Abstract

Biological motion perception, defined as the ability to retrieve information from minimal displays of animal motion, has often been discussed as though it represents a single mechanism. However, depending on the task and details of the stimuli, there have been divergent results as to whether this ability is primarily based on motion or form processing, whether it relies more on local or global information, and whether the knowledge that is applied is learned or innate. These results can be reconciled by a multi-component framework, with five major components: local motion invariant processing, structure-from-motion, figure-ground segregation, action categorization, and style recognition. Several experiments are reported that are motivated by this framework.

To investigate the sensitivity of the visual system to local motion invariants, performance was tested on a direction discrimination and a detection task with a point-light walker based on real motion-capture data or a synthetic walker created by Cutting (Chapter 2). When the walkers were displayed normally, performance was equal for both stimuli. However, when the walkers were spatially scrambled, the direction could only be determined for the natural walker, demonstrating that the invariant is found in subtle features of the local motion trajectories.

Another experiment examined the difference in attentional distribution due to the task requirements (Chapter 3). Participants looked more often at the feet in a direction task than they did in a gender task, lending support to the idea that useful information can be derived from the local motion of the feet.

Finally, Chapter 4 describes a battery of psychophysical tests that assesses each of the components of biological motion perception as independently as possible. The fact that individual
performance does not correlate between tests is evidence that they measure different underlying mechanisms, supporting the multi-component framework.

In the concluding chapter, multi-component computational models of biological motion perception are evaluated within this framework, and the evidence relating the components to activity in particular brain regions is reviewed.
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Co-Authorship

I. Co-Authorship Declaration

In all cases, the data analysis, interpretation and manuscript preparation were performed by the author. The experiments of Chapter 4 were designed and conducted by the author. The experiment of Chapter 3 was designed by the author and carried out by the co-author.

II. Declaration of Previous Publication

This thesis includes three original papers that have been submitted to peer reviewed journals, as follows:

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Chapter 1
General Introduction

1.1 The significance of biological motion perception

In 1973, the Swedish psychologist Gunnar Johansson observed that humans are strikingly sensitive to the motion of living creatures. When exposed to movies consisting of only a handful of light patches tracking the major joints of a moving person, all observers had a vivid impression of seeing a human figure. These point-light displays of human motion are the foundation of the field of biological motion perception, and represent a significant contribution to the study of vision. Like faces, they are complex social stimuli that can yield rich percepts from relatively small amounts of information. Unlike still images of faces, there is a dimension of time, connecting biological motion perception to both basic motion perception and the perception-action system. Although they contain a great deal of social information, point-light displays can be carefully controlled, and so bridge the gap between the gratings and bars used in research on early vision and the high-dimensional visual input present in video footage or the everyday visual world. There has also been keen interest from the computer vision community, because of the many potential uses for information about people’s activities that could be automatically extracted from video cameras (Aggarwal & Cai, 1999; Poppe, 2010).

From the first studies of biological motion perception, researchers have attempted to describe the processes that make it possible. The proposed mechanisms can be organized according to three attributes: the emphasis on the motion information in the display as compared to the form information, the emphasis on global or spatially local information, and whether the knowledge that is involved is learned or innate. The fact that experimental results vary for these attributes depending on the task and details of the stimuli is suggestive that biological motion consists of multiple phenomena, reflecting different underlying mechanisms.
1.2 The role of motion and form information

A topic of great interest in biological motion perception has been whether it relies more on motion information or form information. The point-light stimulus might seem to be inherently dynamic, and Johansson (1973) reported that naïve observers did not interpret static frames of point-light displays as human figures. However, later studies found that experienced observers could see humans in single frames (Cutting, Moore, & Morrison, 1988; Thirkettle, Benton, & Scott-Samuel, 2009). When the elements of biological motion displays are manipulated in ways that are known to diminish motion sensitivity, such as randomly changing the polarity of the dots or representing the elements as changes in texture or contrast, many perceptual tasks are unaffected (Aaen-Stockdale, Thompson, Hess, & Troje, 2008; Ahlstrom, Blake, & Ahlstrom, 1997). Conversely, unlike simple motion detection, biological motion perception is harmed by low-light conditions (Grossman & Blake, 1999). From the neuroscience literature, there exist brain lesion patients who are capable of perceiving biological motion but not simple motion (McLeod, Dittrich, Driver, Perrett, & Zihl, 1996; Vaina, Lemay, Bienfang, Choi, & Nakayama, 1990), and some fMRI studies have found more consistent activation in form processing areas in response to biological motion than in motion processing areas (Michels, Lappe, & Vaina, 2005; Peelen, Wiggett, & Downing, 2006; Thompson, Clarke, Stewart, & Puce, 2005).

The role of local motion, in the context of retrieving the global organization of a point-light walker, was addressed directly by Beintema and Lappe’s (2002) experiments using a stimulus they called the sequential position walker. It is derived from a point-light display of human walking, but instead of tracking the joints, the dots are placed at random positions on the limbs, and frequently disappear and reappear at a different location. In one condition, dots were relocated between each frame, and yet observers still identified the display as a human walking despite the absence of local motion signals, and performed at near-normal levels in determining
the walking direction, among other perceptual tasks. When the lifetime of the dots was increased, enhancing the motion signal, no improvement in performance was detected. The effectiveness of the sequential position walker for producing an impression of human action was confirmed by further studies using this display (Beintema, Georg, & Lappe, 2006; Kuhlmann, de Lussanet, & Lappe, 2009; Michels et al., 2005).

These results, showing that continuous local motion is not necessary for perceiving biological motion, led Beintema, Georg and Lappe (2006) to the view that form is of greater importance, and that the only role of motion signals is to segregate the point-light walker from masking noise. Once it is segregated, biological motion is recognized using a collection of form templates, with evidence for a particular interpretation accumulating over time. This view was implemented in a computer model (Lange, Georg, & Lappe, 2006; Lange & Lappe, 2006), in which time was represented as a learned ordering of static form templates resembling frames of an animation. Approaches along these lines have been taken in the computer vision applications with some success (for a review, see Aggarwal & Cai, 1999).

There is also evidence that motion-dependent processes are important for interpreting biological motion. As in simple motion detection, when the dots that define it are isoluminant the figure is much less effective (Garcia & Grossman, 2008). Playing a point-light display of human motion back at non-naturalistic speeds makes perceptual judgments difficult or impossible (Barclay, Cutting, & Kozlowski, 1978; Beintema, Oleksiak, & van Wezel, 2006), as does adding a blank interval between frames (Mather, Radford, & West, 1992; Thornton, Rensink, & Shiffrar, 2002). In fMRI studies, point-light walkers activate the motion area hMT/V5+ (Bonda, Petrides, Ostry, & Evans, 1996; Grossman et al., 2000), and the posterior superior temporal sulcus (Grossman et al., 2000; Peuskens, Vanrie, Verfaillie, & Orban, 2005; Santi, Servos, Vatikiotis-Bateson, Kuratate, & Munhall, 2003; Vaina, Solomon, Chowdhury, Sinha, & Belliveau, 2001),
which receives input from the motion-driven dorsal stream as well as the ventral stream (Felleman & Van Essen, 1991).

Assuming that the motion in point-light displays is important for interpreting them, how is this information used? The main importance of the motion of the dots might be to define the three-dimensional articulated body structure. Once the viewer has recovered this structure, the movement can be interpreted in terms of the deformation of it over time. The initial inference of the rigid segments and joints implicit in a point-light display has been called structure-from-motion, after the ability to recover the 3-D structure of a fully rigid object when it is presented from several views (Ullman, 1979), although disambiguating a piecewise-rigid object like the human body is a somewhat different problem. Johansson (1973; 1976) and Cutting (1981) described how the rigid connections between point lights could be recovered by determining whether the motion of one joint could be more simply described within the reference frame of another. These proposals did not reach the level of detail of a computational algorithm, but later attempts did (Hoffman & Flinchbaugh, 1982; Webb & Aggarwal, 1982), although their analysis revealed that additional constraints were needed for the structure to be uniquely specified.

Outside of the context of retrieving the global shape, recent studies have revealed that the motion trajectories carry valuable information on their own. When the form is destroyed by spatially scrambling the location of the motion trajectories, the display is still sufficient to determine the walking direction (Chang & Troje, 2008; Troje & Westhoff, 2006), create an impression of animacy (Chang & Troje, 2008), and even give reliable cues to the emotion being portrayed (Chouchourelou, Matsuka, Harber, & Shiffrar, 2006). A different source of motion-based information is the opponent motion signal created when the dots representing the feet cross, which some have argued is used for biological motion detection (Casile & Giese, 2005; Thurman & Grossman, 2008).
There is also evidence that human observers are sensitive to the dynamics implicit in a point-light animation, and their consistency with real-world dynamics (Runeson, 1994; Runeson & Frykholm, 1983). Point-light gaits played back at different speeds give different impressions of the size of a walking creature, that are consistent across participants (Jokisch & Troje, 2003), and observers can identify the correct playback speed for a normal gait (Giese & Lappe, 2002). Shipley (2003) created point-light displays from motion capture recordings of people walking on their hands and found that these were more easily perceived in noise than the same displays inverted, which had a more familiar form but unrealistic dynamics. And Bingham, Schmidt, and Rosenblum (1995) found that observers could distinguish animate from inanimate events when they were represented by single dot trajectories. Consistency with realistic dynamics may be what makes the local motion of the feet compelling and informative, since these trajectories become much less useful when inverted (Chang & Troje, 2009a; Troje & Westhoff, 2006). It appears that the human visual system is tuned to certain types of motion that reflect gravity, inertia, friction, muscle performance, and other constraints of the natural world.

Some studies have directly compared the contribution of motion and form to specific tasks. Mather and Murdoch (1994) set dynamic cues of a point-light display in opposition to structural cues in a gender discrimination task and found that the dynamic cues dominated. Similarly, Troje (2002) compared the effect of removing either the form or the motion information that specified gender from a point-light display, and found that the loss of motion information harmed gender discrimination more than loss of structural information. However participants remained above chance with only the structural cues available, at least when the point-light walkers were viewed from the front.

Overall, the evidence suggests that the output of simple motion integration mechanisms such as are found in brain area MT may not be necessary for most biological motion perception
tasks, but that motion still contributes, and carries valuable information about the dynamics of the scene. It is unlikely that an approach based solely on static form templates could fully explain biological motion perception.

1.3 The role of global and local processes

Another important dimension along which proposed mechanisms for perceiving biological motion differ is in whether they are spatially localized or distributed across the display. Results in face perception emphasize the role of global information, which is often referred to as ‘configural’ information in this domain. Configural information has been defined different ways, including as the spatial relations between pairs of features, or as holistic representations that do not decompose the image into features (Maurer, Le Grand, & Mondloch, 2002). One result that supports the importance of configural information is the whole-part advantage (Tanaka & Farah, 1993), in which studied facial features are better recognized when presented in their proper location within a face, even if it is not the originating face. There is also the composite face effect (Young, Hellawell, & Hay, 1987), which describes how participants find it more difficult to recognize the top half or bottom half of faces when they are aligned with the other half of a different face. In both cases, the overall holistic impression interferes with perception of the local features, indicating faces are not recognized in terms of their component parts.

Further evidence for configural processing comes from a widely observed property of face perception, that it is impaired when the face is turned upside-down (for a review see Valentine, 1988). Several experiments manipulated either the configural or featural information in face images to determine which contributed more to this characteristic inversion effect. Searcy and Bartlett (1996) distorted either the spatial relations between features or the features themselves, and found that participants only detected distortions in the configuration when the
faces were presented upright. Freire, Lee, and Symons (2000) swapped facial features between photographs of individuals or changed the position of their facial features, and found that the spatial changes were only detected in an upright face. These and other findings led to the view that only configural processes are orientation-dependent, and so particularly important in face perception.

Therefore some researchers have interpreted the fact that there is a pronounced inversion effect for biological motion (Pavlova & Sokolov, 2000; Sumi, 1984) as evidence that it is primarily recognized through global, configural information, based on the converse proposition that orientation dependency is an indicator of configural processing. However, as Sekuler, Gaspar, Gold, and Bennett (2004) pointed out, logically this is not necessarily the case. Another source of evidence for global processing is experiments where local biological motion features are made uninformative. For example, when point-light displays are placed in a mask constructed from dots that move according to the trajectories of dots in the walker (a ‘scrambled walker mask’), so that the presence of valid local motion does not guarantee the presence of a walker, participants can still perform a detection task (Bertenthal & Pinto, 1994). Further evidence of global processing is that when stick figures of human walking are viewed through apertures, such that the local motion is ambiguous, participants are able to identify it as a person, as they cannot do for a car or a pair of scissors (Shiffrar, Lichtey, & Chatterjee, 1997). Integration across space is essential for these abilities to be possible.

However, the local motion of dots in point-light displays can also contribute valuable information, and this can be affected by inversion. Troje and Westhoff (2006) found that even when the configuration of a point-light walker was completely disrupted by randomly relocating the dot trajectories, observers could still judge the direction of walking, but only when the display was presented in an upright orientation. This implies that there is a distinct inversion effect for
local motion features, unlike facial features, and therefore that there are at least two process at work, one operating on local motion and one more globally. Chang and Troje (2009a) showed participants pairs of dots following the same trajectories as feet, and there was an inversion effect for judging the walking direction implied by these dots. The fact that visually-inexperienced baby chicks show an interest in a scrambled point-light display of a hen, but not when it is inverted (Vallortigara & Regolin, 2006) also suggests the existence of a local inversion effect. An alternative type of local motion that may also play a role, but is not affected by inversion, is the opponent motion signal that is created when two point-lights representing the hands or feet cross each other in walking viewed from the side (Casile & Giese, 2005; Thurman & Grossman, 2008), although the fact that scrambling the phase of the dots does not affect performance is an argument against the importance of this feature (Troje & Westhoff, 2006).

Between these two scopes, that is, the whole walker or individual dot trajectories, there are processes that operate on an intermediate scale, dealing with relations among groups of dots. Structure-from-motion uses pairings of points to construct subconfigurations that could be recognized on their own or built into larger figures. Pinto and Shiffrar (1999) found that when groups of points representing limbs were randomly placed in a noise mask, they could still be detected at about the same rate as intact point-light walkers, and were spontaneously described as human figures (also see Neri, 2009). Movement of point-light limbs in isolation can portray socially-relevant information, such as the emotion being portrayed by an actor performing a drinking or knocking action, based only on dots representing the arm and head (Pollick, Paterson, Bruderlin, & Sanford, 2001), or animacy and the type of gait from a small number of dots representing only the legs and feet (Todd, 1983).

To identify the spatial distribution of information that is used in a direction discrimination task, Thurman, Giese and Grossman (2010) employed ‘bubbles’, meaning spatial
and temporal samples of a display. The feet and arms were highlighted as particularly important, specifically when they were crossing the midline (creating an opponent motion signal). However, in a study which presented walkers in white noise and constructed classification images based on performance on a forward-backward walking task, Lu and Liu (2006) came to the conclusion that information from all of the point lights in the display contributed to performance. Besides the difference in the tasks that were used, these differing conclusions may have been the result of details of the mapping technique, since bubbles emphasize local features whereas classification images emphasize holistic information.

From the studies described above, it is apparent that there is a double dissociation in the displays that are capable of producing an impression of biological motion: some contain only local motion information, while others contain only global information. This suggests multiple processes, some of which have a narrow spatial focus while others integrate information across the entire figure.

1.4 The role of knowledge

A third way to group biological motion perception processes is by the nature of the knowledge they rely on. Structure-from-motion at first does not appear to require any knowledge about the human body or its characteristic motion patterns, but promises to reconstruct any piecewise-rigid object, with biological motion only representing an instance of such an object (Cutting, 1981; Johansson, 1973, 1976). However upon further analysis, extra assumptions that apply particularly to the human body were required to fully disambiguate a figure, such as the fixed axis assumption (Webb & Aggarwal, 1982) or the planarity assumption (Hoffman & Flinchbaugh, 1982). Although these are simple, bottom-up constraints, they represent domain-specific knowledge.
More detailed knowledge is clearly required for certain demonstrated biological motion abilities. People can recognize individuals by their walks as represented in point-lights (Cutting & Kozlowski, 1977; Hill & Pollick, 2000; Loula, Prasad, Harber, & Shiffrar, 2005; Westhoff & Troje, 2007), and short point-light animations are sufficient to portray gender (Barclay et al., 1978; Kozlowski & Cutting, 1977; Mather & Murdoch, 1994; Troje, 2002) emotional state (Clarke, Bradshaw, Field, Hampson, & Rose, 2005; Dittrich, Troscianko, Lea, & Morgan, 1996), and even the intention to deceive (Runeson & Frykholm, 1983; Sebanz & Shiffrar, 2009). All these abilities can be grouped under the heading of style recognition, and depend on complex, top-down representations that are the product of experience, unlike structure-from-motion. This is also true for the ability to determine the action category from point-light displays (Dittrich, 1993; Johansson, 1976; Prasad & Shiffrar, 2009).

What form does the knowledge take that is used in style recognition and action categorization? Some evidence suggests that it is a type of expert pattern recognition, as described for face perception and other types of visual expertise, for example perception by professional dog show judges of their specialist breed (Diamond & Carey, 1986). Detection of point-light walkers in noise improves with practice (Grossman, Blake, & Kim, 2004) and visual experience (Giese & Poggio, 2003), although this ability may be primed by the evolutionary relevance of biological motion, since people have difficulty learning arbitrary motion patterns (Hiris, Krebeck, Edmonds, & Stout, 2005; although see Jastorff, Kourtzi, & Giese, 2006). Only experts in basketball could distinguish a real from a feigned pass (Sebanz & Shiffrar, 2009). Also in common with other types of visual expertise, when motion style is exaggerated it is easier to recognize the individual (Hill & Pollick, 2000) and the action category (Pollick, Fidopiastis, & Braden, 2001).
However, there is reason to think that the knowledge that is used in interpreting local biological motion trajectories is of a different nature, which is innate and evolutionarily old rather than the product of learning. Movement of individual dots representing animal motion are compelling to baby chicks with no previous visual experience (Vallortigara & Regolin, 2006; Vallortigara, Regolin, & Marconato, 2005), and to human infants within the first few days of life (Méary, Kitromilides, Mazens, Graff, & Gentaz, 2007; Simion, Regolin, & Bulf, 2008). Processing of local motion also appears to not be affected by learning (Chang & Troje, 2009b). This predisposition towards certain types of dynamics and visual stimuli resembles that found in other domains, such as the result that infants preferentially gaze at face-like objects (Johnson, Dziurawiec, Ellis, & Morton, 1991); the research on ‘core knowledge’ that infants have of constraints on the motion of inanimate objects (Spelke, Breinlinger, Macomber, & Jacobson, 1992); and the finding that lab-raised rhesus monkeys could be conditioned to fear the sight of toy snakes but not toy rabbits (Cook & Mineka, 1989).

This type of innate perceptual knowledge could be entirely in the form of bottom-up perceptual filters for invariants in the local motion. These invariants would capture aspects of the dynamics discussed in section 1.2, such as the influence of gravity, that correlate with the presence of a legged, terrestrial animal, making it function as a ‘life detector’ (Johnson, 2006; Troje & Westhoff, 2006). The body-based constraints that make structure-from-motion possible could also be a case of this type of knowledge.

In summary, although knowledge is required for all biological motion processes, the form of the knowledge that is used to accomplish certain tasks may be quite different, and may not necessarily require learning.
1.5 A hierarchy of biological motion processing

Does biological motion perception depend more on motion or on form processing? Is it more spatially localized or distributed? Is the knowledge it uses innate or acquired through learning? The varying answers to these questions that different experiments have provided can be reconciled by a framework in which there is not one phenomenon of biological motion perception, but several, representing multiple underlying components. Each distinct biological motion task calls on a different mixture of these components.

The results reviewed above suggest that there are at least five components (Troje, 2008), which are more or less distinct and have different levels of reliance on motion, localization, and innate knowledge:

1. Processing of local motion invariants: Purely motion-based; purely local; innate knowledge.
2. Structure-from-motion: Motion-based; localized but not as much as the previous component since it is based on relations between dots rather than single dot trajectories; knowledge takes the form of limited constraints which may be innate or learned.
3. Figure-ground segregation: Motion-based; in the case of a scrambled walker mask, dependent on global processing; both learned and innate knowledge may contribute.
4. Action categorization: Unknown whether it is more motion or more form-based; likely more global than local; learned knowledge.
5. Style recognition: Both form and motion contribute; all levels of localization; learned knowledge.

These components may be regarded as a hierarchy of processing levels. When a living creature enters the visual field, the presence of invariants in the motion of their extremities (1.)
alerts the viewer and perhaps primes the other components. Then a structure-from-motion mechanism (2.) recovers the form of the creature. Once the form is obtained, the motion can be described in terms of the form, assisting in categorizing the action that is being executed (4.). Subsequently the style of the action (5.) can be recognized based on the possibilities inherent in the action category. Many of these components can also operate independently of each other, for example recognizing style from local motion trajectories without recovering the structure (Chouchourelou et al., 2006), but the output of earlier stages can assist in the judgments of later stages. The transition between stage (4.) and (5.) resembles the hierarchy of object categorization described by Rosch and colleagues (1976), in which objects are first categorized at their basic level (in this case the action being performed, such as ‘dancing’), after which they can be categorized at the subordinate level (the style of the action, an adverb such as ‘angrily’). However, the local motion ‘life detector’ bypasses this system to provide superordinate category information, that a living creature is present.

The purpose of the studies that make up this thesis is to better define the nature of the processing hierarchy and its components. Chapter 2 assesses the specificity of the local motion invariant detectors. Chapter 3 explores the localization of information in a style recognition task compared to the lower level biological motion task of direction discrimination. Chapter 4 investigates the relations among the proposed components, using a battery of tests that assesses individuals’ abilities on all these components in one session, making it possible to use individual differences to test for dependencies.

1.6 Specificity of local motion invariant processing (Chapter 2)

Although the ability to extract information from local motion trajectories of point-light displays is established, as well as the fact that the majority of local information about walking
direction is in the foot motion (Troje & Westhoff, 2006), it has not been clear how important the
details of the local motion are to activating this processing stage. What level of distortion can it
tolerate? This is a question of practical importance in the laboratory, since a type of point-light
display that has been in widespread use in biological motion psychophysics since its publication
(Cutting, 1978a, 1978b) has dot motion that is significantly different from motion-captured
walking. It is synthesized from pendular and sinusoidal functions that have been adjusted by eye
to resemble a walker, an approach that was criticized by Runeson (1994) as failing to capture
important features of real human motion.

Chapter 2 presents evidence that the local motion of this display lacks the characteristic
invariants that allow human observers to determine direction and other properties even when the
global structure is disrupted by scrambling. Participants were at chance for determining direction
from a scrambled walker prepared according to Cutting’s algorithm, whereas participants were
above chance for determining the direction of a scrambled walker derived from motion capture
recordings. However when the structure was intact, such that structure-from-motion or other
global processes would be capable of interpreting it, direction performance using Cutting’s
walker was equivalent to the more naturalistic stimulus, as was detection performance.

The conclusion from this study is that observers are sensitive to details of the local
motion when it is the primary source of information, and that the vertical acceleration of the feet
is likely part of the invariant since it is one of the most prominent differences between the two
types of displays. On the other hand, structure-from-motion does not appear to be as sensitive to
distortions in the local motion pattern. This contrast suggests two distinct processes with different
characteristics.
1.7 Localization of information for different components (Chapter 3)

The components that are engaged for different tasks differ in their degree of localization, and for those that do not process the entire display in parallel it is useful to investigate which regions contain the relevant information for a task. The identified regions are clues to how the task is accomplished. Since different tasks emphasize different components of biological motion perception, we would expect that attention will be differentially directed to the areas that are most informative for the relevant component, and that this difference in attentional focus will be reflected in eye fixations.

Chapter 3 investigates the localization of information in the display for different components, by tracking participants’ gaze while they performed a direction task and a gender task, with identical stimuli in both tasks. For the direction task, the typical viewer fixated more frequently at the feet, which suggests that they were making use of the local motion invariants there. By contrast, in the gender task there were relatively less concentration of fixations at the feet and more fixations at the shoulders, which is congruent with previous research on diagnostic regions of a point-light walker for determining gender.

1.8 Assessing biological motion perception components (Chapter 4)

The studies in chapters 2 and 3 are based on the assumption of separate components involved in biological motion processing. Besides the distinct characteristics that are revealed in those studies, such as the sensitivity of local motion processing to velocity features and the different information sources used by local invariant extraction and style recognition, it was necessary to further test the hypothesis of separate components and to develop methods to assess them.
Chapter 4 introduces an online battery of psychophysical tests, called BMLtest, that efficiently measures multiple components of an individual’s biological motion perception in the same session. Literature is reviewed showing that many commonly used biological motion tasks confound measurement of more than one of these underlying components, and this has led to apparently conflicting results. The tests use tasks that are designed to measure the components as independently as possible. Each of the levels described in section 1.5 has a corresponding test based on point-light or stick figure stimuli, with two different style recognition tests: one that assesses gender discrimination ability and one that assesses person recognition.

Besides providing a resource for future biological motion studies, in particular studies characterizing the abilities of special populations, BMLtest can be used in individual differences analyses to test for functional dependencies between processes. 65 young adults completed the battery of tests. Since there were no significant correlations, it could be concluded that the underlying constructs measured by each test were relatively disjoint.

The studies described in these chapters are situated within the framework of biological motion perception as a hierarchy of processing components, and provide support for the framework. Instead of using a single task to measure biological motion perception, assessments should target a specific component, whether it is local motion invariant processing, structure-from-motion, figure-ground segregation, action categorization or style recognition. These processes differ in terms of their use of motion information, of global and local information, and of learned or innate knowledge.
References


Chapter 2
Specificity of local motion invariant processing

2.1 Introduction

Humans have a striking ability to detect and extract information from the motion of living creatures. A major tool to study biological motion was introduced into visual psychology by Johansson (1973). His point-light displays, which consist of only a few dots moving along with the major joints of a human in motion, elicit a clear percept of actor and action without being confounded with other cues, such as clothing or facial information. The ease with which the human visual system retrieves information from these degraded displays reflects a number of different processing levels. Troje (2008a) has suggested that the complex phenomenology of biological motion perception can only be understood if we carefully distinguish between two different processes in particular. One uses local motion signals directly, in particular for tasks such as detecting the presence of an animal or inferring its orientation, while another integrates the articulated motion of a body to infer its structure and how that structure deforms over time.

The majority of work in the field of biological motion perception has focused on this second process. Early studies proposed models for how hierarchical structures of limbs and joints could be determined by subtracting motion trajectories from one another (Cutting, 1981; Johansson, 1973, 1976). Later it was shown that the global form can be recovered even when the local motion of the dots is rendered entirely uninformative. Point-light displays of walking human figures can still be detected, at least to some degree, when masked by randomly displaced copies.

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of the motion of the dots making it up (Cutting, Moore, & Morrison, 1988), indicating that the
global form is sufficient for detection. On the other hand, this ‘scrambled walker mask’ did prove
to be a much more effective mask than any of the other types of masks used in the study. The loss
of local motion as a cue to the presence of a walker caused a substantial decrease in performance,
suggesting that it does play an important role.

Further experiments with the scrambled walker mask showed that the global structure of
the point-light display is sufficient to determine the direction of walking, and to show the
pronounced inversion effect that seems to be characteristic of biological motion processing
(Bertenthal & Pinto, 1994). Similar results were observed when local motion patterns were
eliminated in a different way, namely by randomly repositioning the dots on the body in every
frame of a motion sequence (Beintema, Georg, & Lappe, 2006; Beintema & Lappe, 2002; Neri,
Morrone, & Burr, 1998).

In contrast to this emphasis on deriving the global, articulated structure from the point-
light display, some researchers have asked whether there might be valuable information used by
the visual system in the local motion alone. Mather, Radford and West (1992) tested performance
on a direction task (asking participants whether the walker was facing left or right) when sets of
dots were removed from the point-light display. Removal of shoulder and hip dots, or elbow and
knee dots, had no effect on performance (which was approximately 90% correct), but removing
the wrist and ankle dots while leaving all other dots resulted in a reduction of performance to
near-chance levels. Troje and Westhoff (2006) also found evidence that motion of the extremities
is important in biological motion perception. However, while the stimuli used by Mather et al.
(1992) still retained most of the coherence of their shape, Troje and Westhoff used spatially
scrambled walkers in their study which entirely lacked coherent global form. They found that
even though observers had no idea what kind of creature they were seeing, they had a clear
percept of facing direction – but only if the displays were shown in upright orientation. A subsequent experiment showed that this inversion effect was dependent only on the orientation of the trajectory of the ankle dots, while no effect resulted from inverting any of the other trajectories. The results were not limited to human motion but were also obtained with point-light displays derived from a walking pigeon and from a walking cat.

Troje and Westhoff (2006) suggested that cues contained in the local motion of an animal’s feet may play a role as a general detection system that signals the presence of a terrestrial, articulated animal in the visual environment (see also Johnson, 2006). They hypothesized that such a system might be evolutionary old and possibly innate rather than being based on individual learning. Some evidence for this idea has since been demonstrated. For instance, Vallortigara and Regolin (2006) have shown that newborn chicks already respond preferentially to upright, but not to inverted, biological motion, but that the particular shape of the point-light animal (cat or chicken, coherent or scrambled) has no effect on this preference behaviour (Vallortigara, Regolin, & Marconato, 2005). Simion and colleagues (2008) have shown that 2-day-old human neonates also prefer upright biological motion over inverted biological motion. They used point-light displays of chicks, and therefore a shape which is probably as meaningless to a 2-day-old infant, as a scrambled display. Finally, Chang and Troje (2008) showed that adult observers consistently rated scrambled point-light displays (of humans, pigeons and cats) as more animate when presented upright rather than inverted.

In all these studies, observers responded either to scrambled biological motion, or to biological motion carried by structures they are entirely unfamiliar with. All they could use was the local motion of the trajectories of individual dots, most likely the dots representing the feet. Since such local motion alone can specify the direction of walking, affect the impression of animacy, and control preference behaviour of chicks and human neonates, we must consider the
processing of local motion signals as an essential component of biological motion perception. Destroying the structure by spatially scrambling the walker cannot therefore be considered a means to eliminate the biological aspect of the motion, as has been assumed in many psychophysical studies (e.g. Fujimoto, 2003; Thornton, Pinto, & Shiffrar, 1998; Verfaillie, 2000) and functional imaging studies (e.g. Vaina, Solomon, Chowdhury, Sinha, & Belliveau, 2001).

One reason why the important role of local motion has been overlooked for a long time relates to the way many researchers in the field have generated their biological motion stimuli. In 1978, James Cutting published a paper in which he described a simple algorithm that would create a point-light display of a human walking in sagittal view. He was inspired by work in speech recognition that involved synthesizing auditory stimuli to determine which features produce a particular judgment in the listener about what phoneme they are hearing, and wanted to create a point-light stimulus that could be manipulated in a similarly principled way. Using his walker stimulus, Cutting showed that changing the synthesized relative motion of the hips and shoulders affected observers’ judgment of the gender of the figure, providing evidence for the earlier conjecture (Kozlowski & Cutting, 1977) that the center of moment is the critical invariant which discriminates between male and female walking. Cutting’s walker was a breakthrough for biological motion research because allowed for both ease of generation of biological motion displays in laboratories around the world and for unprecedented control over the spatial and temporal characteristics of the display. The walker was generated by a FORTRAN program which animated each of the 13 dots with some combination of ellipsoid and pendular motion, defined within the hierarchical organization of an articulated body. The trajectories of shoulders and hips were modeled as simple ellipses, while the elbow and knee motions were modeled as pendulums with the motion of the shoulders and hips, respectively, added to them. He qualitatively compared the resulting motion pattern to existing knowledge about human gait.
(Murray, 1967), but the final parameter settings were chosen by eye. Many important results have been established with this display, and even 30 years after its publication Cutting’s walker is still widely used (e.g. Hiris, 2007; Lange & Lappe, 2007; Neri et al., 1998; Saygin, Driver, & de Sa, 2008; Thornton & Vuong, 2004).

However, while the general shape of Cutting’s synthetic walker resembles point-light displays derived directly from human walking, it differs in important ways from recordings of real human motion, particularly in the critical invariants contained in the kinematics of the feet (see Figure 2.3). Runeson (1994), among others, warned that such synthesized displays may be missing information about dynamics, in particular the effect of gravity and the ways the organism’s musculoskeletal system responds to it. Given the role that the feet seem to play (Mather et al., 1992; Troje & Westhoff, 2006), we expect that these differences seriously affect the invariants in the local motion that convey direction and animacy in more natural displays. We assume that the wide usage of Cutting’s walker is also the reason why the presence of these cues in normal scrambled biological motion displays was overlooked for such a long time.

In the first experiment of the present study, we compare the performance of human observers in a direction discrimination experiment between Cutting’s walker and a walking display derived from recorded human motion, which we will refer to as the BML (BioMotionLab) walker (Troje, 2008b). Both coherent and scrambled versions of the walkers were used, but the critical condition is the one in which scrambled walkers are presented and so decisions must be based entirely on local motion cues.

Whereas in the first experiment we presented local motion without global motion, in the second experiment we chose a paradigm that makes available global motion but not local motion. Specifically, we employed a detection task, with coherent versions of the walkers embedded in a
mask made of scrambled walkers. Differences as to the presence of the critical cues in the local motion should only affect Experiment 1, and, specifically, the condition in which scrambled walkers are used.

2.2 Methods

2.2.1 Participants

Twelve people participated in the experiment (four male, eight female, age 17-44 years, mean = 20.2 years), all had normal or corrected-to-normal vision. All but one subject had no experience with biological motion and point-light stimuli, and all were naïve with regard to the purpose of the study. All but one were undergraduate students at Queen’s University. They received course credit for their participation.

2.2.2 Stimuli

Displays consisted of point-light walkers presented in the center of a 17-inch CRT monitor, operated with a 100 Hz refresh rate. Each point-light display subtended 6° of visual angle vertically, with the observer’s viewing distance fixed at 80 cm by means of a chinrest. The displays consisted of 11 white dots on a black background representing a sagittal view of the head, elbows, wrists, knees, and ankles, as well as one shoulder and one hip, of a human walking without forward displacement, as if on a treadmill. Gait frequency was 0.93 Hz.

Scrambled walkers were produced by subtracting from each trajectory its mean position, and then adding to it a random constant position which would place it within a rectangular range of 4° × 6° of visual angle. In addition, the trajectory was also randomly offset in time. The result was that the local motion of each dot’s trajectory remained unchanged, while its position within the display and its temporal phase with respect to the other dots was randomized.
Two different masks were used. The scrambled mask was based on a similar technique used to create the scrambled walker. The only difference was that depending on the required number of masking dots, multiple copies of the scrambled walker were used, and the dots comprising these walkers were randomly distributed over a rectangular range of $9.6^\circ \times 9.6^\circ$ of visual angle. Scrambled masks were always made from the same walker (BML walker or Cutting walker) they were intended to mask. The random dot mask was produced by randomly placing stationary dots within this $9.6^\circ \times 9.6^\circ$ mask area. Each dot was displayed for 300 ms at a given location, after which it disappeared and then reappeared elsewhere.

The point-light walkers were generated by two different methods.

**Cutting’s walker.** The animation of Cutting’s walker was created following the algorithm published in Cutting (1978).

**The BML walker.** The BML walker was produced from the walks of 50 women and 50 men obtained by means of a motion capture system. Participants were instructed to walk at a normal pace in a straight line for several steps. Each participant had 41 reflective markers attached to his or her body, which were tracked in 3-D space by an array of nine high-speed (120 Hz) cameras making up part of a motion capture system (Vicon 512, Oxford Metrics). A biomechanical model of the human body was then applied to estimate the movement over time of 13 ‘virtual markers’ located inside the body. They included the joint centers of the shoulders, elbows, wrists, hips, knees, and ankles, as well as the center of the head. One of the shoulder and one of the hip points were discarded, resulting in an 11-dot display with each point corresponding to a point in Cutting’s walker. The forward component of the movement was subtracted. Each trajectory was then transformed into a discrete Fourier expansion. All but the average position and the first two harmonics, which together represented 98% of the variance, were discarded,
resulting in a smoothed, continuous walking loop. The final BML walker was obtained by averaging these Fourier representations over all 100 individuals (for details see Troje, 2002, 2008b).

2.2.3 Procedure

All participants completed both the detection task and the direction task in two separate blocks. The order of the two experiments was counterbalanced, so that half of the participants started with the direction task first and the other half started with the detection task. Before each experiment participants read the instructions and then completed several practice trials to get familiar with each task.

**Direction task.** In each trial either the BML walker or Cutting’s walker was displayed facing to the left or to the right. The walker was either presented in its coherent shape or was scrambled. We included trials with coherent walkers to make the task more rewarding for our participants. The critical condition, however, was the one employing scrambled biological motion in which only local motion cues can be used to solve the task.

In all cases, walkers were embedded in a random-dot mask consisting of 50, 200, or 750 dots. Participants were instructed to press the left or the right arrow key as quickly as possible to indicate the direction in which the walkers seemed to face. Each trial was shown until a key was pressed, or until a maximum duration of 1 s. Between trials, the screen remained blank for 1 s. No feedback was provided. There were 20 repetitions for each cell in our design, resulting in a total of 240 trials (2 walkers × 2 directions × 3 mask densities × 20 repetitions).

**Detection task.** A two-alternative forced-choice design was used. Each trial consisted of a pair of displays shown sequentially. Each display was presented for 1 s and contained a scrambled walker mask with 50, 100, or 250 masking dots. In addition, either the first or second
interval also contained a coherent walker – either the BML walker or Cutting’s walker. Walkers were shown in sagittal view, facing randomly either to the left or to the right. Participants pressed the left arrow key to indicate that the walker was in the first interval, or the right arrow key to indicate the second interval. There were 20 repetitions for each condition, for a total of 240 trials (2 walkers × 2 intervals × 3 mask densities × 20 repetitions).

2.3 Results

2.3.1 Direction task

The results of this task are shown in Figure 2.1. For coherent walkers, direction errors were low and similar between the two walker types. For scrambled walkers, direction errors were generally higher. In this condition, performance with Cutting’s walker was at chance, while error rates fell below chance level for the BML walker. Mask density seemed to affect the coherent walkers much more than the scrambled versions. A three-way repeated-measures ANOVA conducted on error rates confirmed these observations. It showed significant main effects of walker type,
There was also a significant interaction between walker type and walker scrambling, $F(1, 11) = 9.56, p < .05$, and a significant interaction between scrambling and mask density, $F(2, 22) = 4.13, p < .05$. All other interactions were non-significant. A simple main effects analysis was conducted to assess the effect of walker type within the scrambled walker and coherent walker conditions. There was no significant difference between walker types in the coherent condition, $F(1, 11) = 0.44, p = .52$, but there was a significant difference between walker types in the scrambled condition, $F(1, 11) = 16.02, p < .01$, confirming that the scrambled BML walker still retained information about its facing direction while Cutting’s walker did not. A one-sample $t$ test was conducted for Cutting’s walker within the scrambled condition. The average error rate was not significantly different from 50%, $t(11) = 0.84, p = .42$. The interaction between walker scrambling and number of masking dots was due to a significant simple main effect of number of masking dots within the coherent walker condition, $F(2, 22) = 32.54, p < .001$, but there was no effect of number of masking dots in the scrambled walker condition, $F(2, 22) = 2.36, p = .118$. The same pattern of results was seen in a three-way repeated-measures ANOVA performed on $d'$ values (significant effects of walker type, walker scrambling, and number of masking dots, with an interaction between walker type and walker scrambling, $F(1, 11) = 7.98, p < .05$, as well as between walker scrambling and number of masking dots, $F(2, 22) = 6.18, p < .01$, with the average $d'$ for the scrambled Cutting walker not significantly different from 0, $t(11) = 0.86, p = .41$).

### 2.3.2 Detection task

The results of this task are shown in Figure 2.2. Larger numbers of masking dots caused more errors, but there was no difference between the BML walker and Cutting’s walker. A two-way ANOVA conducted on error rate confirms this observation. There was a significant effect of
Figure 2.2. Mean error rates across twelve participants for the detection task. Coherent versions of either Cutting’s walker or the BML walker were presented in a scrambled dot mask. The error bars represent ±1 SEM.

number of masking dots, $F(2, 22) = 28.55, p < .001$, but no other main effect or interaction. The same pattern of results was seen in a two-way repeated-measures ANOVA performed on $d'$ values (effect of masking dots, $F(2, 22) = 28.89, p < .001$, effect of walker type, $F(1, 11) = 0.02, p = .89$)

2.4 Discussion

Our results demonstrate that Cutting’s (1978) walker lacks the cues that signal facing direction in scrambled displays derived from recordings of natural locomotion patterns such as the BML walker. Performance on the detection task, where only the global shape was available to identify the presence of the walkers, did not differ between the walker types.

What are the differences between the walker types that account for the differences in performance? The most obvious differences are in the feet (as represented by the ankles in the motion capture recordings), which Troje and Westhoff’s (2006) findings suggest are the principal carriers of direction information. A plot of the trajectory of the ankle (Figure 2.3) shows the
Figure 2.3. Plots of the trajectory of the ankle movement of a walker seen in sagittal view and facing to the right. Overall forward displacement was subtracted, leading to a treadmill-like display. A) The ankle trajectories of six randomly selected recordings from the database used to generate the BML walker. B) The ankle trajectory of the BML walker. C) The ankle trajectory of Cutting’s walker.

similarity between the foot movement of the BML walker and that obtained from typical motion capture recordings. The feet of Cutting’s walker have several differences visible in this figure. Where the feet of the BML and the real walkers are stopped abruptly by the ground, the foot of Cutting’s walker appears to extend slightly below the ground. The Cutting foot lacks at least in part the horizontal asymmetry of the teardrop shape of the real foot movement, and assumes a more symmetrical shape, almost approaching a horizontal figure eight. Examining the vertical component of the movement alone, a plot of position, velocity and acceleration over time (Figure 2.4) reveals more differences. During the swing phase, the feet of Cutting’s walker decelerate less sharply and for a longer duration. The longer deceleration of Cutting’s walker also reflects a longer relative duration of the swing phase within the gait cycle.

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For the kinematics along the horizontal axis, which are likely more directly relevant to the direction discrimination task, there are at least two noticeable differences in the foot acceleration between the two types of walkers (Figure 2.5). There is an additional small peak in acceleration at the start of the swing phase of the BML walker’s gait cycle, and a sharper deceleration just after the end of the swing phase, as the heel strikes the ground. Though these differences can be clearly seen in comparing the acceleration plots for the BML walker and Cutting’s walker, they are only subtly visible in the plot of horizontal position. In addition, the motion of the foot of Cutting’s walker on the horizontal axis is more symmetrical with respect to time; by comparison the feet of the BML walker form more of a ‘saw tooth’ pattern, which may assist in cueing direction.

*Figure 2.4. Vertical displacement, velocity, and acceleration of the ankle dot over two full gait cycles for the BML walker and Cutting’s walker. Shaded regions indicate swing phases and white regions indicate stance phases.*
These differences in the kinematics along both the vertical and the horizontal axes are particularly relevant in light of the recent finding by Chang and Troje (2009) that the pattern of acceleration plays a key role in determining the direction of feet motion in isolation. They showed that the apparent facing direction of a point-light display consisting of feet motion only (of the BML walker, but also from point light displays of other animals) could be determined from only brief segments of the motion of the two ankle dots, and that this ability was hampered when the dot motion was presented upside down. However, when the characteristic acceleration profile was removed by replacing the motion with a constant-velocity tracing of the same trajectory, there was no longer an effect of inversion on performance. They concluded that observers make use of the characteristic pattern of acceleration which the feet follow as the walker responds to gravity in

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*Figure 2.5. Horizontal displacement, velocity, and acceleration of the ankle dot over two full gait cycles for the BML walker and Cutting’s walker. Shaded regions indicate swing phases and white regions indicate stance phases.*
an energy efficient way. From the present results, it appears that at least some of the features that characterize the typical acceleration profile of a natural foot are missing in Cutting’s walker.

Johansson (1976) first demonstrated the sensitivity of the human visual system to deviations from normal biological motion, finding that all of his observers could tell that a point-light display was produced by a puppet rather than by a person in less than 500 ms. Runeson and Frykholm (1983) showed that observers can derive the true weight of a box by seeing it lifted, even if the lifter tries to deceive the observer by acting as though the box is heavier or lighter than it actually is. Runeson (1994) argued that to ensure point-light displays contain all the information that may be perceptually relevant, recordings of real motion should be used, and “we must require that even its finer details are intact.” (p. 392) He pointed out that Cutting’s walker has “no involvement of dynamic factors such as mass, elasticity, energetic processes, or neural mechanisms, nor are any of the law of motion taken into account.” (p. 392)

The finding that direction of the scrambled walker could be determined for the BML walker but not for Cutting’s walker may explain a discrepancy between two previous results. Troje and Westhoff (2006) found a 64% accuracy, significantly above chance, for the direction task using the scrambled walker masked by random flickering dots, whereas Lange and Lappe (2007) found that participants in their study viewing a scrambled walker display did not perform significantly better than chance on the direction task. Troje and Westhoff used the BML walker to generate its point-light display, while Lange and Lappe used Cutting’s walker.

Increasingly, biological motion perception studies use motion-capture-derived stimuli (e.g. Ikeda, Blake, & Watanabe, 2005; Jastorff, Kourtzi, & Giese, 2006) as this technology becomes more accessible. New techniques other than conversion to a point-light display are also being explored for controlling the non-motion cues in a scene (e.g. Kuhlmann & Lappe, 2006).
However, there is still a large body of research which is based on Cutting’s walker. For most of these studies, Cutting’s walker is probably an appropriate approximation. In particular, it can be used to study the mechanisms underlying perceptual organization of the individual dots into a coherent shape. However in other cases, the use of Cutting’s walker may have produced results different from the ones a real walker would have revealed. Caution needs to be exercised particularly in experiments in which a scrambled walker is used as a ‘non-biological’ control stimulus in order to eventually conclude that a phenomenon must be based exclusively on coherent structure rather than on local information. For instance, Fujimoto and Yagi (2008) showed that a coherent walker induces apparent motion in a flickering background, but a scrambled walker does not. The difference between the two displays is certainly the presence versus the absence of coherent shape, and it can therefore be concluded that coherent shape is able to induce background motion. However, from this experiment it cannot be excluded that information in the local motion (for instance, of the feet) could also induce background motion. Other examples of experiments that follow a logic that is potentially vulnerable to the particular kind of walker used are the work of Thornton and Vuong (2004) on the effect of coherent and scrambled flanker stimuli, and the work of Saygin, Driver, and de Sa (2008) on sensitivity to audio-visual asynchrony in point-light displays.

Further studies will investigate in detail which aspects of the motion of the extremities are used to determine direction, and what role they play in detection of biological motion, in biological motion perception in the visual periphery, and in perception of animacy. Our conclusion from the present study is that Cutting’s (1978) walker differs in systematic and perceptually relevant ways from real human motion. Invariants contained in the local motion alone, without the possibility of deriving structure from the display, can play an important role in
perception of biological motion, but this role may not be detected in experiments that employ Cutting’s walker since it does not represent these invariants realistically.
References


Chapter 3

Localization of information for different components

3.1 Introduction

Human vision can extract much high-level information from the motion of living creatures, even when all visual cues are removed except for the movement of the major joints. Experiments using these point-light displays have shown that we are sensitive to the gender, heading direction, identity, and even the emotional state and personality of the producer of the motion (Blake & Shiffrar, 2007; Troje, 2008a). However, more remains to be learned about which parts of the display are most informative for these different abilities. Determining the primary areas from which information is gathered will provide important clues to how each perceptual task is performed.

In this study, we observed the effect of differing task instructions on eye movement patterns when viewing the same set of biological motion point-light walker stimuli. We used two commonly studied biological motion tasks: determining the direction in which a walker is facing, and determining the gender of the walker. The facing direction can be derived from different sources. One of them is the articulated structure of the body revealed by the dynamic configuration of subsets of dots (Bertenthal & Pinto, 1994; Troje, 2002). Like the explicit structure of a static stick-figure, the motion-mediated structure of a point-light display clearly indicates whether a sagittal-view figure is facing left or right. A second source of information

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about facing direction is the local motion of individual dots, in particular the ones representing
the extremities. Both Mather, Radford and West (1992) and Thurman and Grossman (2008)
found that removal of central dots in the displays did not harm walking direction judgment,
whereas removal of dots representing the extremities did. Other experiments showed that the feet
are sufficient for the perception of direction by making the overall form of the point-light walker
inaccessible, whether by inverting all the dot trajectories of the walker except for the feet (Troje
& Westhoff, 2006) or by showing the feet in isolation (Chang & Troje, 2009). Although Pinto
and Shiffrar (1999) argued that the importance of the extremities may be limited to determining
direction, the hypothesis they addressed involved the configuration of the extremities rather than
the intrinsic local motion of the extremities, since the local motion was made uninformative by
the study’s use of simultaneous scrambled walker masks. Therefore the study did not contradict
the premise that local motion of the extremities is relevant to more than judging walking
direction.

For the task of judging a point-light walker’s gender, a number of structural and
kinematic cues have been suggested, most involving the shoulders and hips. Murray, Kory, and
Sepic (1970) described greater lateral movement of the hips during walking in women as
compared to men, observations that were later quantified by motion capture studies (Cho, Park, &
Kwon, 2004; Smith, Lelas, & Kerrigan, 2002). Several researchers manipulated structural and
kinematic properties of synthetic point-light displays in order to investigate which inform gender
judgments. Barclay, Cutting and Kozlowski (1978) and Cutting, Proffitt and Kozlowski (1978)
showed that observers required movement in the dots of the walkers to correctly determine
gender, but framed the movement differences as mediating underlying differences in the body
structure, in particular in the relative width of shoulders and hips. However, in another study
using a synthetic walker, Mather and Murdoch (1994) showed that even with structure held

constant, lateral shoulder and hip movement could produce specific gender impressions. In contrast to these studies evaluating individual candidate cues to gender, Kozlowski and Cutting (1977) removed subsets of dots from point-light walkers derived from videotaped motion, and found that participants were above chance at judging gender with every subset. This view of gender information being distributed across the body was supported by Troje’s (2002) creation of caricatured male and female walks using a linear ‘walker space’. However, the differences between the extreme male and extreme female walk were most visible in the frontoparallel movements of hips and shoulders.

The importance of different regions of a visual display to a perceptual judgment can be studied using eye movements. The location of central fixation is a good proxy for the focus of attention (Findlay & Gilchrist, 2003), and it is known that eye movement strategies change depending on the task, with relatively more fixations occurring in regions that are more informative for the task (Buswell, 1935; Castelhano, Mack, & Henderson, 2009; Yarbus, 1967). This technique was used by Buchan, Paré, and Munhall (2007) to explore where people look on a speaking human face when either comprehending the words or judging the emotion being expressed, and how the fixation locations change as the task became more difficult. Johnson and Tassinary (2005) observed gaze patterns for biological motion perception of gender, using solid animated walking figures created with the animation software Poser. Their participants spent the most time fixating on the waist and hips, followed by the chest, followed by the legs and head. However, only the specific candidate cues of shoulder swagger and hip sway were varied, and the walkers were shown only in a frontal view.

We used eyetracking and a point-light walker derived systematically from recorded human movement (Saunders, Suchan, & Troje, 2009; Troje, 2008b), presented with different view angles and different degrees of maleness or femaleness, to investigate where viewers were
gathering information during a direction and a gender biological motion task. The fact that the displays were synthesized from real walking patterns allowed us to avoid preconceived ideas of where directional or gender information was located, while still precisely modulating the availability of relevant cues.

We hypothesized that there would be a difference between the tasks in the number of fixations that fell in different regions, with more fixations at the feet in the direction task as compared to the gender task, and more fixations at the pelvis and shoulders in the gender task as compared to the direction task. We also hypothesized that task difficulty would have an effect on fixations in both tasks, with relatively more fixations being directed to informative locations as the task becomes more difficult. We evaluated these hypotheses using fixation maps and a region-of-interest analysis on the number of fixations. We also looked at individual differences, assuming that individuals would demonstrate different gaze strategies and that some strategies would be more effective than others for each task.

3.2 Methods

3.2.1 Participants

Twenty participants (7 men and 13 women) completed this study, primarily recruited from an introductory psychology course (several other recruited participants withdrew or did not complete due to technical problems). Their ages ranged between 17 and 38 (M = 20) and they were not experienced with biological motion stimuli. Participants were compensated with money or course credit.
3.2.2 Materials

Eye positions during the trials were recorded using a head-mounted EyeLink II system at a 250 Hz sampling rate. It has a gaze position accuracy of < 0.5°. The stimuli were presented on a CRT monitor and controlled by a computer running MATLAB and the Psychophysics Toolbox (Brainard, 1997) which also received eyetracking data.

3.2.3 Stimuli

The stimuli consisted of point-light displays of human walking. They were created from the walks of 100 individuals, including 50 men and 50 women. Walking sequences were obtained using a motion capture system (Vicon 512, Oxford Metrics) and data were converted into Fourier-based representations, before averaging across several walking trials recorded for each individual (Troje, 2008b). We applied principal component analysis (PCA) to the collection of averaged individual walks to reduce the dimensionality of the space, and then used discriminant function analysis to create an axis that best captured gender-specific differences (Troje, 2002). A unique walker was generated for each trial by first choosing the point on the gender axis corresponding to the desired level of maleness or femaleness, and then adding a small random vector in the walker space, orthogonal to the gender axis. This had the effect of producing a novel animation that was a valid-looking walker with the desired gender level. The point-light displays consisted of 15 white dots on a black background, 10 cm tall, subtending 9.5° of visual angle at the viewing distance of 60 cm.

The synthetic walkers were either male or female, with one of three gender levels in units of standard deviations of the original 100 individuals’ gender scores: 0.5, 1.5, and 2.5, where larger values indicate more exaggerated gender cues. The gender level affected all the dots of the walker, changing both static and dynamic properties. For example, the most extreme male walker
had a shoulder-to-hip ratio of 2.4, whereas the most extreme female walker had a shoulder-to-hip ratio of 1.6. The shoulder sway also decreased from extreme male to extreme female, with the lateral distance traveled over a gait cycle by the dot representing the clavicle decreasing from 0.35 to 0.11° of visual angle. The orientation of the walkers in the horizontal plane was also varied, with leftward or rightward walkers at one of three view angles: 3°, 30°, and 90°, with 90° corresponding to a sagittal (side) view and 3° corresponding to a near-frontal view.

To combat ceiling effects by increasing the difficulty, we introduced a mask of linearly moving dots traveling at 1.7° per second in random directions. The dots had a lifetime of 1 second after which they reappeared at a different location. There were 150 dots on the screen at any time in a 13° by 13° square centered on the point-light walker.

The walker and surrounding noise could be centered at one of four locations all offset from the initial fixation mark at the center of the screen by 4.8° of visual angle, shifted either up and left, up and right, down and left, or down and right. The purpose of this was to force viewers to make at least one saccade to reach the walker, reducing a possible bias towards fixating the center of the figure.

3.2.4 Procedure

The study was run in two blocks. The same set of point-light displays was used in both blocks; only the instructions were different. Following calibration of the eyetracker, participants were instructed to either identify the gender of the walker (male or female), or the direction the walker was walking (left or right) in the following trials. After those trials and a short break, they were given the other set of instructions and continued. The order of the blocks was counterbalanced across participants.
The six gender conditions, six view angle conditions, and four possible offsets from the center combined to make a total of 144 possible stimuli. Each of them occurred twice, resulting in a total of 288 trials for each block, with the order randomized within the block.

Trials began with a central fixation point that also served the purpose of drift correction for the eyetracker. Once the participant had fixated this point, that is, their eyes were relatively stationary, they pressed the spacebar and the fixation point disappeared to be replaced by the walker. The walker was displayed for 2 seconds, with the eyetracker collecting eye position data, after which the participant was prompted to press one of the arrow keys on the computer’s keyboard to indicate their judgment of either the gender or the direction of the walker, depending on which of the two blocks they were running.

**Data analysis** We analyzed the eyetracking position data to determine fixation locations. To detect saccades, we first smoothed the instantaneous eye velocity information with a moving average window 40 ms in width. Then we determined a saccade velocity threshold for each trial separately by computing the median and standard deviation of eye movement velocity after removing data points over 50 °/s, and labeling all samples with velocities of more than 3 standard deviations above the median as part of a saccade. All samples between the saccade periods were labeled ‘fixations’, even though by examination of eye velocity plots there was some movement during fixation periods, indicating possible microsaccades, fixation drift, or pursuit eye movements. Therefore the eye gaze events that we distinguished might be more accurately called saccades and non-saccades. However we will refer to the latter as fixations for the rest of the paper. The location of a fixation was defined as the average eye position during the fixation period.
For visualization purposes, we constructed fixation maps (Wooding, 2002) for different sets of conditions. We began with a matrix of all zeros corresponding to individual pixels of the display, and then progressively added circular Gaussian blobs for each fixation, centered on the fixation location (independent of fixation duration). These Gaussians had a standard deviation of 27 pixels or 1° of visual angle, roughly the size of the foveola, and equal heights of 1 unit. After all fixations for the relevant conditions were added to the map, the values at each pixel were scaled relative to the highest peak in the map, meaning that the darkest part of the image was the most fixated and corresponded to a value of 1. To compare fixations between the two tasks, we combined maps, creating dual-color fixation maps, by using the saturation of red at a given point to indicate fixations in the direction task, and the saturation of blue to indicate fixations in the gender task. Locations that were fixated frequently in both tasks appeared dark purple or black.

To test our hypotheses about fixation regions quantitatively, we defined three regions of interest (ROI) relative to the center of the walker (Figure 3.1). The vertical coordinates of the

![Figure 3.1. The predefined shoulder, pelvis and foot regions of interest, with the dots of an example point-light walker in frontal and side view respectively superimposed for reference. The rectangles indicating the regions and the lines connecting the dots are for illustration purposes and were not part of the stimuli.](image-url)
ROIs were selected to fully contain the pairs of dots that represented either the shoulders, pelvis or feet across all view angles, gender levels, and frames of the animation, with an additional margin of approximately 0.5° of visual angle above and below to capture fixations that were near the dots.

We conducted three-way repeated-measures ANOVAs to investigate the role of task and difficulty levels on gaze patterns, in the three predefined regions separately. The factors were Task (Direction or Gender), View Angle (walker rotated 3°, 30° or 90° from front-facing), and Gender Level (0.5, 1.5 and 2.5, with 2.5 being the most distinctively male or female). We used as a dependent measure the number of fixations that fell into each of the three ROIs, making a total of three $2 \times 3 \times 3$ ANOVAs. All repeated-measures $F$ values were computed using an assumption of sphericity.

### 3.3 Results

#### 3.3.1 Accuracy

A repeated-measures analysis of variance applied to the proportion correct showed that we successfully manipulated the difficulty of the direction and gender tasks (Figure 3.2) by varying the viewing angle and the degree of maleness or femaleness, respectively. The factors were Task (direction or gender), View Angle (3°, 30° or 90° offset from a frontal view) and Gender Level (0.5, 1.5, or 2.5, where larger values are less ambiguous). There was a significant main effect of Task, $F(1, 19) = 170.42, p < .001$, with the gender task ($M = 0.64$) being more difficult than the direction task ($M = 0.86$). There were significant interactions between Task and View Angle, $F(2, 38) = 174.84, p < .001$, and between Task and Gender Level, $F(2, 38) = 44.76, p < .001$. 


Within the direction task results, a follow-up 3 × 3 repeated-measures ANOVA showed an effect of View Angle on accuracy, $F(2, 38) = 381.36, p < .001$, but no effect of Gender Level, $F(2, 38) = 0.43, p = .65$. The effect of View Angle was due to accuracy differences between each pair of view angles, as shown by follow-up pairwise comparisons, $p < .003$, with the direction of walkers at near-frontal view angles being most difficult to judge.

Within the gender task results, a follow-up 3 × 3 repeated-measures ANOVA showed an effect of Gender Level on accuracy, $F(2, 18) = 21.01, p < .001$, and also View Angle, $F(2, 18) = 63.59, p < .001$, with no interaction between View Angle and Gender Level, $F(4, 16) = 2.16, p = .12$. Follow-up pairwise tests showed significant differences between gender accuracy at gender levels 0.5 and 1.5 and between levels 0.5 and 2.5, $p < .001$, as well as between 1.5 and 2.5, $p = .011$, with performance improving as the gender of the walkers became more distinct. Though performance at the 0.5 gender level, the most ambiguous, was close to chance level (which was 50%), it was still significantly above chance, $t(19) = 2.87, p = .01$. Follow-up tests showed that the effect of View Angle was due to significant differences between gender

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**Figure 3.2.** The average proportion correct for the direction and gender task at A) the different walker view angles, where 90° represents a side view and 3° represents a near-frontal view, and at B) the different gender levels, where 0.5 is most ambiguously male or female and 2.5 is most exaggerated. The error bars represent ±1 SEM.
Figure 3.3. A) Fixation map across all participants and all conditions. The red areas indicate where there were the most fixations in the direction task and the blue where there were the most fixations in the gender task, while dark purple or black areas were frequently fixated in both tasks. B) Corresponding vertical distribution of fixations, showing the average height of the fixation map at each vertical position for the two tasks. Normalized so that the highest peak across both tasks is 1.

accuracy at the 3° and 90° view angles, and between 30° and 90°, $p < .001$, but not between 3° and 30°, $p = .21$. Performance at the 90° view angle was much worse than in the more frontal conditions, but significantly better than chance, $t(19) = 2.76$, $p = .01$. The gender task performance results fell within the range reported by studies included in a meta-analysis conducted by Pollick, Kay, Heim, and Stringer (2005): these studies found accuracies for point-light walker gender tasks between 58% and 76% ($M = 71\%$) for walkers presented from frontal or oblique views, and between chance level and 72% ($M = 66\%$) for walkers presented in a sagittal view.

3.3.2 Fixation maps

The dual-color fixation map averaged across conditions and participants (Figure 3.3 A) showed that in both tasks, participants fixated most often in the area of the point-light walker
stretching between and including the pelvis and shoulders. The corresponding plot of the vertical
distribution of fixations (Figure 3.3 B) confirms this. An overall tendency of participants to fixate
more on the feet in the direction task can be seen in the light red areas of the fixation map. Only
in the direction task were there any number of fixations that fell below the pelvis.

Qualifying these overall results, individual dual-color fixation maps (Figure 3.4) showed
a large amount of variation among the 20 participants in where they fixated, and in how their
fixation patterns were between the two tasks. Most participants’ fixations fell horizontally at the
midline of the body on average. Vertically they landed frequently at the pelvis, shoulders, or both.
Some participants did not show much difference in fixation patterns between the two tasks, which
can be seen from their primarily black and dark purple maps, while others changed their fixations
greatly, which can be seen in clearly separated red and blue regions. Of those who did change
their fixation strategy to an appreciable degree, the majority placed a greater emphasis on the feet
in the direction task than in the gender task (and so have red areas near the feet). Some
participants (e.g. #2, #11, #15) also showed a horizontal shift of their typical fixation locations
between the two tasks, but these shifts did not show a consistent pattern across participants.

In the next section these observations are quantified using the average number of
fixations that fell within three pre-defined regions of interest.
Figure 3.4. Fixation maps for each participant, showing the fixations for the two tasks across all view angles and all gender levels. The red areas indicate where there were the most fixations in the direction task, and the blue where there were the most fixations in the gender task. Dark purple or black areas were frequently fixated in both tasks.
3.3.3 Regions of interest

We analyzed the average number of fixations in each ROI. In the foot region, a three-way repeated measures ANOVA on the average number of fixations, with Task, View Angle, and Gender Level as factors, showed a significant main effect of Task, $F(1, 19) = 10.64, p = .004$, with participants fixating more often on the feet in the direction task (Figure 3.5). There was also a main effect of View Angle, $F(2, 38) = 3.91, p = .03$ and an interaction between Task and View Angle, $F(2, 38) = 3.49, p = .04$. Follow-up ANOVAs within the direction task and gender task revealed that the main effect of View Angle was carried exclusively by mean differences in the direction task, $F(2, 38) = 4.13, p = .02$, with no effect of View Angle in the gender task, $F(2, 38) = 0.32, p = .73$. Pairwise tests in the direction task showed that there were more fixations on the feet in the 3° than in the 90° (side view) view angle (Figure 3.6), $p = .04$, but no significant difference between the number of feet fixations in the 3° and 30° view angles or the 30° and 90° view angles, $p = .10$ and $p = .07$ respectively. A test for a linear trend was significant, $p = .04$. 

*Figure 3.5. The average number of fixations within the three ROIs for the two tasks. The error bars represent ±1 SEM. * indicates significance at the .05 level, ** significance at .01.*
The average number of fixations during the direction task in the three ROIs for the different view angles, where 90° is a side view and 3° is a near frontal view. The error bars represent ±1 SEM. * indicates significance at the .05 level.

There was no significant main effect of Gender Level in the foot region, as well as no other interactions.

Participants fixated more often in the shoulder region during the gender task than during the direction task, $F(1, 19) = 4.89, \ p = .04$ (Figure 3.5). There were no effects of View Angle or Gender Level in the shoulder region, and no interactions.

In the pelvis region, there was a significant main effect of View Angle on number of fixations, $F(2, 38) = 4.51, \ p = .02$, and no other main effects or interactions. Pairwise tests indicated more fixations on the pelvis in the 3° (near frontal) view angle than in the 90° view angle, $p = .02$. Although differences between the other pairs of view angles in number of fixations did not reach significance, a test for a linear trend was significant, $p = .02$, indicating a pattern of fewer pelvis fixations at the less frontal view angles.

Figure 3.6. The average number of fixations during the direction task in the three ROIs for the different view angles, where 90° is a side view and 3° is a near frontal view.
3.3.4 Individual gaze patterns and accuracy

To determine whether the regions typically favored by each individual in the direction and gender task could predict their performance, we conducted a multiple regression for each task. The predictors were an individual’s average number of fixations in the three ROIs (shoulder, pelvis and foot) during the task, and the dependent variable was their proportion correct. In the direction task, the regression was not significant, $R^2 = .03$, adjusted $R^2 = -.15$, $F(3, 16) = .18$, $p = .90$. In the gender task, the regression was also not significant, $R^2 = .21$, adjusted $R^2 = .06$, $F(3, 16) = 1.40$, $p = .28$. Average number of fixations in the three ROIs could not predict an individual’s performance.

3.4 Discussion

Our results reveal that observers look at the feet of a point-light walker when performing a direction task, but rarely when performing a gender task. Furthermore, they look at the shoulders in a gender task relatively more often than in a direction task. These findings provide evidence for task-specific cues in those locations.

The premise that the primary cues to gender are found in the shoulder and pelvis of a walker is supported by our results. Participants looked often at the hips and shoulders in the gender task and rarely at the feet. The importance of the shoulders may be that they most clearly reveal the degree of lateral body sway (Mather & Murdoch, 1994). Kozlowski and Cutting’s (1977) experiment on gender judgments when sets of dots were removed from a point-light walker found that upper body (head, arms and shoulders) information was more helpful than lower body (pelvis and legs). The fact that our participants on average did not look at the shoulders more often than the pelvis may have been because people are not aware that the shoulders carry gender cues: in Barclay et al.’s (1978) study, when asked which body movements
they thought best differentiated men and women, all participants mentioned the hips while only 76% mentioned the shoulders. Other candidate cues involve the shoulders and pelvis equally, such as the shoulder–hip ratio. In contrast to our finding of approximately equal number of fixations on the shoulders and pelvis, Johnson and Tassinary (2005), also studying gaze patterns in a gender classification task, observed almost twice as many fixations on the pelvis and lower torso as on the chest and shoulders. This could be accounted for by the fact that they used solid models for their walking displays, which, unlike point-light walkers, depicted the size of the waist and other non-skeletal anatomical measures. Exaggerating this effect, half the trials in their eye movement analysis only varied the shape of the waist, with the kinematics remaining constant, so that the only gender information was in the waist–hips area. The present study extends Johnson and Tassinary (2005) by showing the role of different tasks, and by defining gender levels in terms of a linear space of motion-captured male and female walking patterns rather than by varying isolated motion cues such as hip and shoulder sway. We also observed fixations with the walkers displayed at different view angles.

On average, our participants fixated more often in the region containing the feet when they performed the direction task than when they performed the gender task. Along with the frequent fixations on the pelvis and upper body in this task, this points to a contribution of both local and configural cues in judging direction. Although the present study cannot directly address the relative importance of these two sources of information, the localized change in attentional emphasis is predicted by previous findings (Chang & Troje, 2009; Mather et al., 1992; Troje & Westhoff, 2006) that the local motion of the extremities, in particular the feet, are useful in perception of direction. A hypothesized aspect of the foot motion that triggers biological motion processing is the characteristic acceleration profile: Chang and Troje (2009) found that inversion reduced accuracy in direction judgments from two foot dots, but not when the acceleration profile
of the foot movement was distorted. The bottom-most ROI always contained the foot dot, but for certain walkers the knee dots passed into it during part of the gait cycle. This means that our findings of increased fixations in this ROI are also consistent with an alternative model, that local subconfigurations of limbs (Pinto & Shiffrar, 1999) are what are critical for some biological motion tasks. In either case, the task-related shift in attention to the lower body indicates that participants likely made use of directional information located there.

Our prediction that increased task difficulty would enhance the difference in gaze patterns between the two tasks was partially verified. In the direction task the most difficult condition, in which the walker was displayed at a near-frontal view angle, resulted in the most foot fixations on average, whereas the sagittal view angle resulted in significantly fewer foot fixations. Changes to both the gender level and to the view angle affected performance in the gender task, replicating Mather and Murdoch’s (1994) and Troje’s (2002) finding that the gender of a point-light walker is easiest to determine in frontal views; however neither gender level nor view angle affected the number of fixations in the ROIs. This discrepancy between the effect of task difficulty in the two tasks may be the result of view angle being more immediately accessible (as reflected in the higher accuracies), allowing the participant to adjust their gaze strategy dynamically based on the difficulty, whereas in the gender task the 2 second display time may not have been sufficient to both determine the difficulty of the task and then carry out an appropriate strategy. Alternatively, 2 seconds may have been more than sufficient to determine the walking direction in the easier direction task conditions, after which the viewer could explore the display freely, including areas not necessary for performing the task. This interpretation is supported by the finding of ceiling and near-ceiling direction performance at the 90° and 30° view angles.

Contrary to our expectations, performance in the two tasks could not be predicted by how an individual distributed their fixations among the shoulders, pelvis and feet. This may have been
due to lack of variability in individual behavior. In the direction task there was relatively low between-subjects variability in the proportion correct ($SD = 0.04$) and high variability in the distribution of fixations, whereas in the gender task there was somewhat higher variability in the proportion correct ($SD = 0.08$) and low variability in the regions fixated (almost exclusively the shoulders and pelvis). This could be improved in future studies by making the direction task more difficult, and by using a larger walker, so that observers would have to decisively fixate on the pelvis or shoulders to gather information from them. Nevertheless, the fact that participants fixated more on the feet in the direction task than in the gender task, and more on the upper body in the gender task than in the direction task, shows that they sought task-relevant information in these two areas.

Future eyetracking studies using human motion stimuli should examine the temporal patterning of fixations. The time dimension is critical for describing gaze behavior when viewing non-periodic actions; in particular, displays of social interaction. It is already known that there are predictable temporal and spatial patterns of fixations for normal observers on video of social scenes, but which are different for people with autism (Klin, Jones, Schultz, Volkmar, & Cohen, 2002). This phenomenon could be further investigated with point-light displays, which offer a large number of possibilities for controlling aspects of social interaction animations: information can be removed and altered, and even the degree of interactivity can be modulated.
References


Chapter 4

Assessing biological motion perception components³

4.1 Introduction

Tools for assessing basic visual capabilities have been available for centuries, and are widespread and effective. Visual acuity, contrast sensitivity, and color perception are all measured in clinical settings on a regular basis. However, as social animals one of the main uses of our visual system is to observe the people around us, and less attention has been paid to measuring this aspect of vision. Social perception encompasses both static and dynamic stimuli: we obtain valuable information not only from a person’s posture and facial expression, but also from their movement. Johansson’s (1973) point-light display isolates important aspects of social recognition from motion. In this display, a person performing an action, such as walking, is represented by only a few dots tracking the major joints and head. Nevertheless, the observer experiences an instantaneous, rich percept of a human in action, including a strong sense of the movement style. In contrast to more realistic stimuli, point-light displays allow for well controlled manipulations and a rigorous, quantitative psychophysical approach. Therefore they have become the basis for the study of the complex social ability known as biological motion perception.

Biological motion perception is characterized by a number of interesting observations: like faces, point-light displays are subject to an inversion effect (Pavlova & Sokolov, 2000); they can be organized in as little as 100 ms (Oram & Perrett, 1996); and they are extremely salient and

³ This chapter is reprinted from “A battery of tests for assessing biological motion perception” by D. R. Saunders and N. F. Troje, manuscript submitted to Seeing and Perceiving, Feb. 28, 2011.
therefore difficult to mask, even with dynamic noise (Cutting, Moore, & Morrison, 1988). Furthermore, humans are sensitive to biological motion within two days of birth (Simion, Regolin, & Bulf, 2008); both the dorsal and ventral visual pathways play a part in it (Vaina, Solomon, Chowdhury, Sinha, & Belliveau, 2001); and perceiving it appears to be possible even in the absence of simple motion detection ability (McLeod, Dittrich, Driver, Perrett, & Zihl, 1996). These intriguing properties and the questions they raise have made it important to develop adequate methods for assessing biological motion perception.

Early demonstrations of biological motion perception simply showed point-light displays to naïve observers and asked them to describe what they saw, but more sensitive measures were soon developed. For example, a popular task involves determining the heading direction of a point-light walker shown in profile view, often superimposed with masking noise consisting of moving dots with the same size and luminance (e.g. Cutting et al., 1988). However, what is measured by this task is ambiguous. Past research has suggested that there are at least three complex perceptual abilities that could contribute to good performance. First, distinguishing the walker from the noise, a type of figure-ground segregation, appears to be a distinct ability that can be selectively impaired. For example, Ikeda, Blake and Watanabe (2005) found that detection of biological motion is particularly poor in the periphery, beyond what could be explained by decreasing visual acuity. It was later discovered, however, that this phenomenon depended on the walker being masked by noise dots. In a paradigm that did not require masking noise, Gurnsey, Roddy, Ouhnana, and Troje (2008) were able to compensate for the worse performance at greater eccentricities by increasing the size of the stimulus (see also Thompson, Hansen, Hess, & Troje, 2007). Only detecting a walker in masking noise was ‘unscalably poor’. Similarly, in a study comparing older and younger adults in a direction discrimination task, performance was equal
between the two groups except when masking noise was added, in which case the older adults were significantly worse than the younger adults (Pilz, Bennett, & Sekuler, 2010).

A second ability that can be used to perform the walking direction task is retrieval of structure from motion (Cutting, 1981; Hoffman & Flinchbaugh, 1982; Johansson, 1976; Webb & Aggarwal, 1982). Structure-from-motion processing, in the context of a point-light display, infers the articulated structure of the walker using the local relations of the moving dots representing the joints. If the motion trajectory of the dot representing the wrist is more simply described in the reference frame of another dot, such as the elbow, then they are interpreted as being rigidly connected. In this way, the observer recovers the organizing ‘skeleton’, and the form that is revealed can then give the direction of walking. With structure-from-motion, the motion relations play the same role as the explicit connections in an animated stick figure. This process is not absolutely essential for the perception of biological motion, as was shown by studies in which the pairwise relations between points were disrupted or absent (Beintema, Georg, & Lappe, 2006; Beintema & Lappe, 2002; Casile & Giese, 2005; Cutting, 1981); however when these stimuli are compared to the standard point-light display there is a decrease in performance, indicating that human observers do make use of information about the hierarchical, articulated structure when it is available.

Finally, the same direction task can be accomplished using an independent source of information: the local motion trajectories of the dots, in particular the feet. The first suggestion that this is the case came from Mather, Radford and West (1992), who observed that removing the dots representing the hands and feet particularly harmed ability to determine walking direction. Stronger evidence came with Troje and Westhoff (2006) showing that when the structure of the point-light walker was destroyed by spatially displacing the dot trajectories to random locations (that is, ‘scrambling’ the walker) observers could still determine the walking
direction at above chance levels. Furthermore, when the trajectory representing the feet was
inverted, participants could no longer perform the task, even if the remaining trajectories were
shown in upright orientation. They concluded that the local motion trajectories of the feet contain
invariants that represent a distinct source of information about the walker. They speculated that
these may be used as an early warning signal for the presence of a terrestrial animal nearby,
which could be predator, prey, or conspecific. Subsequent studies have found that local motion
alone can create an impression of animacy (Chang & Troje, 2008), and that it attracts the interest
of both human infants (Simion et al., 2008) and baby chicks (Vallortigara, Regolin, & Marconato,
2005).

A direction task with masking noise therefore requires at least three distinct abilities, and
all three are confounded in measurements of performance on this task. Biological motion
perception is not a single phenomenon (Troje, 2008), as is implicitly assumed when only a single
task is used to assess it. It is important to conceptually distinguish the different major abilities
involved in biological motion perception, and to devise ways to isolate them experimentally.
Tasks differ in the demands they place on underlying processes, and as a result will exhibit
different properties. For example, local biological motion processing appears to not be affected
by repeated exposure (Chang & Troje, 2009b), whereas more global types of processing are
(Grossman, Blake, & Kim, 2004). In addition to tasks calling on figure-ground segregation,
structure-from-motion, and the processing of local motion invariants, higher-order biological
motion abilities are also often assessed, in particular action recognition (Dittrich, 1993; Olofsson,
Nyberg, & Nilsson, 1997) and perception of high-level stylistic properties relating to identity
(Cutting & Kozlowski, 1977; Loula, Prasad, Harber, & Shiffrar, 2005; Westhoff & Troje, 2007),
emotional state (Dittrich, Troscianko, Lea, & Morgan, 1996; Pollick, Paterson, Bruderlin, &
Sanford, 2001), or biological properties such as sex, weight, or age (Troje, 2002b). Like the
lower-level abilities, they are typically measured using point-light displays, in which case good performance also reflects structure-from-motion ability. The stimuli are also sometimes presented in a mask of noise dots, requiring figure-ground segregation. Therefore higher-level biological motion abilities are rarely measured independently of lower-level abilities.

Lack of attention to what a task measures has led to apparent inconsistencies in the biological motion perception literature. For example, there have been varying answers to the question of whether people with Autism Spectrum Disorders are impaired in biological motion perception. Studies in which people judged walking direction of point-light displays have generally found no differences between ASD participants and controls (Herrington et al., 2007; Murphy, Brady, Fitzgerald, & Troje, 2009; Saygin, Cook, & Blakemore, 2010), although one study using adolescents did find impairment (Koldewyn, Whitney, & Rivera, 2010). On the other hand, a set of studies in which participants labeled the emotion being portrayed by a point-light actor (Atkinson, 2009; Hubert et al., 2007; Moore, Hobson, & Lee, 1997; Parron et al., 2008) consistently found lower accuracy in the ASD group. These same studies also found no evidence of a difference in ability to recognize actions. Taken together, these results suggest a pattern of relative normalcy in determining structure from motion, but difficulty in interpreting high-level style properties. However, apart from the combination of action categorization and style recognition, groups of participants with ASD have rarely been assessed on more than one of these biological motion tasks at a time. As well, no published study has assessed people with ASD on their ability to recognize individuals by their walk, which might be expected to be impaired from their difficulties with face recognition (Klin et al., 1999). All of these tasks might be said to assess biological motion perception, and all likely have parallels in the day-to-day social world. Therefore the question of whether people with ASD have deficits in perceiving biological motion
will have multiple answers, that should derive from conducting multiple tests on the same sample of individuals.

Another set of questions that can be clarified by a multiple-ability framework comes from neuropsychology. Researchers have measured the performance of patients with brain lesions in biological motion tasks in an attempt to distinguish and localize the underlying cognitive mechanisms. However nearly all studies use no more than two biological motion tasks, and few choose the same set. Therefore it can be difficult to interpret the implications of patients who appear to show dissociations between different abilities. Even patients with deficits in low level motion detection may be normal in action categorization (McLeod et al., 1996; Vaina, 1994; Vaina, Cowey, LeMay, Bienfang, & Kikinis, 2002), person identification (Vaina, 1994; Vaina et al., 2002), sensitivity to scrambling (McLeod et al., 1996), and detection in noise (Billino, Braun, Bohm, Bremmer, & Gegenfurtner, 2009). Conversely, low-level vision can be preserved while there are deficits in action categorization and walker identification (Cowey & Vaina, 2000), or detection in noise (Billino et al., 2009). Two studies also showed dissociations between 2D or 3D structure-from-motion tasks and biological motion tasks (Battelli, Cavanagh, & Thornton, 2003; Vaina & Gross, 2004). Most interestingly for the multiple-ability viewpoint, there exist patients who are impaired in some biological motion tasks and not others: two patients of Schenk and Zihl (1997), FM1 and FM2, were poor at detecting the walker in noise, but were able to judge direction and scrambling within the normal range. LM as measured by McLeod et al. (1996) had normal scores on a scrambling detection task, but not a direction determination task (and in particular was strongly affected by masking noise), while JL (Battelli et al., 2003) had the inverse pattern, with normal performance on a direction task but a deficit in identifying the scrambled walker. Although these are only single cases, they suggest that these tasks call on different underlying processes. For these latter patients in particular, who would be considered impaired in
biological motion perception by some measures but not by others, a spectrum of tasks would more clearly illuminate the nature of their deficits.

Not only the task but also details of the stimuli can strongly affect biological motion assessment. A common method of generating point-light displays may have eliminated the role of local motion invariants in many experiments. Cutting’s (1978) algorithm produces a synthetic point-light walker animation that broadly resembles a person walking, but differs from real human motion in the shape and velocity profiles of its local motion trajectories, with this particularly evident in the feet. Observers are not able to determine direction of walking from the local motion of this stimulus as they are from displays derived from motion capture recordings (Saunders, Suchan, & Troje, 2009). Cutting’s walker was used, and continues to be used, in many studies, including some experiments in which the direction of a scrambled walker is relevant (Bosbach, Prinz, & Kerzel, 2004; Fujimoto, 2003; Lange & Lappe, 2007; Mather et al., 1992; Pilz et al., 2010), and where, therefore, the lack of local directional cues almost certainly affected the measured performance.

Assessment of the abilities of special populations, as well as many other investigations of biological motion perception, would benefit from a standardized package of biological motion tasks that is efficient to administer, and that uses motion stimuli derived from recordings of human movement. This paper introduces such a battery of tests (Figure 4.1), BMLtest (for Bio Motion Lab). Three of the tests assess figure-ground segregation, structure-from-motion, and local motion invariant extraction, as independently as possible. For example, only the test of figure-ground segregation (the Detection test) uses a dynamic dot mask. Additional tests measure higher-level biological motion abilities: action categorization and style recognition, the latter of which is divided into gender classification and person identification. In these tests the human figures are portrayed as animated stick figures rather than point-light displays, so that
measurements will be independent of structure-from-motion ability, and even participants who cannot organize point-light displays can be assessed with them. Finally, the Coherence test measures simple motion detection ability. The current paper describes the tests that make up BMLtest, and evaluates it using undergraduate samples.

Besides providing normative data and an estimate of the reliability of the tests, full-battery data from undergraduate participants can test the hypothesis that these tasks measure relatively independent constructs. Using the logic of individual differences (e.g. Kosslyn, Brunn, Cave, & Wallach, 1984; Wilmer, 2008), if there is little correlation between performance on the tests then it is evidence that they are assessing mostly or entirely independent mechanisms. Conversely, if individual performance on two tasks is highly correlated, then they share some underlying mechanisms. This type of data can serve as a complement to other evidence of independence, such as neuropsychological data, regarding such questions as whether low-level motion perception relies on distinct pathways from biological motion perception.

Since BMLtest is intended to be used in a variety of settings, Experiment 1 tested the robustness of performance when the tests are administered via the internet, in a home setting. We assessed participants twice in counterbalanced order, once at home and once in a lab setting, and compared the test scores. We also gave participants a questionnaire to determine features of the home testing environment, a topic which is of increasing interest to psychologists weighing the advantages and disadvantages of internet-based studies (e.g. Birnbaum, 2004). Given the evidence from the first experiment in support of the effectiveness of online testing, Experiment 2 relied on home-based internet testing to collect normative data and test-retest reliability estimates for the final version of BMLtest.
<table>
<thead>
<tr>
<th>Test Name</th>
<th>Assesses</th>
<th>Stimulus</th>
<th>Task</th>
<th>Test score</th>
</tr>
</thead>
<tbody>
<tr>
<td>1 Coherence test</td>
<td>Low-level motion processing</td>
<td>Dots that change their position from frame to frame either randomly, or coherently in one direction.</td>
<td>Indicate whether average motion goes to the left or to the right.</td>
<td>Percent coherence for 82% correct performance.</td>
</tr>
<tr>
<td>2 Detection test</td>
<td>Figure-ground segregation for biological motion.</td>
<td>Point-light walker in profile view in a scrambled walker mask or an empty scrambled walker mask.</td>
<td>Indicate whether the first or second interval contained the walker.</td>
<td>Number of noise dots for 82% correct performance.</td>
</tr>
<tr>
<td>3 Direction test</td>
<td>Processing of local biological motion invariants.</td>
<td>Horizontally scrambled point-light walker at different viewing angles.</td>
<td>Indicate whether the walking direction is left or right.</td>
<td>Viewing angle in degrees for 75% correct direction discrimination (smaller values indicate greater discrimination ability).</td>
</tr>
<tr>
<td>4 Distortion test</td>
<td>Structure-from-motion ability for biological motion.</td>
<td>Partially scrambled or intact point-light walker.</td>
<td>Indicate whether the first or second interval contained the intact walker.</td>
<td>Percent scrambling for 82% correct scrambling discrimination.</td>
</tr>
<tr>
<td>5 Action test</td>
<td>Action recognition.</td>
<td>Stick figure animations depicting 10 different actions.</td>
<td>Choose the name of the action from a menu.</td>
<td>Percent correctly labeled.</td>
</tr>
<tr>
<td>6 Gender test</td>
<td>Sensitivity to the expressed gender of a walker.</td>
<td>Stick figure animations depicting male or female walkers.</td>
<td>Indicate the gender of the walker.</td>
<td>Maximum slope of the psychometric curve for gender discrimination (larger values correspond to greater sensitivity).</td>
</tr>
<tr>
<td>7 Identity test</td>
<td>Person identification from walking style.</td>
<td>Stick figure animations of unique individuals, including 3 walkers to memorize (target walkers) and 45 others (distractor walkers).</td>
<td>Two blocks of tasks: Naming. Choose the name of the depicted walker from a menu of three names. Old-New. Indicate whether the walker is one of the three originally studied, or a new walker.</td>
<td>Naming. Percent correctly labeled. Old-New. Sensitivity in terms of $d'$ (larger values correspond to greater sensitivity).</td>
</tr>
</tbody>
</table>

*Figure 4.1. Descriptions of the tests that comprise BMLtest.*
4.2 BMLtest: A Comprehensive Assessment of Biological Motion Perception

4.2.1 Overview

BMLtest is a web application, implemented in Adobe Flash, that assesses multiple abilities which play a role in biological motion perception (Figure 4.1). After entering their name and other information, participants are presented with a list of the seven tests that make up the test battery. Once they have completed a test, that option is disabled. Researchers can instruct participants to take all of the tests in a particular order, or to take a subset of the tests. The stimuli in each test appear in the same fixed order across all participants.

Each test begins with a screen containing instruction text and a button to perform a practice trial. The practice trials are easier versions of the task, designed to familiarize the participant with the task requirements without providing much training. The participant may practice until they feel ready, and then click the “Begin” button to start the test. Once a test is completed, a summary of the results and the performance on individual trials can be accessed online by the researcher.

In the following descriptions of the individual tasks, stimulus sizes are reported in degrees of visual angle, under the assumption of a viewing distance of 60 cm, a monitor that measured 18 inches diagonally, and a screen resolution of 1280 × 1024 pixels. All the stimuli consist of moving white dots or lines shown on a black background. In the case of the stick figures, there are dots located at the major joints in addition to the lines. Dots are 6 pixels in diameter, corresponding to 0.16° under the assumed distance and screen size. Lines are 3 pixels (0.08°) thick. All point-light walker (and stick figures) are 4° of visual angle in height. Walkers remain in one place as if on a treadmill.
4.2.2 Adaptive threshold procedure

Five of the tests use a QUEST threshold procedure (Watson & Pelli, 1983) to measure a stimulus intensity threshold, which becomes the test score. The advantage to using a maximum-likelihood procedure like QUEST is that a high precision can be achieved with a relatively small number of trials, and it does not require prespecifying a range of thresholds in the population, as the method of constant stimuli does (Watson & Fitzhugh, 1990). Both are advantages for using the tests with special populations, such as children, where there might be difficulties in maintaining focus for long periods of time and where there might be little a priori knowledge of the distribution of thresholds.

For all tests using the QUEST procedure, a Weibull-shaped psychometric function is assumed with parameters set based on the suggestions of Watson and Pelli (that is, $\beta = 3.5$ and $\delta = 0.01$). The procedure terminates after a fixed number of trials, as recommended by Anderson (2003), specifically 40. The initial test intensities, which also determined the mean of the prior probability distribution, were chosen separately for each test to be approximately one standard deviation easier than the predicted mean threshold in the population based on pilot data.

4.2.3 Coherency test

4.2.3.1 Introduction

The Coherency tests assesses low-level motion processing, based on the task used by Newsome and Paré (1988).

4.2.3.2 Stimuli

The display consists of 15 dots which are initially placed at random within a 10 degree-diameter circular region. At a frame rate of 20 Hz, on each subsequent frame each dot will either be moved to a randomly chosen new location within the circular region, or displaced by 0.15° to
the left or the right. The percent coherence represents the probability that a dot will be selected to move rather than to be randomly relocated.

4.2.3.3 Procedure

In each trial, the dynamic random dot display is shown for 1 s, with the percent coherence set by a QUEST procedure as described above. Then the display disappears and participants click on the button labeled “Left” or “Right” to indicate the direction of motion. A blank screen is displayed for 1 s before the next trial. The test terminates after 40 trials.

4.2.4 Detection test

4.2.4.1 Introduction

The Detection test assesses the ability to extract biological motion from noise (e.g. Bertenthal & Pinto, 1994; Cutting et al., 1988), a form of figure-ground segregation.

4.2.4.2 Stimuli

The stimuli consist of point-light displays created from the averaged Fourier-transformed motion recordings of 50 men and 50 women walking (Troje, 2002a). The motion of a human shown in profile view and facing either to the left or to the right is represented by 11 dots. The walker is placed within a mask of scrambled walker noise. This mask consists of a specified number of dots with the identical movement trajectories of the dots of the walker, with their average location placed randomly within a square 5 × 5 degree region. The masking dots are the same size and luminance as the walker dots. When the walker is absent, the number of masking dots is increased by 11 to match the number of dots in the walker-present display. The walker, if present, is offset from the center of the mask region by a random value between -0.5 and 0.5° horizontally, and -1.5 and 1.5 vertically, so that it is not always in the same location but it also
remains completely within the masked region. The entire display was offset from the center of the screen by 4° to the left in the first interval, and 4° to the right in the second interval.

4.2.4.3 Procedure

Two intervals are presented in each trial, with a 0.5 s delay between them. One interval displays a region of masking noise containing the walker, while the other displays a region of masking noise alone. The number of masking noise dots is set by a QUEST procedure. The order of the intervals is randomized. After both intervals are displayed, participants click on a “First” or “Second” button to indicate which interval contained the walker. A blank screen is displayed for 1 s before the next trial. The test terminates after 40 trials.

4.2.5 Direction test

4.2.5.1 Introduction

The Direction test assesses sensitivity to local biological motion invariants (e.g. Chang & Troje, 2008, 2009a; Saunders et al., 2009; Troje & Westhoff, 2006), by measuring a participant’s ability to determine walking direction from a spatially scrambled walker.

4.2.5.2 Stimuli

The stimuli consist of scrambled point-light displays created from averaged motion capture recordings of human walking (Troje, 2002b). 15 dots represent the movement of the major joints, as well as the chest and center of the pelvis. In each trial, the horizontal location of each dot making up the walker is placed randomly within a 3 degree range, approximating the width of a walker, while the vertical position is unchanged, for example keeping the dots representing the feet near the bottom of the display. Compared to scrambling vertically and horizontally, this provides a perceptually stronger local motion direction cue, while still
completely disrupting structural cues to the direction (Hirai, Chang, Saunders, & Troje, 2011). The relative phase of the dots was also randomized.

4.2.5.3 Procedure

In each trial, the horizontally-scrambled walker display is presented for 1 s from a certain horizontal viewing angle. Whether it is walking left or right is randomized, but the magnitude of the angle offset from facing the viewer is set by a QUEST procedure. The walker disappears and participants click on a Left or Right button to indicate the walking direction. A blank screen is displayed for 1 s before the next trial. The test terminates after 40 trials.

4.2.6 Distortion test

4.2.6.1 Introduction

The Distortion test assesses structure-from-motion ability, that is, the ability to reconstruct the hierarchical body structure from moving dots representing the joints of a walking figure (Cutting, 1981; Hoffman & Flinchbaugh, 1982; Johansson, 1976; Webb & Aggarwal, 1982). If participants can correctly discriminate the partially scrambled walker, then they are sensitive to disruption of the pairwise joint relationships that define the hierarchical structure.

4.2.6.2 Stimuli

The point-light displays for this test consist of 13 dots representing the major joints of a human walking. The walker is presented from a horizontal viewpoint that is between 10° and 15° offset from a frontal view. The possible values for percent scrambling range between 0.01% and 100%. In the 100% case, the dots of the walker are placed randomly within an area 4° of visual angle high and 2° wide on each trial (so roughly occupying the area of an intact walker).
Scrambling values close to 0.01% produce only slight distortions in the locations of the walker dots.

Walkers are placed randomly within a 3 × 3 degree region and presented from an elevated vertical viewpoint such that the walker is tilted forward by 20°. These steps prevent simple strategies for detecting distortion, such as comparing the second walker to the afterimage of the first, or observing whether the hip dots are horizontally aligned. The entire display was offset from the center of the screen by 4° to the left in the first interval, and 4° to the right in the second interval.

4.2.6.3 Procedure

There are two intervals presented in each trial, with a 0.5 s delay between them. One interval displays an intact walker and the other displays a partially scrambled walker, each for 1 s. The percent scrambling is set by a QUEST procedure. The order of the intervals is randomized. After both intervals are displayed, participants click on a “First was normal” or “Second was normal” button to indicate which interval they thought contained the intact walker. A blank screen is displayed for 1 s before the next trial. The test terminates after 40 trials.

4.2.7 Action test

4.2.7.1 Introduction

The Action test assesses action categorization (Dittrich, 1993; Johansson, 1973). The nature of action categorization from point-light displays has been less carefully studied than other abilities, but it has been frequently used as a task in special populations, such as brain lesion patients (e.g. Heberlein, Adolphs, Tranel, & Damasio, 2004; McLeod et al., 1996; Vaina et al., 2002). The Action test does not use an adaptive threshold procedure.
4.2.7.2 Stimuli

The stimuli consist of animations of a stick figure performing an action. The source for the animations were motion capture recordings of men and women. Three unambiguous depictions of each of the 10 actions were chosen, and the most representative 1 s segment from each one was used.

4.2.7.3 Procedure

Each action is presented for 1 s. The point-light display disappears, and the participant chooses the name of the action from a menu of buttons, consisting of “Catching”, “Climbing stairs”, “Jumping”, “Jumping jacks”, “Kicking”, “Lifting”, “Running”, “Sitting”, “Throwing”, and “Walking”. A blank screen is displayed for 1 s before the next trial. Each action is presented at three different viewpoints – 0° (facing the viewer), 30° and 90° (facing to the right) – for a total of 30 trials.

4.2.8 Gender test

4.2.8.1 Introduction

The Gender test assesses sensitivity to the gender of a walker (e.g. Kozlowski & Cutting, 1977; Mather & Murdoch, 1994; Troje, 2002a), which is an example of biological motion style recognition.

4.2.8.2 Stimuli

The stimuli consist of stick figures of human walking in place as viewed from an angle randomly selected to be slightly offset from facing the viewer. The animations have a specified gender level for each trial, which is achieved by using discriminant function analysis to create an axis in the vector space defined by the motion-captured walking sequences of 50 men and 50
women that best captures the differences between male and female walks (Troje, 2002b). The walkers also have an individual walking style, that is produced by adding a normally-distributed random vector in the walker space with the same standard deviation as the real walking sequences. Each participant sees the same set of individuals, but with different gender levels applied to them according to the adaptive procedure.

4.2.8.3 Procedure

Each walker is presented for 1.5 s. When the walker disappears, the participant chooses between two buttons labeled “Male” or “Female.” On each trial the synthesized gender is either male or female, with the degree of expression of that gender selected by one of two interleaved QUEST procedures. Thus the test produces two thresholds at the end, one for detecting maleness and one for detecting femaleness. There are a total of 80 trials, 40 trials for each staircase, with the interleaving of male and female trials randomized once but fixed across participants. A blank screen is displayed for 1 s before the next trial.

Once the two thresholds have been estimated, they represent two points on the psychometric curve for discriminating male from female walkers. These points are fitted with a logistic function. The gender test score is then the maximum slope of the curve, representing the sensitivity. The gender response bias can also be determined from this curve, as the point of intersection with the 50% correct level.

4.2.9 Identity test

4.2.9.1 Introduction

The Identity test assesses the ability to recognize people by their walking style (e.g. Cutting & Kozlowski, 1977; Loula et al., 2005; Troje, Westhoff, & Lavrov, 2005). Memory for
three walkers is evaluated with a Naming and an Old-New subtest (in memory research terms, a
classification task and a recognition task respectively). Besides measuring person identification,
the Naming subtest assists in encoding the target walkers, since feedback is provided. Although
conceptually linked, the test scores from the two subtests were analyzed separately. The Identity
test did not use an adaptive threshold procedure.

4.2.9.2 Stimuli

The stimuli consist of stick figures depicting individual walking styles in the center of the
screen. To create the walkers, first the three target walkers that the participant must memorize
were generated. The goal was to obtain three random target walkers that shared approximately the
same absolute distinctiveness and the same pairwise similarity. Beginning with the motion
capture recordings of 50 men walking, the movement trajectories for each joint along each
dimension were reduced into a 15 dimensional vector space using principle component analysis
(Troje, 2002a). The similarity between walkers in this space can be approximated by the angle
between the vectors representing them, while the Euclidean distance from the origin, which
represents the average walk, roughly corresponds to the visually distinctiveness of the walk. We
randomly generated three vectors in the walker space at a fixed distance from the origin (and so a
fixed distinctiveness), and at a 120° angle from one another, maximizing and equating the
pairwise distance.

To create the distractor walkers, first an initial set of 150 distractors was produced by
randomly generating normally distributed vectors in the walker space. Therefore these walkers
had the same expected mean distinctiveness as the original population, and the same expected
distance from each of the target walkers. Then we used a pilot experiment to choose the
distractors that would produce a certain range of trial difficulties. The pilot experiment
participants were asked to memorize the target walkers, and then had alternating blocks of Naming and Old-New trials as in the regular test. We then ranked the 150 walkers according to the number of correct responses they achieved, then discarded the 15 walkers that had the best performance and picked the 45 easiest walkers that remained to become the distractors.

A different procedure was used to create the stimuli for the first version of the Identity test, used in Experiment 1, which was less effective at controlling the distinctiveness of the walkers and the differences among them.

4.2.9.3 Procedure

There are five phases to the test: a memorization phase, a first Naming phase, a first Old-New phase, a second Naming phase, and a second Old-New phase. Each participant sees the same target and distractor walkers in the same order.

In the memorization phase, each of the three target walkers is presented with its name printed below it, for 5 s at both a 0° and a 45° degree viewing angle (corresponding to a frontal and half-profile view respectively) for a total of 6 presentations. There is a 1 s blank interval between walkers.

In the Naming phases, on each trial one of the three target walkers is presented at a 0° or 45° viewing angle, with buttons corresponding to the three names underneath. The animation continues until the participant chooses one of the names, at which point the participant receives feedback and there is a 1 s blank interval. Each of the two Naming phases had 18 trials.

In the Old-New phases, on each trial either one of the target walkers or a distractor walker is shown at a 0° or 45° viewing angle. Half of the trials are old and half are new, and each of the target walkers is shown an equal number of times, whereas none of the distractors are repeated. The animation continues until the participant clicks on the button marked “Old” (for a
target) or “New” (for a distractor) underneath. There is a 1 s blank interval before the next trial. Each of the two Old-New phases had 45 trials.

4.3 Evaluation of BMLtest – Experiment 1

4.3.1 Introduction

In order to test the robustness of BMLtest in different settings, we had participants complete a web-based version of the test at home and in the lab, and for each test examined the correlation between individual performance in the two settings and differences in mean score.

4.3.2 Methods

4.3.2.1 Participants

There were 30 participants (13 male, 17 female), ranging in age from 18 to 42 (M = 22.8). They were primarily Queen’s University undergraduate or post-graduate students, but several non-students were also included. They reported normal or corrected-to-normal vision. They were compensated with a modest honorarium.

4.3.2.2 Procedure

Participants were randomly assigned to two groups. One group (‘Home first’) conducted BMLtest in the home setting first and then later in the lab setting, while the other (‘Lab first’) encountered the settings in the reverse order. There was an interval of at least 48 hours between the first and second testing session. Each participant ran the complete battery in each setting, with the tests in the order in which they are described above. They were instructed to try to reduce distractions in their home environment, and to complete the tests in one sitting. Upon finishing the test battery in the home setting, they answered a questionnaire that asked about their computer hardware, and also included the questions, “Were there any distractions while you were
running?”, “At any point during a test did you take a break?” and “Did you notice any phenomena while running it that seemed to be a problem with the program?”

4.3.2.3 Data analysis

For all analyses we removed only test scores for which QUEST failed to converge to a threshold, which was defined as an absence of reversals in the last 15 trials and 5 or more trials spent at the maximum stimulus intensity. Across all tests and participants 2.8% of test scores were removed.

4.3.2.4 Apparatus & Stimuli

The lab setting was free of interruptions, and had a consistent display that was a CRT with a display resolution of 1280 × 1024. In both the lab and the home setting participants did not use a chinrest, but were asked to maintain a distance from the screen of approximately 60 cm. Because of differing display devices in the home setting, the questionnaire responses were needed to estimate the range of sizes of the display elements on the retina.

4.3.3 Results

4.3.3.1 Questionnaire responses

According to the post-test questionnaire, the diagonal measurement of the participants’ computer screens in the home setting ranged from 13 inches to over 20 inches, and the screen resolutions from 1024 × 768 pixels to 1920 × 1200 pixels. Based on these ranges, and assuming some inconsistency in the distance to the screen, the visual angle of elements could have varied by as much as a factor of 2. However for a fixed viewing distance, the difference in size due to computer monitors resolutions would have been relatively small: 90% of the reported resolution
and screen size combinations produced between 34 and 41 pixels per degree of visual angle, an increase of only 24% from the smallest to the largest.

Many participants took the tests on a laptop, judging by the fact that 13 of the 30 participants had a monitor that was less than 15 inches diagonally. Of the 15 who replied to a question about the clock speed of their computer, the average was 1.98 GHz, and the range was 800 MHz to 2.83 GHz. For all participants and all tests the average framerate was always over 24 frames per second, and examination of framerates for individual trials strongly suggested that instances of the framerate dropping were rare. In response to the item on the questionnaire about phenomena in the display, only two reported seeing brief stuttering in the animation. Therefore it appears that the computers used by the participants at home were adequate for running BMLtest.

Distractions were a frequent occurrence in the home setting. Of the 30 participants, 18 reported at least one source of distraction during the tests. The most commonly mentioned (6 out of 18) were calls or text messages on their cellphone. Another distraction mentioned more than once was someone in the room talking to the participant. Other distractions included “music playing”, “Alert on computer”, “my cat jumping onto the computer desk”, and “eating at the same time”. Based on responses to the distraction question, at least three participants completed the battery at work, and at least one in a public library. No participant reported taking a break during a test, which was corroborated by the rarity of outlier test durations. However 14 of the 30 reported taking a break between tests.

4.3.3.2 Practice and setting effects

The time between the first and second sessions ranged from 2 days to 33 days, with the median being 5 days. For each test we conducted a mixed-model ANOVA on the mean scores, with the group (‘Home first’ or ‘Lab first’) as a between-subjects factor and session (‘First’ or
Figure 4.2. Mean test scores in the lab and home settings. Units for each of the test scores are explained in Figure 4.1. The error bars represent ±1 SEM. * indicates a significant mean difference, $p < .05$.

Second’) as a within-subjects factor. As expected from the random group assignment, none of the ANOVAs showed a significant main effect of group. However an interaction between the group and session factors would indicate a difference between performance in the lab and home settings.

The Coherent, Detection, Direction, Distortion, and Action tests did not have a significant interaction between group and session, $p > .18$, corresponding to no evidence for a difference between lab and home performance (Figure 4.2). The Coherence test had significant improvement between the first and the second times the participant was tested, $F(1, 28) = 4.89, p < .05$, as did the Detection test, $F(1, 28) = 4.90, p < .05$, and the Action test, $F(1, 28) = 7.39, p < .05$. The ANOVA for the Identity – Old-New test did show signs of an interaction, and therefore an effect of setting, $F(1, 28) = 0.78, p = .06$, but they did not reach statistical significance. There was strong evidence for an improvement in performance between the first and second sessions of the Old-New test ($M = 0.86$ vs. $M = 1.38$), $F(1, 28) = 18.80, p < .001$. 

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On the other hand, the ANOVAs for the Gender and Identity – Naming tests did have significant interactions between group and session, $F(1, 25) = 5.47, p = .028$ and $F(1, 28) = 9.05, p = .005$ respectively, indicating an influence of testing setting on performance. There was no main effect of session for the Gender test, $F(1, 25) = 0.14, p = .71$, so we conducted a paired $t$ test for Home vs. Lab, averaging between participants for whom the lab session was their first time and for whom it was their second time. Participants running the test in the lab were significantly better (had steeper gender discrimination slopes) than participants running the test in their homes ($M = 0.33$ vs. $M = 0.27$), $t(26) = 2.36, p < .05$. Since the Identity – Naming test had a significant main effect of session, indicating a practice effect, we conducted separate paired $t$ tests for Home vs. Lab for the Lab First and Home First participants. There was no difference in mean performance between the home and lab setting for the Lab First participants, $t(14) = 0.52, p = .61$, but performance at home was significantly worse than in the lab setting for the Home First participants, $t(14) = 3.50, p = .003$. Therefore there was a benefit for the Naming test both for running it in the lab vs. home ($M = 83\%$ vs. $76\%$) and for running it a second time vs. the first time ($M = 82\%$ vs. $77\%$).

In summary, experience appeared to improve performance in the Coherent, Detection, Action, and both Identity tests, and a lab setting led to better performance in the Gender and Identity – Naming tests.

A Pearson correlation coefficient between individual performances at home and in the lab found significant correlations for the Coherent, Detection, Distortion, Action, and both Identity tests (Table 4.1). Direction and Gender were not significantly correlated. In the case of the Gender test, both the individual male and female thresholds that are used to compute the slope of the Gender psychometric curve had highly significant test-retest correlations, at $r(28) = .56, p = .001$ and $r(28) = .48, p = .007$, as did the gender bias, at $r(28) = .64, p < .001$. 

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### Table 4.1

<table>
<thead>
<tr>
<th>Test</th>
<th>r</th>
</tr>
</thead>
<tbody>
<tr>
<td>Coherent</td>
<td>.61**</td>
</tr>
<tr>
<td>Detection</td>
<td>.39*</td>
</tr>
<tr>
<td>Direction</td>
<td>.26</td>
</tr>
<tr>
<td>Distortion</td>
<td>.63**</td>
</tr>
<tr>
<td>Action</td>
<td>.61**</td>
</tr>
<tr>
<td>Gender</td>
<td>.05</td>
</tr>
<tr>
<td>Identity – Naming</td>
<td>.60**</td>
</tr>
<tr>
<td>Identity – Old-New</td>
<td>.47*</td>
</tr>
</tbody>
</table>

* indicates $p < .05$,  
** indicates $p < .001$

#### 4.3.4 Discussion

The majority of the tests in the battery had similar mean scores in a lab or home setting, and a significant correlation between scores at home and in the lab, suggesting that the test is robust to different settings with differing visual and attentional parameters. The two tests that did show somewhat poorer performance at home were the Gender test and the Identity test, which may have been more vulnerable to distractions since they required access to high-level representations of movement styles.

Gender test scores and Direction test scores did not correlate significantly between the home and the lab. Some aspects of Gender perception were measured adequately, as can be seen from the male and female test-retest reliability, but estimating the slope of a psychometric curve generally requires a higher level of precision (Turpin, Jankovic, & McKendrick, 2010), so it is possible there was not enough variance in this particular population to reliability detect the test-retest consistency with this sample size. The lack of correlation in the Direction test may have
been the result of a floor effect, since 10 out of the 60 thresholds measured for the viewing angle were the maximum stimulus intensity of 90°, indicating that even for this strongest stimulus (representing a profile-view scrambled walker) many participants may not have been able to achieve the target 82% threshold.

4.4 Evaluation of BMLtest – Experiment 2

4.4.1 Introduction

The purpose of Experiment 2 was to estimate the mean and standard deviation of the scores for all the tests in a particular population (normal undergraduate students), and to estimate the test-retest reliability of the final version of BMLtest. Experiment 2 also provided data for an individual differences analysis to examine whether the tests measured different or overlapping underlying processes. Between Experiment 1 and 2, the method of generating the target and distractor walkers for the Identity test was changed, and the target threshold for the Direction test was lowered from 82% to 75%, so that a greater proportion of test subjects could achieve that performance level within the range of possible stimulus intensities.

4.4.2 Methods

There were 65 new participants (38 female, 27 male), ranging in age from 18 to 43 ($M = 19.2$), recruited from first year undergraduate Psychology and compensated with either course credit or an honorarium. They received instructions by email and completed the tests in the battery in the order in which they are described above using their web browser. They were instructed to run BMLtest again one week later. Only participants whose second session was within one day of this criterion were included in the analysis.
Test scores were selected for elimination if there was evidence the QUEST procedure had not converged after 40 trials, as in Experiment 1. One participant’s Direction test score was eliminated because of a clear loss of attention part way through the task (82% correct in first half, 30% correct in the second half, even though the trials in the second half were easier on average). Including a further participant who did not complete the Identity test, 1.5% of the test scores were absent.

4.4.3 Results

The median time to complete the battery was 32 minutes, and 90% of the sessions took less than 40 minutes including time reading the instructions and practicing.

Test-retest reliabilities for all the tests of the battery were significant at \( p < .001 \) (Table 4.2). We examined the correlations between the scores of individuals on different tests (Table 4.3), using only their first sessions because of the effect of practice. A sequential Bonferonni

<table>
<thead>
<tr>
<th>Test</th>
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<tbody>
<tr>
<td>Coherent</td>
<td>.46**</td>
</tr>
<tr>
<td>Detection</td>
<td>.42**</td>
</tr>
<tr>
<td>Direction</td>
<td>.58**</td>
</tr>
<tr>
<td>Distortion</td>
<td>.50**</td>
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<tr>
<td>Action</td>
<td>.48**</td>
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<tr>
<td>Gender</td>
<td>.55**</td>
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<tr>
<td>Identity – Naming</td>
<td>.46**</td>
</tr>
<tr>
<td>Identity – Old-New</td>
<td>.50**</td>
</tr>
</tbody>
</table>

** indicates \( p < .001 \)
Table 4.3

**Experiment 2 Intercorrelations Between Biological Motion Tests**

<table>
<thead>
<tr>
<th>Test</th>
<th>1</th>
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<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
</tr>
</thead>
<tbody>
<tr>
<td>1. Coherent</td>
<td></td>
<td>.07</td>
<td>.05</td>
<td>.06</td>
<td>.13</td>
<td>-.04</td>
<td>.13</td>
<td>.24</td>
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<tr>
<td>2. Detection</td>
<td></td>
<td></td>
<td>-.28</td>
<td>.30</td>
<td>.00</td>
<td>.33</td>
<td>.32</td>
<td></td>
</tr>
<tr>
<td>3. Direction</td>
<td></td>
<td>-.11</td>
<td>-.07</td>
<td>-.10</td>
<td>.17</td>
<td>-.17</td>
<td></td>
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</tr>
<tr>
<td>4. Distortion</td>
<td></td>
<td>-.26</td>
<td></td>
<td>.24</td>
<td>-.20</td>
<td>-.25</td>
<td></td>
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</tr>
<tr>
<td>5. Action</td>
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<td>-.07</td>
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<td>.23</td>
<td>.24</td>
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<tr>
<td>6. Gender</td>
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<td></td>
<td>.15</td>
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<tr>
<td>7. Identity – Naming</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>-.51*</td>
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<td>8. Identity – Old-New</td>
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<td></td>
</tr>
</tbody>
</table>

* indicates significance at .05 after correcting using sequential Bonferroni
** indicates $p < .001$

correction was applied to the tests of significance. Only the correlation between the Naming and the Old-New subtests of the Identity test was significant. Although all the other correlations were nonsignificant after correction for multiple comparisons, the Coherent and Direction tests in particular had little evidence of an association with the other tests, average $r^2 = 0.014$ and 0.015 respectively.

As an additional test of reliability, we compared test scores for the first session of Experiment 2 with the test scores from the first session of Experiment 1, which used a different set of participants (Figure 4.3). For all the scores, a $t$ test not assuming equal variance did not show a significant difference from the previous experiment, with the exception of the Identity – Naming subtest, $t(72.8) = -2.53, p = .014$.

There was strong evidence for practice effects in Experiment 2, with a significant mean improvement between the first and second sessions in the Detection, Direction, Naming and Old-New tests, paired $t$ tests, $p < .005$. 

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4.4.4 Discussion

The test-retest correlations for all the tests were highly significant, and the similarity of the means with those from Experiment 1 also provided support for the reliability of the tests. The Identity – Naming subtest had higher mean test scores in Experiment 2, which was expected since a different set of target and distractor walkers were used in this experiment than in Experiment 1, and these were designed to make the Identity test easier as well as more consistent in difficulty. Average performance in the Identity – Old-New test was also higher in Experiment 2, although this difference did not reach significance. The thresholds measured in the Direction test were expected to be lower (representing a lower stimulus intensity) in Experiment 2 compared to Experiment 1, since the target threshold was lowered for Experiment 2. The mean difference was in the expected direction (\(M = 27.4\) vs. \(M = 13.7\)), but it was not significance, perhaps due to variance added by the floor effect that was present in Experiment 1. Intertest correlations were not significant, except for the two subtests of the Identity test, which were both designed to measure person identification ability.
4.5 General Discussion

We have presented two experiments which together suggest that BMLtest is effective in both a lab and a home setting. The participants were normal university students, a homogenous population that nevertheless provided an estimate of the distribution of test scores in younger adults, and a basis to test for relationships between the tasks using individual differences. Some aspects of BMLtest might need to be adjusted for special populations such as children and people with autism. However the use of an adaptive threshold procedure for many of the tests means that a wide range of ability levels can be measured.

Estimated test-retest reliabilities fell below the recommended level for psychological tests of 0.7 (Urbina, 2004), but they were approximately within the range measured for tests of memory (Bird, Papadopoulou, Ricciardelli, Rossor, & Cipolotti, 2003) and several visual perception tests, including the DTVP-A, MVPT-3 and Test of Visual Perceptual Skills (Brown, Mullins, & Stagnitti, 2008). The test-retest reliability was also likely underestimated because of the presence of practice effects for several of the tests (McIntire & Miller, 2007). Nevertheless, based on the measurement error that the test-retest reliability implies, a single administration may not be sufficient for assessing individual patients, but rather scores from more than one session should be averaged together. On the other hand, for many studies examining group differences a single session will suffice. The fact that normal participants typically complete all tests in a little over 30 minutes, with individual tests taking 2–5 minutes, makes the individual tests and BMLtest as a whole good candidates to be used in testing sessions alongside other assessment tools, without requiring expertise in biological motion perception or psychophysics.

The fact that performance on individual tests generally did not predict performance on other tests is evidence that the tests measure relatively independent abilities. Although considerations of power mean that all the nonsignificant correlations cannot be confidently
interpreted as dissociations, the particularly low correlation between the test of simple motion
detection (the Coherence test) and the biological motion tests supports the idea that biological
motion perception does not rely on simple motion integration, which was first suggested by the
existence of pairs of brain lesion patients who represent a double dissociation between the two
abilities (Cowey & Vaina, 2000; Vaina et al., 2002). The low correlation of the Direction test
scores with the other test scores demonstrates that local motion was successfully rendered
irrelevant in the lower-level biological motion tasks, for instance by the use of a scrambled
walker mask in the Detection task which ensured that local motion was not a cue to the presence
of the walker, and that when cues other than local motion invariants are present for the higher-
level, style-based tasks, they dominate performance, even though there is evidence that local
point-light motion information is sufficient to discriminate emotional state (Chouchourelou,
Matsuka, Harber, & Shiffrar, 2006).

From the results of Experiment 1, we conclude that BMLtest may be successfully
administered via a web browser, despite differences in the display devices and distractions in the
environment. Therefore the test battery should be effective in a variety of settings where it is
accessed remotely. Beyond that, BMLtest could be used to evaluate biological motion perception
within a large and varied sample of individuals recruited over the internet. In an online survey of
face recognition ability using the Cambridge Face Memory Test, Germaine, Duchaine, and
Nakayama (2011) obtained data from 44,000 participants with a wide range of ages and
background, providing an excellent estimate of ability over the lifespan and revealing that face
recognition peaks at the ages of 30–34. Similar information about biological motion perception
from a demographically rich sample is lacking.
Since biological motion perception is more than one phenomenon, a multiple-task battery is needed to adequately assess it, and to begin to disentangle the underlying processes that make this ability possible. BMLtest represents such a tool.
References


Chapter 5
General Discussion

5.1 Biological motion perception is more than one phenomenon

The results presented in Chapters 2 and 3 support the assertion by Troje and Westhoff (2006) that biological motion perception is not a single phenomenon. Different performance characteristics were seen when different tasks were used: in the case of Chapter 2, different sensitivity to the form of local motion trajectories, and in Chapter 3, different patterns of gaze fixation. The task differences were the result of various underlying processing components playing a greater or lesser role. Chapter 4 explicitly described a model of the components of biological motion perception, including processing of local motion invariants, structure-from-motion, figure-ground segregation, action categorization, and style recognition. It also provided a tool for assessing the components separately, and so avoiding the confounding that is inherent in many tasks that are in use.

The central inspiration for this framework was the evidence from Troje and Westhoff (2006) that there is a separate process of biological motion perception that does not require configural information, and that has its own distinct inversion effect. There has been subsequent empirical support for this idea, in addition to the experiments reported here. Local motion trajectories that are consistent with a living creature are of particular interest to visually inexperienced newborn chicks (Vallortigara & Regolin, 2006; Vallortigara, Regolin, & Marconato, 2005) and 2-day-old infants (Simion, Regolin, & Bulf, 2008); they give an impression of animacy (Chang & Troje, 2008); they can portray emotional state (Chouchourelou, Matsuka, Harber, & Shiffrar, 2006); and they can be processed preattentively, facilitating attentional orientation to them (Hirai, Saunders, & Troje, 2010; Jiang, Wang, Zhang, & He, 2011).
Figure 5.1. Direction discrimination accuracies from an experiment manipulating the global structure and local motion orientation of a point-light walker. (from Hirai, Chang, Saunders, & Troje, 2011)

2010). Most recently, it has been discovered that there is an interaction between the global form and local motion processing, such that if the global configuration of the walker is upright (that is, the trajectories of the feet are in their expected location at the bottom of the display), the implied direction is much more easily perceived (Figure 5.1; Hirai, Chang, Saunders, & Troje, 2011). In fact, direction accuracies in point-light walkers that are scrambled but where the feet retain their vertical position are nearly as high as for intact walkers, which is consistent with the relatively low direction discrimination thresholds for horizontal scrambled walkers described in Chapter 4.

These findings reinforce the notion that there are at least two dissociable perceptual processes operating on biological motion stimuli. There is only limited evidence specifically dissociating the other components, apart from the dissociations based on individual differences reported in Chapter 4. The finding that masking noise has a greater impact on older adults (Pilz, Bennett, & Sekuler, 2010) and in the periphery (Gurnsey, Roddy, Ouhnana, & Troje, 2008;
Thompson, Hansen, Hess, & Troje, 2007) suggests that figure-ground segregation is a distinct stage, while Chapter 2 demonstrated that structure-from-motion is not sensitive to the dynamics of local features in the same way that local motion processing is. Further studies should investigate the degree of dissociation between these components.

5.2 Computational models of biological motion perception

To achieve a detailed understanding of how biological motion perception is accomplished, this basic outline of a hierarchy of five processing stages should be filled in with computational models of the components and how they interact. As an example of a model of a single component, Hoffman and Flinchbaugh (1982) and Webb and Aggarwal (1982) presented algorithms for computing structure-from-motion. In terms of the interaction of components, there are only two major computational models of biological motion perception in the psychological literature: Lange and Lappe’s (2006) and Giese and Poggio’s (2003). However neither provides a full account of the components that are discussed here.

Lange and Lappe’s (2006) model (also see Lange, Georg, & Lappe, 2006) is based on matching frames of biological motion to a library of static form templates. By definition, it is incapable of using information from motion signals, and so cannot account for processing of local motion invariants, or any sensitivity to dynamics. The only representation of time is as a learned ordering of templates. The creators acknowledge that unlike humans, the model is not capable of determining direction from scrambled motion, or distinguishing between Cutting’s walker and a motion-capture-based point-light display (Lange & Lappe, 2007), which have different perceptual properties for humans, as described in Chapter 2.

The two tasks Lange and Lappe’s model has been shown to be capable of accomplishing are a direction discrimination task and a forward-backward walking discrimination task. For the
direction task, the library consists of two sets of templates, one for walking to the left and one for walking to the right, with each template representing a pose that occurs during the walking action. The first stage activates templates based on the current frame, eventually selecting one as the best match, while the second stage of the model integrates the templates over time to make an overall judgment of ‘left’ or ‘right’. This task might be considered a type of action categorization, but to recognize the full range of actions that humans are familiar with, and the range of styles in which each action can be performed, would require a combinatorial explosion of static form templates. The model is also not capable of making use of structure-from-motion, since motion processing is needed to infer the pairwise relationships between points, and it is not clear how the model could segregate a point-light display from scrambled walker noise, although it has some tolerance to masking by limited-lifetime static noise dots (Lange & Lappe, 2006). Nevertheless, this model does represent a proposal of how biological motion could be processed through multiple stages to recognize its global form.

Giese and Poggio’s (2003) computational model also processes biological motion via static templates, but in addition includes a separate mechanism which integrates local motion signals, and so provides a better fit for the multi-component model. Raw visual input is processed in two parallel streams, analogous to the dorsal and ventral streams of the visual system (Mishkin, Ungerleider, & Macko, 1983). Each stream consists of several stages at which the activation of units in the earlier stage is aggregated and simplified, resulting in receptive fields that increase in scale from highly localized to the entire display. The information from the form and motion pathways is only integrated at the final stage. This feed-forward neural network model can be trained on different actions and walking styles, and it achieved good performance in distinguishing among point-light depictions of walking, running and limping, and among the walking patterns of four individuals. It could also detect a figure within a scrambled walker mask.
Although it does not include a role for structure-from-motion, or a separate figure-ground segregation stage, Giese and Poggio’s model can be regarded as a step towards filling in the framework, with the components roughly mapping to units in the model at the different spatial scales and in the different streams. It demonstrates how responses to local features could be summed to create a global impression, and it incorporates critical features that are based on local motion, similar to the local motion invariants explored in Chapter 2. Unlike Lange and Lappe’s model, it has been demonstrated to be capable of action categorization and person identification, although with small datasets so that the true extent of its abilities are not yet known.

The multi-component framework described in this thesis is not a computational model, but it provides constraints for such models, and a way to clarify future psychophysical and neuroscience research in terms of components that should be studied and assessed separately – for instance by a battery of tests as described in Chapter 4.

5.3 Functional dissociation and neural localization of components

The focus of this thesis has largely been behavioral data, but the components of biological motion perception can be further illuminated by results from neuroscience. Chapter 4 reviewed neuropsychological findings that suggest that biological motion is processed separately from low-level motion detection, and in addition some patients’ deficits represent dissociations between different biological motion abilities (McLeod, Dittrich, Driver, Perrett, & Zihl, 1996; Vaina, 1994; Vaina, Cowey, LeMay, Bienfang, & Kikinis, 2002). The patients whose only confirmed deficits were in biological motion perception all had damage to the right temporal lobe (Vaina & Gross, 2004). There have also been valuable investigations using electrophysiology (Oram & Perrett, 1994, 1996; Perrett et al., 1985) and ERP data (Hirai, Fukushima, & Hiraki, 2003; Hirai & Hiraki, 2006; Jokisch, Daum, Suchan, & Troje, 2005; Wheaton, Pipingas,
Silberstein, & Puce, 2001), as well as repetitive TMS (Grossman, Battelli, & Pascual-Leone, 2005), but the tool best suited for localizing these components in the brain is functional brain imaging. Below, fMRI and PET results are reviewed that suggest brain areas associated with each of the components, with the nonoverlap of implicated areas providing further evidence for dissociations.

5.3.1 Processing local motion invariants

Few imaging studies of biological motion perception have identified regions involved in extracting invariants from local motion, because most take the approach of contrasting brain activation due to viewing normal point-light walkers with activation due to scrambled walkers (Bonda, Petrides, Ostry, & Evans, 1996; Grossman et al., 2000; Peelen, Wiggett, & Downing, 2006; Peuskens, Vanrie, Verfaillie, & Orban, 2005). Both stimuli in this contrast contain valid local motion invariants, so the activity in regions that process local motion invariants will be cancelled out. The posterior superior temporal sulcus (pSTS) was most frequently singled out as selective for biological motion processing.

One study did investigate activation due to local motion, in a 2 × 2 design where either the configuration or kinematics of a point-light display was disrupted, or both (Jastorff & Orban, 2009). They identified the pITS, pMTG, pSTG and pSTS as selective for the kinematics of a point-light display of human action, including the scrambled condition. One other study, Jiang and He (2007), contrasted upright with inverted scrambled point-light walkers, and found relatively greater activation in V3/V3a and pIPS in the upright condition. Further fMRI experiments are needed to interpret these results and clarify the areas that are responsible for processing local motion signals.
5.3.2 Structure-from-motion

Since one major difference between the intact and scrambled display is the lack of valid pairwise rigidity, the studies that contrast the two and find activation in the pSTS raise the question of whether this area might primarily be associated with structure-from-motion processing. Peuskens et al. (2005) compared activation to non-biological articulated dot displays versus activation to human point-light displays. The activation in right STS when the dot motion represented a human walking compared to activation in right STS due to the non-biological articulated display was much lower. Another study using articulated non-biological displays also found no activation in pSTS, but activation in pITS, ITG and FFA/FBA (Pyles, Garcia, Hoffman, & Grossman, 2007). Furthermore, pSTS responds equally well to Beintema & Lappe’s (2002) sequential-position walker, which cannot be interpreted using structure-from-motion (Michels, Lappe, & Vaina, 2005). Finally, it appears that pSTS does not respond exclusively to human body motion, but also to mouth movements (Puce, Allison, Bentin, Gore, & McCarthy, 1998) and the sound of footsteps (Bidet-Caulet, Voisin, Bertrand, & Fonlupt, 2005). It has been suggested that the real selectivity of pSTS is to the presence of intentional action (Saxe, Xiao, Kovacs, Perrett, & Kanwisher, 2004).

Better candidates for structure-from-motion processing may be the extrastriate body area (EBA) and fusiform body area (FBA). The EBA overlaps the previously highlighted pITS region (Pyles et al., 2007), and was found by both Jastorff and Orban (2009) and Jiang and He (2007) to be responsive to intact but not scrambled walkers, as was the FBA. However this interpretation is complicated by the fact that these areas are also selective for still photographs of human bodies (Downing, Jiang, Shuman, & Kanwisher, 2001; Peelen & Downing, 2005).
5.3.3 Figure-ground segregation

Like local motion invariant processing, figure-ground segregation of point-light walkers in noise masks has not been localized directly using functional imaging. Although one of the striking characteristics of biological motion is how difficult it is to mask (Cutting, Moore, & Morrison, 1988), only Grossman et al. (2004) have used a stimulus that included masking noise. Rather than using a whole-brain contrast, they took a region of interest approach, limiting the activity that was analyzed to the functionally located pSTS and FFA regions. Therefore brain regions involved in figure-ground segregation could not be freely identified, but the study did determine that there is less activation in pSTS and the FFA when the walker is masked, and that this discrepancy is reduced as participants gain practice with the masked walker stimulus.

5.3.4 Action categorization

One result that speaks to localization of action categorization is Jastorff and Orban’s (2009) report of activation due to viewing complex, varied actions such as aerobics and boxing as opposed to walking displays. Although their participants were not asked to identify the action, there was a distinct different in the brain regions that were activated, with pSTS only responding in the case of complex actions, but EBA and FBA responding to both complex actions and the basic, repeated walking action. This constitutes further evidence that pSTS may not be critically involved in many commonly used biological motion tasks involving point-light displays of walking.

5.3.5 Style recognition

The ability to determine the style of an action from body movements, as acquired through experience, was examined by Calvo-Merino, Glaser, Grezes, Passingham, and Haggard (2005). They used fMRI to determine the regions that were activated while observing videos of ballet or
capoeira dance performances, with the observers being experts in one of the two dance forms or non-experts. They found greater activation in premotor areas when participants viewed a dance style corresponding to their own expertise, as well as the intraparietal sulcus, the right superior parietal lobe, and left pSTS. Non-experts showed no significant differences in activation between the two stimuli. This may indicate that visual representations of motion style is primarily located in the premotor cortex, as predicted by mirror neuron theory (Rizzolatti & Craighero, 2004; Rizzolatti, Fadiga, Gallese, & Fogassi, 1996). However one study of observation of mouth actions suggested that there are two distinct action categorization systems: one for actions the observer has personally experienced, which relies on mirror neuron areas, and a distinct system for actions that the observer has not performed him- or herself (Buccino et al., 2004).

In summary, there is only limited fMRI evidence to date that isolates regional activity due to the components of biological motion processing proposed here, but activation in pSTS appears to be related to processing of complex human activity, while structure-from-motion may be associated with the EBA and FBA. Recognition of specific styles of motion is associated with the premotor cortex, when the viewer is experienced in the action that is portrayed.

5.4 A multi-component approach to studying biological motion perception

This thesis has outlined and provided evidence for at least five distinct components of biological motion perception. There could very well be more components: for example, as suggested by the test battery results of Chapter 4 in which Gender test results did not correlate with Identity test results, person recognition may use additional abilities beyond basic style recognition. Nevertheless, the present framework represents an advance over the treatment of biological motion perception as a single phenomenon even when assessed by very different tasks.
The components function as stages in a processing hierarchy, with intermediate representations generated by one stage serving as input to the next stage, but with the components capable of useful perceptual judgments on their own. When we observe the actions of others in everyday life we likely make use of several of these components simultaneously. However to truly understand biological motion processing, experimenters must carefully construct their task and stimuli so as not to confound the contributions of several components. Once this becomes widespread practice, many questions that have had conflicting answers in the biological motion literature, from its neural concomitants to whether people with autism are impaired in perceiving it, will be untangled.
References


Summary and Conclusions

Biological motion perception is more than one phenomenon. It is best studied as a hierarchy of processing components, ranging from detection of low-level motion invariants to high-level style recognition. The experiments in this thesis derive from this framework, and from earlier results showing dissociations between the processes used in different biological motion perception tasks. In the first study, it was shown that humans are sensitive to the dynamics portrayed in the local motion trajectories of a point-light walker, such that when the trajectories are altered (in this case by the substitution of Cutting’s walker for motion-capture-derived data) they can no longer be used as a source of information (Chapter 2). In the second study, data was presented showing different eye fixation patterns for different tasks, consistent with the hypotheses that the tasks call on fundamentally different mechanisms, and that the local motion of the feet carries directional information (Chapter 3). Finally, a battery of biological motion tests was introduced, with the purpose of measuring aptitude on these different processing levels independently and so avoiding the confounds that are inherent in many tasks that are in current use (Chapter 4). Individual differences data collected using the battery contribute evidence for these components as distinct constructs. The multi-component framework has clarified a number of issues in perception of biological motion, and will help to guide future research.
Appendix

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