

The action dynamics of overcoming the truth

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A convincing deceiver must act in discordance with their knowledge of the truth. To do so requires the deceiver to resolve competition between what is known to be true and what is intended to be false. We investigated the temporal signature of this competition by examining the action dynamics of arm movement while participants responded falsely or truthfully to autobiographical information. The participants answered *no* or *yes* by navigating a Nintendo Wii Remote to NO and YES regions on a large projector screen. Trajectory analyses of the fine-grained arm movements show increased complexity in false responding relative to truthful responding, with the greatest difference in false *yes* answers. The dynamic motor movements also reveal greater strength of competition during the act of false responding, thereby extending traditional response time measures that capture latent competition alone. These results suggest that deceptive processes may be detectable when action is allowed to covary with thought. Supplemental figures and a list of the sentence stimuli may be downloaded from <http://pbr.psychonomic-journals.org/content/supplemental>.

Most people can easily confirm or deny an assertion that they once met Elvis Presley or that they collect clocks in their spare time. Nonetheless, evidence suggests that prior to a final confirmation or denial of such assertions, people temporarily and perhaps nonconsciously believe the assertions to be true (Gilbert, 1991; Gilbert, Tafarodi, & Malone, 1993). The nature of belief appears to be biased toward initially accepting a proposition as true, even when that proposition is unequivocally false. It stands to reason that for someone to dishonestly confirm or disconfirm any proposition, as in agreeing that they met Elvis when they really did not, the person will accept the assertion as true, then assess whether their initial acceptance is correct (which it is not), and then respond in order to violate the conclusion of their assessment (by falsely responding *yes*). In such deceptive behavior, competition exists between the initial belief, the assessment of the belief, and the intention to deceive. Although the initial belief is involuntary, this acceptance of the truth is an impediment that requires active processing to overcome.

By many accounts of deception, the competition and resolution involved in false responding is a far more challenging and time-consuming process than confirming the truth (Vendemia & Buzan, 2004; Walczyk et al., 2005). Accordingly, researchers interested in deception detection have designed clever tasks that exploit this increased processing time (e.g., Gregg, 2007; Sartori, Agosta, Zogmaister, Ferrara, & Castiello, 2008). Participants often respond honestly or deceptively to simple statements, and the time taken to respond, by vocal onset or a manual keypress, is recorded and analyzed. In general, these response time latencies are useful in discriminating certain deceptive behaviors and have thus risen to prominence as a standard

bearer of detecting deception. Unfortunately, response time captures only the outcome of a completed cognitive process, and the real-time cognitive dynamics that occur during the process are lost.

To begin exploring these moment-to-moment changes of response selection, we turn to a growing body of action dynamics research. In this research, actions that occur in conjunction with a cognitive task often reflect ongoing characteristics of processing, ranging from low-level speech perception (Spivey, Grosjean, & Knoblich, 2005) to higher level learning (Dale, Roche, Snyder, & McCall, 2008). The response activity involved in these tasks is usually recorded as arm movements within a set spatial region. Analysis of the arm movement provides insight into what information is important during processing and when that information is most relevant. For example, in Dale, Kehoe, and Spivey (2007), participants used a computer mouse to make typicality judgments on category membership. Each trial involved matching an animal exemplar (e.g., *whale*) to one of two visually copresent response options located in opposite corners of a computer screen. For some trials, classification was potentially ambiguous, such as matching *whale* to a *fish* or *mammal* option. During response movements, the streaming x,y coordinates of the computer mouse were recorded. Dale et al. (2007) found that computer mouse trajectories for atypical animals (e.g., *whale*) curved more toward a featurally similar distractor (e.g., *fish*), suggesting that semantic categorization processes also unfold partly into the dynamics of response execution.

We use this cognition–action interplay to tap the dynamics of false responding. Indeed, there is well-established evidence that deception often leaks into a deceiver's actions, such as with gesture and body posture (Vrij, 2001).

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Here, we employed an action dynamics technique to study response behavior as continuous competition from an initial belief and the goals of deception. To the extent that this competition is expressed in action, we can use x, y coordinate trajectories to expose the dynamics of overcoming this initial true belief and enacting the agenda of a false response.

In the present study, we expose hidden cognitive activity that is involved when one falsely accepts or denies assertions about oneself. To do so, we use a guided lie paradigm that is commonly employed in EEG and fMRI analysis (e.g., Spence et al., 2004; Vendemia & Buzan, 2004). In our version, participants are prompted to respond falsely or truthfully to simple autobiographical facts, such as *Have you ever been to Asia?* Here, rather than answering with a computer keypress, the participants used a Nintendo Wii Remote, and the x, y coordinates of their arm movements were rapidly sampled (see Dale et al., 2008).

With this rich data output, we evaluated signatures of deception in terms of the shape of each movement trajectory and the location of the trajectory over time. We also quantified trajectory properties on dimensions of velocity, stability, and direction. As the results reveal, the unpacking of response time provides not only unique distinctions between false and true responses, but also the more subtle distinction between false responses answered with either a *no* or a *yes*.

EXPERIMENT Revealing the Dynamics of False and True Responding

Method

Participants. Twenty-six undergraduate students (19 female, 7 male) participated for extra credit. Only native English speakers with normal or corrected-to-normal vision were eligible to participate. All of the participants were right-hand dominant.

Procedure. A trial began with a small bull's-eye-shaped circle appearing at the bottom center region of a 3.8×1.8 m screen positioned approximately 2.7 m directly in front of the participant. The participants' task was to click on the circle with the Wii Remote-controlled cursor, and by their doing so, the first word in a biographical question would appear above the circle. With each click of the Wii Remote, the current word was replaced with the next word in the question. This process continued until the final word of the question was encountered (akin to self-paced reading tasks; Just, Carpenter, & Woolley, 1982). At this point, a NO response box appeared in the top right corner of the screen, and a YES response box appeared in the top left corner of the screen (each box was approximately 0.5×0.5 m). Also at this time, the bull's-eye-shaped circle changed to green or red. If the circle changed to green, the participants were simply instructed to answer the question truthfully. If the circle changed to red, the participants were to answer falsely. All responses were made by navigating the Wii Remote cursor to the appropriate NO or YES box.

During each trial, the x, y coordinates of the cursor movement were continuously recorded (sampling at approximately 80 Hz) and stored for later processing. Because the Wii Remote was held with an extended arm in front of the body and toward the screen, subtle directional changes in wrist and arm movement were captured, with 33 pixels traversed for each centimeter of lateral movement.

The participants responded to 84 question trials, including 4 practice trials. The 80 experimental trials were divided equally between false and true color prompts, and the order of the false and true

prompts and the order of the questions were randomized for each participant. The position of the NO and YES response boxes was also reversed for a third of the participants (i.e., the NO response box appeared in the top left corner of the screen, and the YES response box appeared in the top right corner of the screen). After the initial set of questions had been answered, the participants completed a follow-up verification task that required them to truthfully reanswer all of the questions that they had previously viewed.

Question stimuli. The questions used in the study began with the stem *Have you ever . . .* and were completed with 120 possible statements (i.e., *Have you ever eaten pizza?*; the stimuli are available online¹). Eighty of the 120 questions were randomly selected for each participant. The questions were also selected to elicit an equal number of false *no* responses, true *no* responses, false *yes* responses, and false *no* responses. A pilot study confirmed that the responses were approximately evenly distributed within individuals across our target population. All questions were completed with two to three words, following the pattern of verb + object or verb + preposition + object. The object in the question always occurred in the sentence-final position to prevent early guessing of sentence meaning.

Results

Trials were excluded if the responses in the verification task were incongruent with the original responses or if response time was more than 3 *SDs* from the mean. This exclusion criterion eliminated 1 participant (incongruent responses exceeded 50% of total trials) and 92 trials (5% of the data). Of the 1,908 trials remaining, 435 trials occurred in the false/*no* condition, 455 trials occurred in the true/*no* condition, 503 trials occurred in the false/*yes* condition, and 515 trials occurred in the true/*yes* condition.

Trajectory shape. With this analysis, we examined the shape of each trajectory as it moves from the bottom center bull's-eye to the final response box at the top left or top right of the screen. To conduct this analysis, we initialized the response trajectories for each participant to x, y coordinates (0,0) and interpolated to 101 time steps (see Dale et al., 2007; Spivey et al., 2005). At each time step, the x - and y -coordinate positions were then averaged within conditions for each participant. To compare conditions, we performed paired *t* tests at corresponding x -coordinate time steps (a total of 101 *t* tests). A consecutive run of statistically significant tests indicates that the trajectories between conditions diverged during response execution.

The false *no* and true *no* trajectories diverged for 29 time steps ($p < .05$) between the 59th and 88th steps, whereas the false *yes* and true *yes* trajectories statistically diverged for 58 time steps ($p < .05$) between the 40th and 98th steps (Figure 1A). The divergence for each comparison exceeds the minimum number of 8 consecutive time steps that bootstrap simulations have shown to be a standard for statistical significance (see Dale et al., 2007). Accordingly, the trajectory shape analysis revealed false and true responses that were conspicuously different. The true response movements appeared to travel a more direct route to the target response, whereas the false response movements took a more curved route. The bend of the curve was always in the direction of the competing response option (i.e., the true response). This greater curvature suggests that competition was greater for the false responses, whereas the processing for the true responses was relatively unaffected.

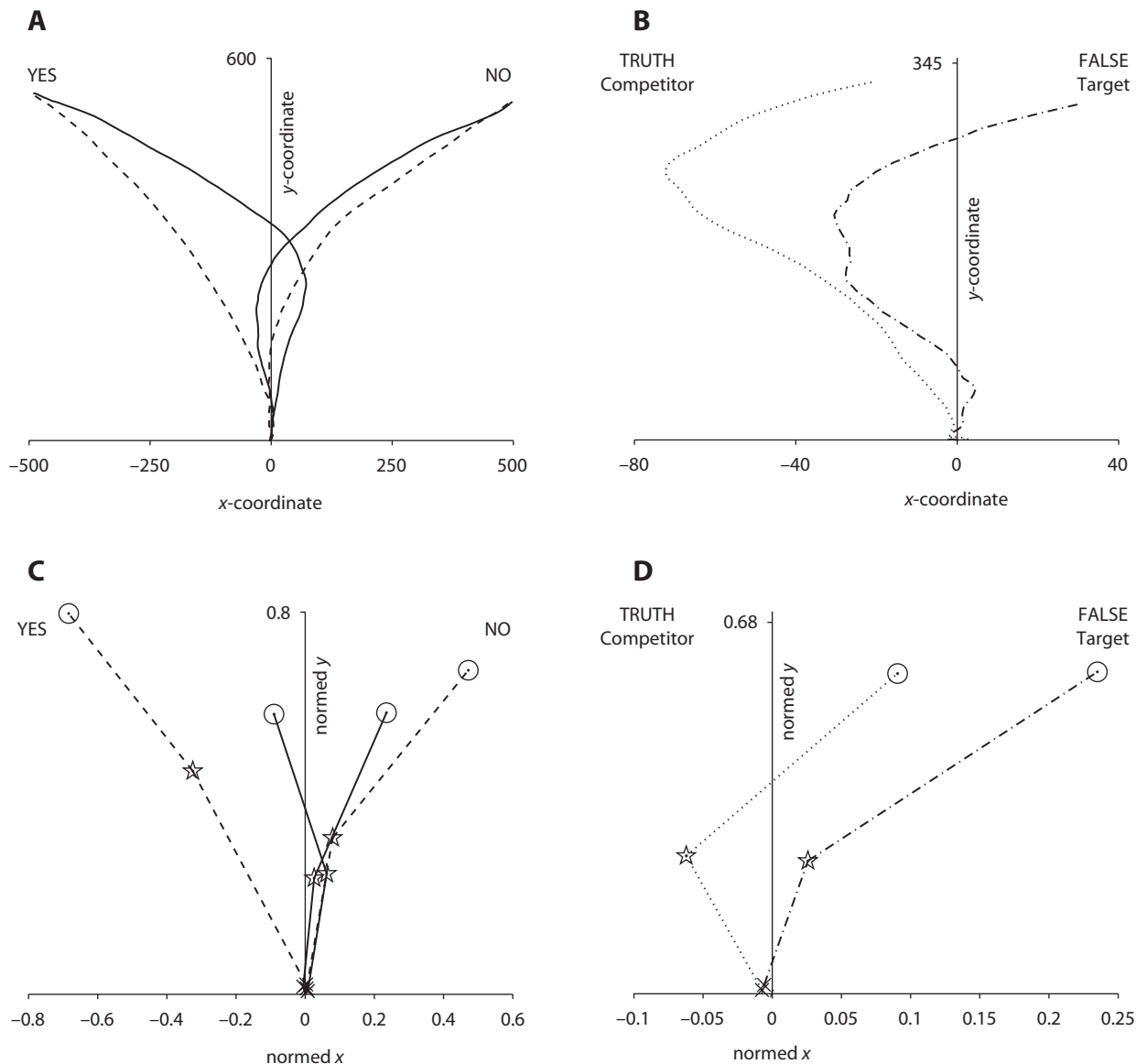


Figure 1. (A) Shape of Wii Remote trajectories for each condition. False answers (solid lines) display a greater arc toward the competing response option than do true answers (dashed lines). (B) Shape of trajectories for false *yes* responses (mirror reversed from trajectory in panel A) compared with the shape of false *no* responses. The false *yes* responses (dotted line) are closer to the competing true option than are the false *no* responses (dash-dotted line). (C) Location of Wii Remote trajectories for each condition. The *x*-coordinate position on each trajectory is plotted at 500 msec (cross), 1,000 msec (star), and 1,500 msec (circle). The false answer positions (connected by solid lines) show slower movements toward the correct response location (e.g., the upper left corner for false *yes*) and are closer to the competing response option (e.g., the upper right corner for false *yes*) than do the true answers (connected by dashed lines). (D) Location of trajectories for false *yes* responses (mirror reversed from trajectory in panel B) compared with the shape of false *no* responses. The false *yes* responses (dotted line) are closer to the competing true option than are the false *no* responses (dash-dotted line).

To further explore this competition, we examined divergence between false *no* responses and false *yes* responses by superimposing mirror-reversed false *yes* trajectories (see Figure 1B). A paired *t* test analysis revealed that the false *yes* responses diverged from the false *no* responses for 16 time steps ($p < .05$) between the 55th and 70th time steps. During these time steps, the false *yes* responses were spatially closer to the competing true response option. Not only was there a greater competition for general false responding, but this effect was most pronounced with false *yes* responses.

Trajectory location. With this analysis, we compared the location of response trajectories after the first, second, and third 500 msec of processing (for 500, 1,000, and 1,500 msec). This information was lost in the previous analysis when temporal information was collapsed into fixed time steps. Now with trajectory location, we can answer the question of when trajectories begin to be statistically divergent. To conduct this analysis, the trajectory coordinates were first normalized to initiate at x, y coordinates (0,0) and end at (1,1). Next, the normalized

x -coordinate positions for false and true responses were captured at the 500-, 1,000-, and 1,500-msec processing mark and placed in corresponding time bins. The average location of each time bin for each condition is plotted in Figure 1C.

A 2 (prompt type: true vs. false) \times 3 (time bin: 500 vs. 1,000 vs. 1,500 msec) repeated measures ANOVA was used to evaluate trajectory position in real time. Beginning with the *no* trials, there were statistically significant effects for prompt type and time bin and a significant interaction between prompt type and time bin. To explore this interaction further, planned comparisons were conducted between prompt types at each time bin. There was a statistically significant difference of the x -coordinate position at the third time bin (1,500 msec) between false ($M = 0.23$, $SD = 0.27$) and true ($M = 0.47$, $SD = 0.23$) trajectories [$F(1,24) = 20.54$, $p < .001$].

The repeated measures results for the *yes* trials also showed statistically significant effects for prompt type and time bin and a significant interaction between them. Planned comparisons for the interaction revealed statistically significant differences of the x -coordinate position at the second time bin (1,000 msec) between false ($M = 0.06$, $SD = 0.10$) and true ($M = -0.31$, $SD = 0.20$) trajectories [$F(1,24) = 81.40$, $p < .001$] and at the third time bin (1,500 msec) between false ($M = -0.09$, $SD = 0.16$) and true ($M = -0.68$, $SD = 0.19$) trajectories [$F(1,24) = 203.29$, $p < .001$].

Taken together, these results indicate that the false *yes* trajectories diverged from the true *yes* trajectories much earlier (at around 1,000 msec) than when the false *no* trajectories began to diverge from the true *no* trajectories (not until at least 1,500 msec). Because *divergence* here refers to movement toward the competing response option, false *yes* trajectories appear to be influenced by a truth competition much earlier than do false *no* trajectories.

As in the trajectory shape analysis, false *yes* and *no* trajectories can be directly compared with each other instead of using the true trajectories as a reference, which have their own idiosyncratic response biases (as in a *yes* bias for *no* responses). As such, the trajectories for the false *yes* responses were mirror reversed (see Figure 1D), and paired t tests at each time bin were assessed. Once again, there was greater divergence for false *yes* responses than for false *no* responses, with divergence recorded at the second time bin [$F(1,24) = 4.31$, $p = .05$] and at the third time bin [$F(1,24) = 7.51$, $p = .01$].

Trajectory velocity. The velocity of response trajectories was evaluated by computing the distance (in pixels) covered each second within a moving window of eight x, y pixel coordinates across total time. Figure 2 shows the average velocity profile for each condition. This figure suggests that, on average, the initial increase in velocity (as the participants committed to a response) and the subsequent decrease in velocity (as the participants completed the response) occurred much later for false responses than for true responses. A repeated measures ANOVA conducted on the moment of peak velocity confirms this observation, such that a significant interaction [$F(1,24) = 25.51$, $p < .001$] and follow-up planned comparisons between

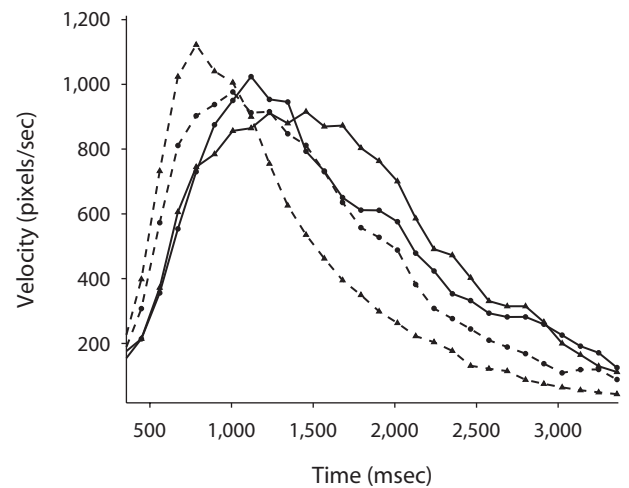


Figure 2. The velocity profiles plotted as a function of total time for true *no* (dashed line with circles), true *yes* (dashed line with triangles), false *no* (solid line with circles), and false *yes* (solid line with triangles) responses. The plotted range of 376–3,400 msec covers 90% of all completed trajectories and eliminates extremely early and late movements that have near-zero velocities. The average velocity profiles are defined over a broad range around the mean total time, because these profiles are partly based on slower trajectories that go beyond the mean total time—that is, trajectories with a total time that is longer than the mean total time.

prompt and response type reveal that false *no* responses peaked later than true *no* responses [$F(1,24) = 22.29$, $p < .001$] and that false *yes* responses peaked later than true *yes* responses [$F(1,24) = 93.32$, $p < .001$].

We also examined differences in the magnitude of peak velocity with the assumption that greater response activation would result in higher peaks. A statistically significant interaction was found between prompt and response type [$F(1,24) = 16.87$, $p < .001$], showing that the peak for false *yes* responses is lower in magnitude than that of true *yes* responses [$F(1,24) = 9.07$, $p < .01$]. There were no significant differences for false *no* and true *no* responses.

Trajectory properties. In this final analysis, we computed eight properties that characterize temporal and trajectory behavior along continuous scales of measurement (Dale et al., 2007). The variables are listed and summarized below.

Total time. The amount of time elapsed between the initiation of the prompt and making a *yes* or *no* response.

Latency. The amount of time that the mouse cursor stays in a latency region, with *region* defined as a 100-pixel radius that surrounds the mouse cursor position that initiated the response prompt (Dale et al., 2008).

Distance. The Euclidean distance traveled by the trajectory after leaving the latency region and making a *yes* or *no* response.

Motion time. The amount of time elapsed between the cursor's moving between the latency region and the participants' completing a *yes* or *no* response.

High x value. A measure of how close (in coordinate position) each trajectory curves toward the *NO* response box, with the *NO* response box at the maximum x -value position.

Table 1
Means and Standard Errors of the Wii Remote Trajectory Variables
by Prompt and Response Type

Variable	Yes				No			
	False		True		False		True	
	<i>M</i>	<i>SE</i>	<i>M</i>	<i>SE</i>	<i>M</i>	<i>SE</i>	<i>M</i>	<i>SE</i>
Total time (msec)	2,806	83	1,996	86	2,802	115	2,408	111
Latency (msec)	1,247	73	999	59	1,246	67	1,111	57
Distance (pixels)	1,423	102	996	98	1,426	145	1,273	114
Motion time (msec)	1,558	103	997	65	1,556	122	1,297	107
High <i>x</i> value	229	24	69	12	515	4	515	4
Low <i>x</i> value	-505	17	-509	22	-190	25	-139	2
<i>x</i> flips in latency	8.67	0.66	6.87	0.50	9.00	0.78	8.32	0.79
<i>x</i> flips in motion	3.87	0.36	2.20	0.24	4.22	0.48	3.42	0.39

Low x value. A measure of how close (in coordinate position) each trajectory curves toward the YES response box, with the YES response box at the minimum *x*-value position.

x flips in latency. The number of times that a trajectory moves back and forth on the *x*-axis within the latency region.

x flips in motion. The number of times that a trajectory moves back and forth on the *x*-axis while in motion to a *yes* or *no* response.

The total time and latency variables are primarily temporal measures, whereas the remaining variables capture dynamical processes that occur along the trajectory of motion. For example, *x* flips in latency and *x* flips in motion provide an intuitive measure of response instability, and the high *x* value and low *x* value variables are indicators of competing attractor strengths that occur when the participants react either to *no* falsely (moving leftward on the *x*-axis toward the competing YES response region) or to *yes* falsely (e.g., moving rightward on the *x*-axis toward the competing NO response region).

A 2 (prompt type: false vs. true) × 2 (response type: *yes* vs. *no*) repeated measures ANOVA was conducted for each of the eight dependent variables. Each variable, its mean value, and its standard error for each condition are provided in Table 1. The results of the repeated measures ANOVA are provided in Table 2.

To ensure that the results for trajectory properties were not unduly influenced by possible confounds, we also performed a hierarchical linear model (HLM) analysis

with random factors that controlled for variance due to practice effects (by nesting trial number in participants), for NO/YES response position, and for items. A noted advantage of an HLM analysis is that it is also quite robust against unequal *ns* in conditions, as is the case with our data. In all analyses, the results were consistent with the repeated measures ANOVA, except that the interaction terms for latency and *x* flips in latency were no longer significant.

DISCUSSION

The present study is the first investigation into the action dynamics of deceptive behavior. The movements of the arm revealed distinct signatures of cognitive activity as the participants made false and true responses to autobiographical questions. During false responses, the dynamics were slower and more disorderly than those during true responses, and they were also curved toward a competitor true region that was visually copresent with the target response region. This curvature suggests the presence of a truth-bias attractor that pulls processing off course during the production of a false response (Gilbert et al., 1993; McKinstry, Dale, & Spivey, 2008). For truthful responses, there was no equivalent pull in the direction of a competitor false region.

The competition effects for false responding are similar to those in decision tasks that involve competition between featurally similar exemplars (Dale et al., 2007), ambiguous syntactic completions (Farmer, Cargill, Hindy, Dale, & Spivey, 2007), and phonological competitors (Spivey et al., 2005). As in these studies, the competition exhibited in false responding is the result of cognitive components that evolve smoothly over the response movement itself, suggesting overlapping processes in overcoming an initial belief and generating a false response. This view of deception naturally extends response time measures by incorporating the fine-grained changes that occur during the response. By doing so, a clearer distinction between deceptive and truthful behavior is possible.

One notable distinction is the greater trajectory curvature and slower responses for false *yes* responses than for false *no* responses. This differentiation within the false responses was not found for the truth responses. Interestingly, the greater difficulty of falsely responding *yes*

Table 2
F Scores for the Repeated Measures Analysis
Using Wii Remote Movements

Variable	Yes Versus No Response	True Versus False Prompt	Prompt <i>x</i> Response
Total time (msec)	9.92**	80.67**	11.85**
Latency (msec)	4.68*	25.82**	
Distance (pixels)	6.05*	16.03**	4.70*
Motion time (msec)	8.04**	44.12**	6.90*
High <i>x</i> value	693.63**	68.38**	72.72**
Low <i>x</i> value	288.49**	7.18*	6.77*
<i>x</i> flips in latency	5.09*	12.83**	
<i>x</i> flips in motion	9.41**	38.50**	

Note—The degrees of freedom for all analyses were 1,24. **p* < .05. ***p* < .001.

is at odds with an earlier finding that normal *yes* responses elicit faster and smoother trajectories than do *no* responses—a *yes* bias (McKinstry et al., 2008). Indeed, on the basis of Gilbert and colleagues' (Gilbert, 1991; Gilbert et al., 1993) findings, automatic acceptance of propositions should give *yes* responses a facilitative advantage. However, when deception is involved, this *yes* automaticity conflicts with the more deliberative goal to respond falsely.

To lay the initial foundation in quantifying deceptive response movements, we chose a guided lie paradigm that generally permits a straightforward contrast between false and true response behaviors. Unfortunately, this distinction is not always realized in real-world scenarios, in which elements of truth are intermingled with the motives and content of a lie. There is also a limitation in our experimental paradigm that concerns the intention to deceive. Ekman (1997) argued that deception is an act of conscious volition that requires the deceiver to know what is accurate and then to purposefully violate that knowledge with false information. Clearly, deception requires a certain degree of motivation that is absent from a guided lie paradigm. Nevertheless, this type of responding is still closely aligned with deceptive behavior and is widely used and accepted in the deception literature (DePaulo et al., 2003; Vrij, 2001). Of course, consensus does not negate further investigation, and future work will allow participants greater choice in the scope of their deception, both in when they deceive and under what circumstances they do it (e.g., personal or social gain, avoiding embarrassment).

It is clear that deception is a complex behavior that garners both theoretical and applied attention. To concoct a false reality requires one to maintain a mental representation of the truth and then to violate this representation with all appearances of sincerity. Not only is this behavior cognitively challenging, but it also interacts with a host of social, motivational, and emotional factors. Notwithstanding this complexity, researchers have devised a multitude of techniques to identify cues of deception. This study provides the first steps toward applying an action dynamics framework to the exploration of false response behavior. The results suggest that dynamic measures capture deceptive processes that are unavailable to response time measurements alone. If so, these measures could improve existing prediction models that have been touted in recent years (Gregg, 2007; Sartori et al., 2008; Walczyk, Mahoney, Doverspike, & Griffith-Ross, 2009), as well as supplement techniques for detecting online deception (Monroe & Rubin, 2000). Although there is much more work ahead, we admonish deceivers everywhere: Your arm might just reveal when you are lying.

AUTHOR NOTE

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NOTE

1. <http://actiondeception.nickduran.com/>, as well as in the supplemental materials.

SUPPLEMENTAL MATERIALS

Supplemental figures depicting the apparatus and a sample trial sequence, as well as a list of the sentence stimuli, may be downloaded from <http://pbr.psychonomic-journals.org/content/supplemental>.

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