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Sensorimotor brain dynamics reflect architectural affordances

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Anticipating meaningful actions in the environment is an essential function of the brain. Such predictive mechanisms originate from the motor system and allow for inferring actions from environmental affordances, and the potential to act within a specific environment. Using architecture, we provide a unique perspective on the ongoing debate in cognitive neuroscience and philosophy on whether cognition depends on movement or is decoupled from our physical structure. To investigate cognitive processes associated with architectural affordances, we used a mobile brain/body imaging approach recording brain activity synchronized to head-mounted displays. Participants perceived and acted on virtual transitions ranging from nonpassable to easily passable. We found that early sensory brain activity, on revealing the environment and before actual movement, differed as a function of affordances. In addition, movement through transitions was preceded by a motor-related negative component that also depended on affordances. Our results suggest that potential actions afforded by an environment influence perception.

sensorimotor | predictive processing | mobile brain/body imaging | architectural cognition | mobile EEG

The affordance of a given spatial environment—defined as the perception of possibilities for, or restraints on, an action that the environment offers—is essential for an agent to produce meaningful behavior. Thus, the affordance of the spatial environment become a central concept for humans interacting with their world. The term “affordance” was introduced by Gibson (1) and subsequently refined by various authors, including Clark (2), who defined it as “the possibilities for use, intervention, and action which the physical world offers a given agent and are determined by the ‘fit’ between the agent’s physical structure, capacities, and skills and the action-related properties of the environment itself.” In light of emerging theories of embodied cognition, the perception of the environment may be dependent on proprioceptive mechanisms. According to predictive processing, a neuroscientific-based theory of embodied cognition (3–5), motor systems, similar to perceptual processes, aim at canceling out continuously incoming bottom-up sensory signals with top-down predictions. In this perspective, movement emerges as a result of an active inference that attempts to either minimize motor trajectory prediction errors by acting, and thus perceiving the unfolding of the predicted movement, or by changing perception itself (6–8).

From the standpoint of active inference, motor systems suppress errors through a dynamic interchange of prediction and action. In other words, there are two ways to minimize prediction errors: to adjust predictions to fit the current sensory input and to adapt the unfolding of movement to make predictions come true. This is a unifying perspective on perception and action suggesting that action is both perceived by and caused by perception (9). Thus, action, perception, and cognition coordinate to move the body in ways that conform to a transitional set of expectations (10).

The claim that we seek to investigate in the present study is that perception is rooted in action, creating an action–perception loop informed by dynamically (top-down/bottom-up) generated prediction errors. Ultimately, the argument is that perception is not the sole result of sensing the physical world but unfolds as an ongoing interaction between sensory processes and bodily actions. Such a claim has philosophical and neuroscientific significance, because the neural dynamics underlying perception would be intimately dependent on the affordances of a given environment.

To investigate this claim further, we used electroencephalography (EEG) recordings to address the neural dynamics of action–perception interactions through affordance manipulations in architectural experiences. More specifically, we investigated the affordances of transitions, which are ideal candidates due to their dynamic nature concerning the duration of altering one condition to another (11). Here we confine transitions to the passage between spaces, which according to the enactivists’ proposed action–perception loop will be an experience dependent on the affordances offered by the passage itself. Because of the dynamic nature of architecture, an essential part of transitions and experiencing architecture is the ability to act (12). Traditionally, investigations of architectural experiences are phenomenological—the description of phenomena in how experience gives access to

Significance

Using electroencephalography and virtual reality, our research provides a unique perspective on the centuries-old open-ended debate in cognitive neuroscience and philosophy on the relationship among cognition, movement, and environment. Our results indicate that cortical potentials vary as a function of bodily affordances reflected by the physical environment. First, the results imply that cognition is inherently related to the potential movement of the body; thus, we posit that action is interrelated with perception, actively influencing the perceivable environment. Second, these results indicate that moving in space is to continuously construct a prediction of a world of affordances, suggesting that architects take up the continuity of spaces, given that the unfolding of bodily movement alters perception and experience.

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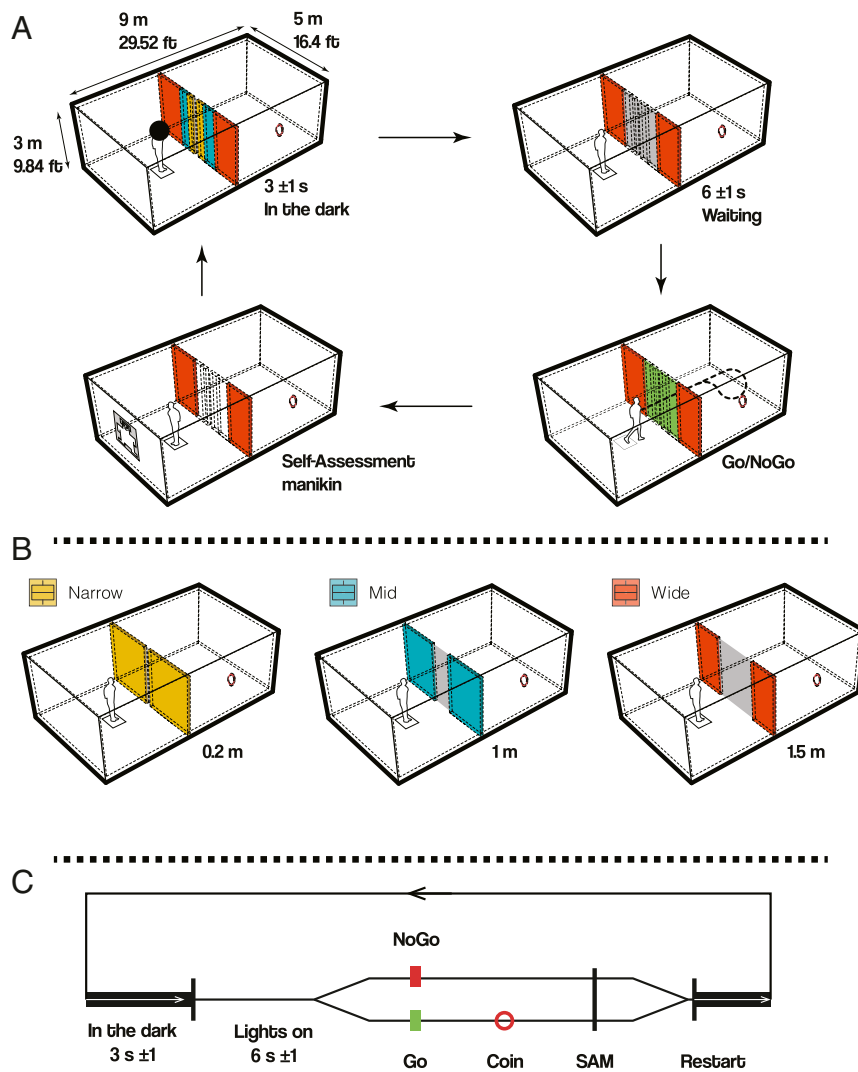


Fig. 1. (A) Participants are instructed to stand in the start square. A black sphere restricts their vision to pure black for 3 s ($\sigma = 1$ s). The moment the black sphere disappears, participants perceive the door they have to pass. They wait for the imperative stimulus, either a green door (*Go*) or a red door (*NoGo*), for 6 s ($\sigma = 1$ s). In the case of *Go*, participants were instructed to pass the opening, virtually touch the red circle (which in turn releases a monetary bonus), return to the start square, and complete the virtual SAM questionnaire. In the case of *NoGo*, participants were instructed to turn around and complete the virtual SAM questionnaire. (B) The three different doors had the following dimensions: *Narrow*, 0.2 m; *Mid*, 1 m; *Wide*, 1.5 m. Note the color code for each door as used throughout the paper. (C) Diagrammatic timeline depicting the sequences of events for a single trial in a conceptual manner.

trial always required approaching the door. When a participant touched the surrounding walls, the walls turned, indicating that the participant failed to pass and thus must return to the start square, complete the virtual SAM, and start the next trial.

Participants would quickly notice that the narrow door (20 cm) was impossible to pass without producing the warning feedback that they have failed to pass, and yet they were required to try passing. All participants underwent a training phase to become accustomed to the VR environment and the different conditions. The experimenter observed the participants from a control room, separated from the experimental space, using two cameras and a mirrored display of the VR environment to reduce interactions to a minimum during the experiments.

Subjective and Behavioral Data. To investigate the subjective experience of the task, we introduced the participants to a virtual SAM questionnaire after each trial. The SAM is a pictorial assessment of pleasure, arousal, and dominance on a 5-point Likert scale (27). The manikin display ranges from smiling to frowning (pleasure), from a dot in the stomach to an explosion (arousal), and from being very small to very big (dominance). Participants were asked to self-assess their current state after each trial. Furthermore, regarding behavioral measures, we recorded the reaction time from the onset of the *Go* stimulus (door color change) to reaching the opening

threshold itself, to assess the behavior. The data were analyzed using analysis of variance (ANOVA) with the width of the doors as a repeated-measures factor. In the case of violation of normality and homogeneity, corrected *P* values are reported. For post hoc analysis, the data were contrasted using Tukey's honest significant differences (HSD) test.

EEG Recording and Data Analysis. We investigated the impact of transitional affordances on human cognition and brain dynamics using a MoBI approach (24–26, 28) recording human brain dynamics in participants actively transitioning through virtual rooms. All data streams were recorded and synchronized using LabStreamingLayer (35). Participants wore a backpack, which held a high-performance gaming computer to render the VR environment (Zotac; PC Partner Limited) attached to two batteries and an EEG amplifier system. We combined a Windows Mixed Reality (WMR; 2.89", 2,880 × 1,440 resolution, update rate at 90 Hz, 100° field of view with a weight of 440 g, linked to the Zotac computer through HDMI) headset and one controller (Acer) to display and interact with the virtual environment based on Unity (Fig. 2). Events for recordings of performance and physiological data were triggered by the position of the participant in the tracking space or by the respective response buttons of the remote control. Specific events, such as touching the wall, all button presses, transitioning through the door, answering the questionnaire, and all cases of "lights on" (and

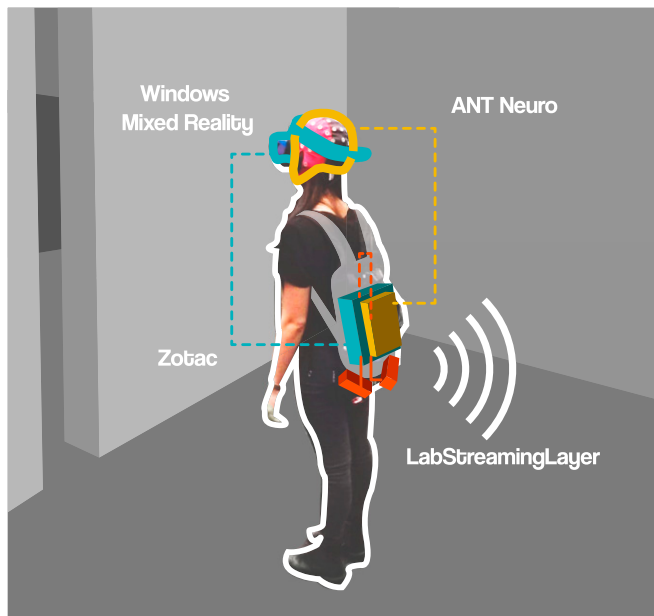


Fig. 2. MoBI setup. The participants wore a backpack, carrying a high-performance gaming computer (Zotac, cyan), powered by two batteries (red). An EEG amplifier (ANT eegoSports, yellow) was attached to the backpack and connected to the computer. The participants wore a VR head-mounted display (Windows Mixed Reality) on top of a 64-channel cap. This setup allowed participants to move freely around while recording data.

“lights off”), were synchronized with the recorded brain activity and the presented VR environment through LabStreamingLayer.

EEG data were acquired continuously with a 64-channel EEG system (eegoSports, ANT Neuro), sampled at 500 Hz. Impedances were kept below 10 k Ω . The computational delay generated by the interaction of ANT Neuro software, Windows Mixed Reality, and Unity was measured as 20 ms ($\sigma = 4$ ms), which was taken into account during the analysis by subtracting the average delay from each event latency. With a jitter of 4 ms, we considered the delay to have little to no impact on the ERPs. Offline analysis was conducted using MATLAB (MathWorks) and the EEGLAB toolbox (36). The raw data were bandpass-filtered between 1 Hz and 100 Hz and down-sampled to 250 Hz. Channels with more than five SDs from the joint probability of the recorded electrodes were removed and subsequently interpolated. The datasets were then re-referenced to an average reference and adaptive mixture independent component analysis (ICA) (37) was computed on the remaining rank of the data using one model with online artifact rejection in five iterations. The resultant ICA spheres and weights matrices were transferred to the raw dataset that was preprocessed using the identical preprocessing parameters as for the ICA dataset, except for the filtering, which used a bandpass filter from 0.2 Hz to 40 Hz. Subsequently, independent components reflecting eye movements (i.e., blinks and horizontal movements) were removed manually based on their topographic, spectral, and temporal characteristics.

Epochs were created time-locked to the onset of the room including the closed door (lights on) from -500 ms before to $1,500$ ms after stimulus onset for *Narrow*, *Mid*, and *Wide* door trials. Similarly, another set of epochs was time-locked to the second *Go/NoGo* stimulus from -500 ms before to $1,000$ ms after onset of the stimulus for *Narrow*, *Mid*, and *Wide* door trials. On average, 15% ($\sigma = 10.8$) of all epochs were automatically rejected when they deviated by >5 SDs from the joint probability and distribution of the activity of all recorded electrodes.

The visual evoked potentials and MRCPs were analyzed at central midline electrodes (*Fz*, *FCz*, *Cz*, *Pz*, *POz*, and *Oz*) covering all relevant locations, including the visual cortex and the motor cortex, as reported previously (31, 38). Because stimuli were distributed across the complete visual field and participants walked through the virtual spaces, we did not expect to see any lateralization of ERPs. All channels were analyzed; however, only three channels (*FCz*, *Pz*, and *Oz*) are discussed here, according to findings reported by Bozzacchi et al. (31). The analysis results of all six channels are provided in *SI Appendix*. For peak analysis of the P1-N1 complex, the grand average peaks were estimated, and individual peaks were defined as the maximum

positive peak and negative peak in the time window surrounding the grand average P1 and N1 peaks (± 10 ms from the peak), respectively. An automatic peak detection algorithm detected the peaks in the averaged epochs for each participant. Multiple peaks were detected and systematically weighted depending on the magnitude, the distance to the grand average peak latency as determined by visual inspection of grand average ERP, and the polarity. The algorithm is provided in *SI Appendix*. For anterior N140 and posterior P140, by visual inspection of the grand average ERPs, the estimated grand average latency was 140 ms, with a search window for individual peaks ranging from 50 to 200 ms. For the anterior P215 and posterior N215, the estimated grand average peak latency was 215 ms, with a search window for individual peaks ranging from 140 to 290 ms.

Mean peak amplitudes were analyzed by 3×3 repeated-measures ANOVA using the door width (*Narrow*, *Mid*, or *Wide*) and electrode as repeated measures. The results descriptions focus on the visual evoked P140 component at posterior electrodes (*Pz*, *POz*, and *Oz*) and the N140 component at frontal leads (*Fz*, *FCz*, and *Cz*) based on separate ANOVAs. For the N215 and P215 components at posterior electrodes (*Pz*, *POz*, and *Oz*) and frontal leads (*Fz*, *FCz*, and *Cz*), separate ANOVAs were computed in the time range of 140–290 ms. For the later motor-related potentials, an ANOVA was computed for the mean amplitude in the range of 600–800 ms. The data were analyzed using a $2 \times 3 \times 6$ factorial repeated-measures ANOVA with the factors imperative stimulus (*Go* and *NoGo*), door width (*Narrow*, *Mid*, and *Wide*), and electrode location (*Fz*, *FCz*, *Cz*, *Pz*, *POz*, and *Oz*) within the time window (600–800 ms). For post hoc analysis, the data were contrasted using Tukey’s HSD. In cases of violation of sphericity, corrected *P* values are reported. All ANOVAs were computed as linear mixed models.

Results

We obtained subjective, behavioral, and electrophysiological data, with a focus on electrophysiology. All data underwent statistical processing and are presented here in the order processed.

Subjective Data: SAM Ratings. The SAM questionnaire was completed for *Go* or *NoGo* and for all door conditions. A 2×3 factorial repeated-measures ANOVA with the factors imperative stimulus (*Go* and *NoGo*) and door width (*Narrow*, *Mid*, and *Wide*) for each emotional dimension of the SAM questionnaire revealed differences in the main effect for width in Arousal ($F_{2,90} = 3.35$, $P = 0.0393$, $\eta^2 = 0.048$), Dominance ($F_{2,90} = 10.03$, $P < 0.0001$, $\eta^2 = 0.138$), and Valence ($F_{2,90} = 5.31$, $P = 0.0065$, $\eta^2 = 0.073$). For the imperative stimulus, differences were found for Arousal ($F_{1,90} = 36.81$, $P < 0.0001$, $\eta^2 = 0.266$), Dominance ($F_{1,90} = 25.26$, $P < 0.0001$, $\eta^2 = 0.173$), and Valence ($F_{1,90} = 28.59$, $P < 0.0001$, $\eta^2 = 0.196$). Interaction effects revealed significant differences for Dominance ($F_{2,90} = 4.14$, $P = 0.0189$, $\eta^2 = 0.056$) and Valence ($F_{2,90} = 7.04$, $P = 0.0014$, $\eta^2 = 0.096$) but only tendencies for Arousal ($F_{2,90} = 0.92$, $P = 0.4000$, $\eta^2 = 0.0134$). Post hoc contrasts using Tukey’s HSD test (Fig. 3) showed no significant differences for *NoGo* in Arousal but identified significant differences for *Go* between *Narrow* \times *Mid* ($P = 0.0386$). For *NoGo* in Dominance, no significant differences were revealed as opposed to *Go* for *Narrow* \times *Wide* ($P < 0.0001$), *Mid* \times *Wide* ($P = 0.0335$), and *Narrow* \times *Mid* ($P < 0.0345$). Similarly, for valence, in *Go* significant differences were revealed for *Narrow* \times *Mid* ($P = 0.0326$), *Narrow* \times *Wide* ($P < 0.0001$), with a tendency seen for *Mid* \times *Wide* ($P = 0.0625$).

Behavioral Data: Door Approaching Times. This analysis was possible only for *Go* trials, as it required actually approaching the door. The time it took participants from the *Go* stimulus to pass the door was calculated using one-way ANOVA with repeated measures for different door widths, which revealed a significant difference in door widths ($F_{2,36} = 6.07$, $P < 0.0053$, $\eta^2 = 0.232$; Fig. 4). Post hoc comparison with Tukey’s HSD test showed no significant differences in behavior when approaching the *Narrow* doors compared with the *Mid* doors ($P = 0.3073$), but had a tendency to be slower when approaching *Mid* doors compared with *Wide* doors ($P = 0.1312$) and a significant difference between approaching *Narrow* doors compared with *Wide* doors

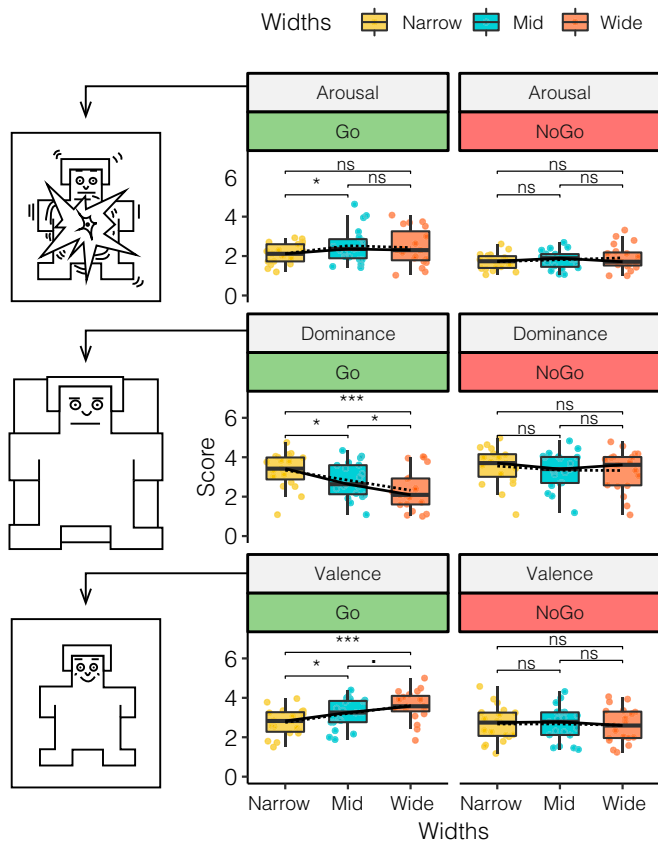


Fig. 3. Boxplot of the SAM questionnaire results for the three different SAM scales (Arousal, Dominance, and Valence) as a function of the door width (*Narrow*, *Mid*, or *Wide*). (Left) Pictorial representation of the SAM manikin for the highest value of each condition presented. (Middle) SAM ratings for the *Go* condition. (Right) SAM ratings for the *NoGo* condition. Means are indicated by dashed lines; medians, by solid lines. * $P < 0.1$, * $P < 0.05$, and *** $P < 0.001$; ns, not significant.

($P = 0.0038$), with significantly faster approach times for the *Wide* door condition.

Electrophysiology: Early ERP.

Posterior P140. With onset of the lights that allowed participants to see the room including the door (i.e., lights on), the ERPs clearly demonstrated a P1-N1 complex that was most pronounced over the occipital midline electrode, with a first positive component around 140 ms, followed by a negative peak around 210 ms (Fig. 5; all six channels shown in *SI Appendix*, Fig. S1). At the frontal midline electrode, this pattern was inverted, and a negative component around 140 ms was followed by a positive peak observed around 215 ms. The 3×3 repeated-measures ANOVA on P140 amplitudes for posterior channels revealed significant main effects for both door width ($F_{2,108} = 8.163$, $P = 0.005$, $\eta^2 = 0.096$) and channel ($F_{2,36} = 15.868$, $P < 0.0001$, $\eta^2 = 0.187$). The interaction effect was not significant ($F_{4,108} = 1.669$, $P = 0.1624$). Post hoc comparisons using Tukey's HSD test revealed significant differences in peak amplitudes at channel *Oz* between *Narrow* and *Mid* transitions ($P = 0.0021$) and between *Narrow* and *Wide* transitions ($P = 0.0065$) and at channel *POz* between *Narrow* and *Wide* transitions ($P = 0.028$).

Posterior N215. The 3×3 repeated-measures ANOVA on N215 amplitudes for posterior channels revealed a significant main effect for the factor door width ($F_{2,108} = 4.348$, $P = 0.0153$, $\eta^2 = 0.066$) but no significant impact for the factor channels ($F_{2,36} = 0.0893$, $P = 0.9147$, $\eta^2 = 0.001$). Post hoc Tukey HSD contrasts revealed no significant differences for *Pz* and *POz*.

However, similar to posterior P140, significant differences at *Oz* for the comparison of *Narrow* and *Mid* transitions ($P = 0.0113$) and for the comparison of *Narrow* and *Wide* transitions ($P = 0.0372$) were found (Fig. 6).

Anterior P215. An inverse pattern was observed for amplitudes over anterior leads, with a main effect of door width that differed depending on the affordances ($F_{2,108} = 11.071$, $P < 0.0001$, $\eta^2 = 0.139$). The main effect of channels also reached significance ($F_{2,36} = 5.3627$, $P = 0.0092$, $\eta^2 = 0.067$). Tukey HSD contrasts revealed significant differences only between *Narrow* and *Wide* transitions for *FCz* ($P = 0.0071$) and *Cz* ($P = 0.0214$), with a tendency at *Fz* ($P = 0.0717$). The interaction was not significant.

Anterior N140. The 3×3 repeated-measures ANOVA on N140 amplitudes for anterior channels revealed no significant main effect for the factor door width ($F_{2,108} = 1.823$, $P = 0.1663$, $\eta^2 = 0.024$). In contrast, the main effect of channels reached significance ($F_{2,108} = 8.109$, $P = 0.0012$, $\eta^2 = 0.107$). The interaction did not reach significance.

EEG-motor-related processes. After onset of the imperative stimulus, a positive peak at anterior leads and a negative peak at posterior leads were observed. For the sake of brevity, this potential complex is referred to as the early postimperative complex (EPIC). Reflecting its similar cortical polarity to the P1-N1 complex, the EPIC was analyzed in a similar way, separating anterior leads (*Fz*, *FCz*, and *Cz*) from posterior leads (*Pz*, *POz*, and *Oz*), and detecting single peaks in individual averages.

Anterior EPIC. A $2 \times 3 \times 3$ repeated-measures ANOVA revealed significant differences in the main effect for widths ($F_{2,270} = 4.21$, $P = 0.0157$, $\eta^2 = 0.025$), imperative stimulus ($F_{1,270} = 23.66$, $P < 0.0001$, $\eta^2 = 0.071$), and channel ($F_{2,36} = 6.70$, $P = 0.0033$, $\eta^2 = 0.040$). No interaction effect was observed. The post hoc Tukey's HSD test revealed no significant differences between the transition widths for the various channels and for the imperative stimuli.

Posterior EPIC. The identical ANOVA for the posterior potentials of the EPIC revealed no significant impact of transition width ($F_{2,270} = 2.001$, $P = 0.1371$, $\eta^2 = 0.013$) or imperative stimulus ($F_{1,270} = 2.30$, $P = 0.1298$, $\eta^2 = 0.007$). Significant differences in EPIC amplitudes were observed for the factor channel ($F_{2,36} = 5.45$, $P = 0.0085$, $\eta^2 = 0.035$). Because topographical differences were not the focus of this study, no further post hoc contrasts were computed. No interaction was significant.

PIV. In the preparation time before the onset of the door color change, indicating that the participant was either to walk through the door or remain in the same room, we observed no systematic negative going waveform as reported in previous studies (29, 39). However, after the onset of the color change, a pronounced

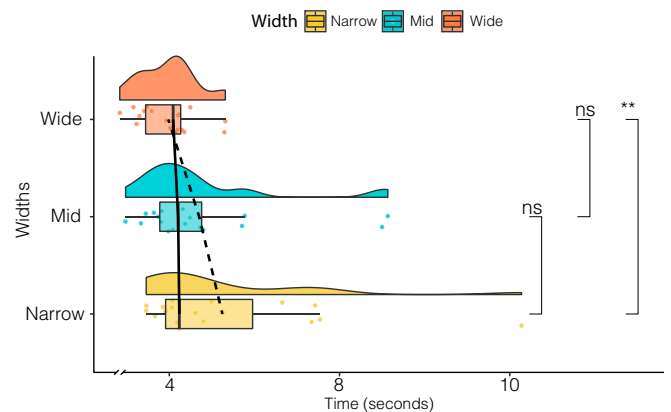


Fig. 4. Raincloud plot of approach times for each door width condition. Post hoc comparisons using Tukey's HSD test. Means are indicated by dashed lines; medians, by solid lines. ** $P < 0.01$; ns, not significant.

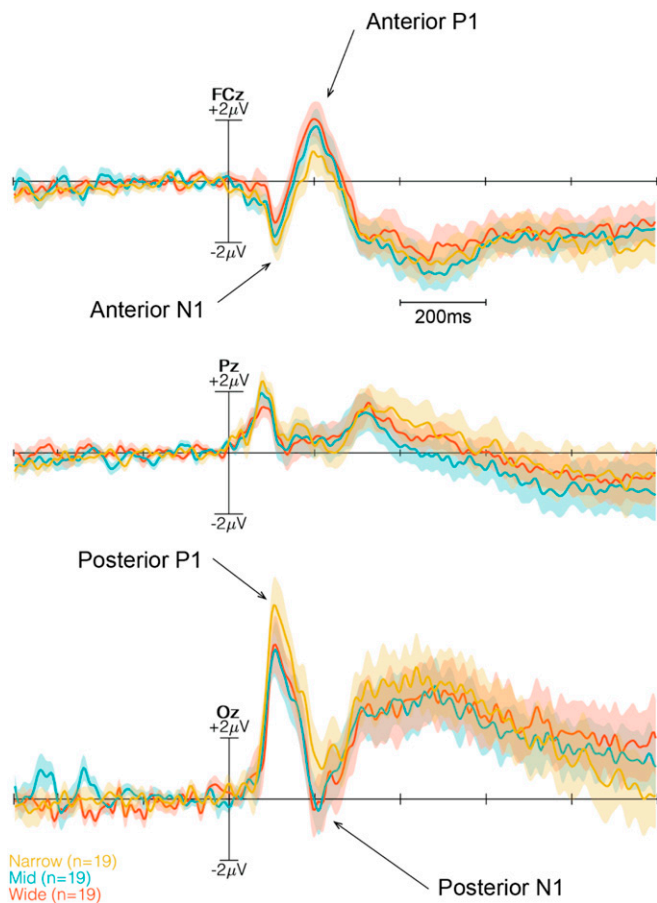


Fig. 5. Three time-locked ERPs (FCz, Pz, and Oz) at the onset of the lights on event. The *Narrow* condition is in yellow, the *Mid* condition is in blue, and the *Wide* condition is in red. Two time windows are indicated with dashed lines and a gray transparent box. The first time window (50–200 ms) marks the anterior N140 and posterior P140, while the second window (140–290 ms) marks the anterior P215 and posterior N215. The components are marked with arrows.

positivity, the EPIC, followed by a long-lasting negative waveform over frontocentral locations was observed in the ERP (Fig. 7; all six channels shown in *SI Appendix*, Fig. S2). This negative waveform resembled a PINV, as described previously (32, 34, 40). The PINV component was observed at 600–800 ms after the imperative stimulus (color change of the door) and varied as a function of the affordance of the environment (door width). A global $2 \times 3 \times 6$ factorial repeated-measures ANOVA was computed to analyze the MRCs using *Go/NoGo*, width, and channel as repeated measures. ANOVA revealed significant differences in the main effect for *Go/NoGo* ($F_{1,540} = 19.54$, $P < 0.0001$, $\eta^2 = 0.039$) and for channel ($F_{5,90} = 16.69$, $P < 0.0001$, $\eta^2 = 0.112$). Significant differences were reported for the interaction effect of *Go/NoGo* \times channel ($F_{5,540} = 5.25$, $P = 0.0001$, $\eta^2 = 0.035$) and for width \times channel ($F_{10,540} = 2.61$, $P = 0.0042$, $\eta^2 = 0.035$). A tendency toward an interaction of the factors *Go/NoGo* \times Width ($F_{2,540} = 2.33$, $P = 0.0975$, $\eta^2 = 0.006$) was observed.

Post hoc contrasts using Tukey's HSD test revealed significant differences only for the *Go* condition as opposed to the *NoGo* condition (Fig. 8). Similar to the early evoked potentials, differences were observed only at frontal and occipital sites and between *Narrow* and *Mid* doors over FCz ($P = 0.0059$) and Oz ($P < 0.0001$), as well as between *Narrow* and *Wide* doors at FCz ($P = 0.0323$) and Oz ($P < 0.0001$). No differences were observed between the *Mid* and *Wide* doors (*SI Appendix*, Fig. S3).

Discussion

The main goal of this study was to assess whether brain activity is altered depending on the affordances offered by the environment. If this were the case, then affordances should systematically modulate behavior and brain activity. Specifically, we hypothesized that perceptual processes would covary with the environmental affordances, leading to behavioral changes, and that MRCs would vary as a function of affordances.

SAM and Approach Time. The analysis of subjective ratings revealed significant differences among *Go* trials but no differences across *NoGo* trials for all ratings. Notably, in cases of *NoGo*, all participants perceived a similar scene standing in front of a red (*NoGo*) door, turning around, and answering the virtual SAM. Varying door sizes for *Go* trials yielded differences for Dominance, with *Narrow* doors more dominant than *Mid* doors and even more dominant than *Wide* doors. The increase in Dominance for *Narrow* doors is inversely reflected in Valence because we observed increasing values with increasing door widths. Regarding Arousal, participants reported less arousal for *Narrow* doors compared with *Mid* and *Wide* doors. Furthermore, it is noteworthy that Dominance for *NoGo* is relatively high in value compared with *NoGo* in Arousal and Valence, which score low and central values, respectively. Taken together, these findings indicate that subjective reports differ significantly depending on whether participants who received a *Go* actively moved through the rooms, implying an impact of action affective ratings of an environment. However, our findings should be considered with caution, as the subjective ratings might have been influenced by several factors beyond affordance, including monetary reward, different trial durations, physical activity, and varying skills of subjective/introspective emotional evaluation.

Thus, performance data might provide a better basis for interpreting the impact of affordances on behavior. The time it took participants to reach the door after the onset of the imperative color change varied according to the environmental affordance. Participants approached the *Wide* doors nearly significantly faster and significantly faster than the *Mid* and *Narrow* doors, respectively, while there was no significant difference between *Mid* and *Narrow* transitions. While the *Wide* door clearly offered a passage without greater demands regarding the motor plan and execution, the *Mid* door width, being ambiguously wide/narrow, might have triggered motor processes simulating a transition to estimate whether the door was passable. In this sense, the *Mid* and *Narrow* doors, causing uncertainty, might have delayed approach times due to increased processing demands. Admittedly, results derived from the approach time are limited, due in part to fatigue from performing a physically demanding task for a relatively long period, and in part to passing a door that is seemingly impossible to pass. This led participants to develop different approach strategies, such as twisting their bodies, peeking inside from different angles, or walking directly into the virtual wall to trigger a failed attempt, causing different delays. Given that no participant was told beforehand that one opening was impassable, participants' enthusiasm and creativity diminished over the course of the trials when they learned that it was the *Narrow* door. However, the fact that participants in general spent significantly more time approaching the *Narrow* doors compared with *Wide* doors provides sufficient guidance for the analyses of cortical measures associated with these differences.

Cortical Measures.

Early evoked potentials. As an initial insight into the association of affordances and cortical potentials, we analyzed the early visual-evoked potentials. We expected to find differences in the stimulus-locked ERP at occipital channels, reflecting differences in sensory processing of affordance-related aspects of the transition. Based on the assumption of fast sensorimotor active inferences that should

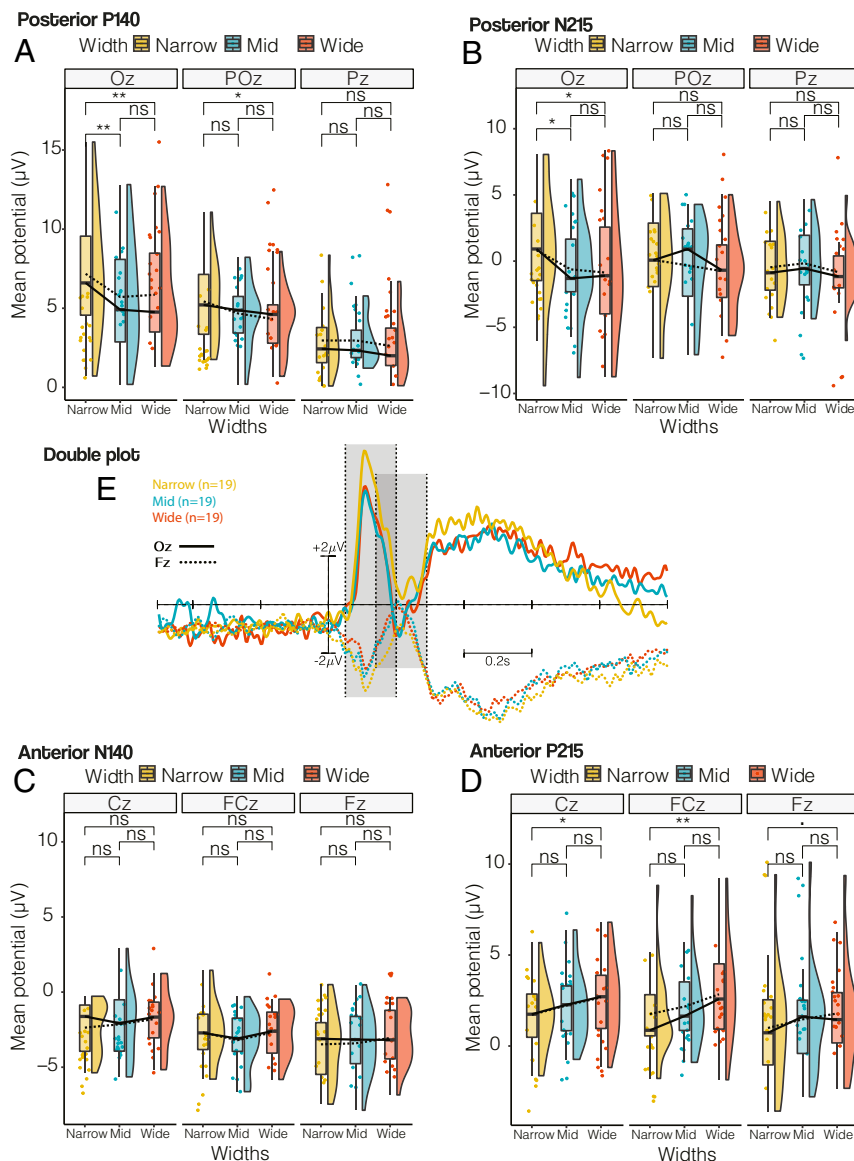


Fig. 6. (A) Posterior P140. Raincloud plot of detected mean amplitude of the positive peak in the time-locked lights on event in the time range of 50–200 ms for Pz, POz, and Oz. Means are indicated by dashed lines; medians, by solid lines. The significance was calculated using Tukey’s HSD test. We observed significant differences for Oz in *Narrow × Mid* ($P = 0.0021$) and *Narrow × Wide* ($P = 0.0065$), while POz in *Narrow × Wide* revealed a significant difference ($P = 0.028$); however, no significant differences were observed in other electrodes and other contrasts. (B) Posterior N215. Raincloud plot of detected mean amplitude of the negative peak in the time-locked lights on event in the time range of 140–290 ms for Pz, POz, and Oz. We observed significant differences only for Oz in *Narrow × Mid* ($P = 0.0113$) and *Narrow × Wide* ($P = 0.0372$). (C) Anterior N140. Raincloud plot of detected mean amplitude of the negative peak in the time-locked lights on event in the time range of 50–200 ms for Fz, FCz, and Cz. We observed no significant differences for any electrode. (D) Anterior P215. Raincloud plot of detected mean amplitude of negative peak in the time-locked lights on event in the time range of 140–290 ms for Fz, FCz, and Cz. We observed significant differences in all channels in *Narrow × Wide*, with the exception of only a tendency in Fz ($P = 0.0717$), FCz ($P = 0.0071$), and Cz ($P = 0.0214$). (E) Double plot. Frontal (dashed line) and posterior (solid line) time-locked ERPs (Fz and Oz) at the onset of the lights on event. The *Narrow* condition is in yellow, the *Mid* condition is in blue, and the *Wide* condition is in red. Two time windows are indicated with dashed lines and a gray transparent box. The first time window (50–200 ms) marks the anterior N140 and posterior P140, while the second window (140–290 ms) marks the anterior P215 and posterior N215. * $P < 0.1$, * $P < 0.05$, and ** $P < 0.01$; ns, not significant.

be reflected in action-directed stimulus processing influencing not only sensory activity, but also motor-related activity, we hypothesized that we would also find differences in the ERP over motor areas in the same time window as sensory potentials (i.e., between 50 and 200 ms). As illustrated in the analysis, we found significant differences in amplitudes of the visually evoked P140 component over the central occipital electrode varying with the affordance of the transition. In addition, and in line with our hypothesis, we also found a difference over frontocentral leads starting around 50 ms and lasting until 200 ms after onset of the door display. Taken

together, these findings indicate that no significant differences in peak amplitudes were found between the passable *Mid* and *Wide* doors, while peak amplitudes associated with both door widths significantly differed from those of the impassable *Narrow* doors.

Note that the visual scene of the three doors was comparable, as they contained the same physical contrasts in the *Go* and the *NoGo* condition. In addition, being merely introduced to the environmental setting, participants did not know whether they would have to attempt to pass. These results indicate that impassable doors with poor affordances produce significantly different

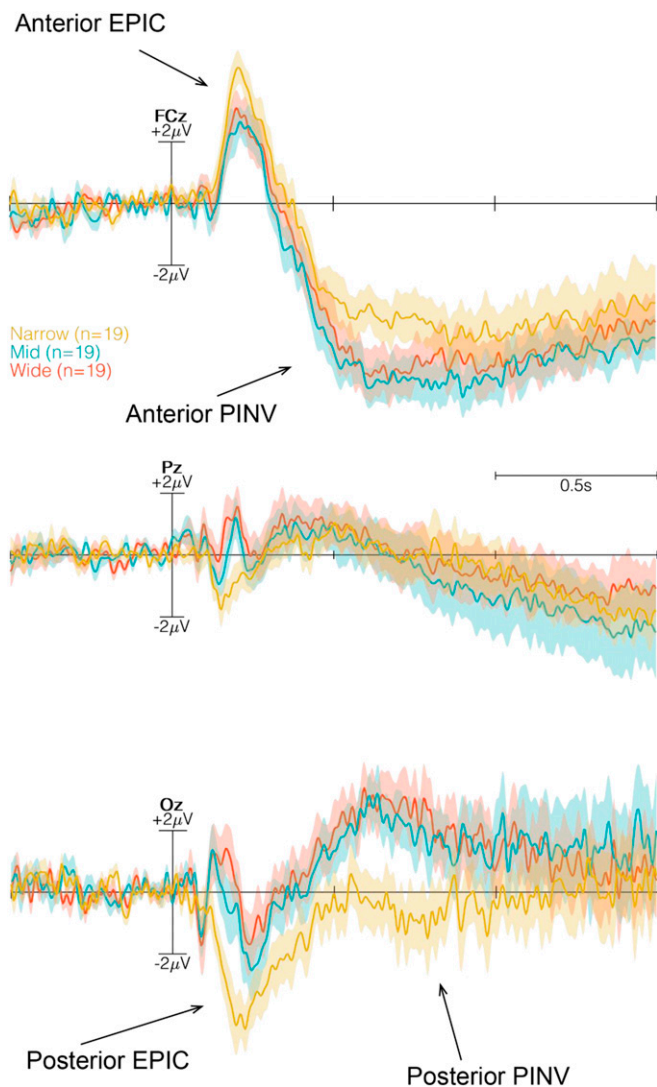


Fig. 7. Three time-locked ERPs (*FCz*, *Pz*, and *Oz*) at the onset of *Go/NoGo*. The *Narrow* condition is in yellow, the *Mid* condition is in blue, and the *Wide* condition is in red. The time window, indicated with dashed lines and a gray transparent box, illustrates the selected time window for analyzing the MRCP by a global $2 \times 3 \times 6$ factorial repeated-measures ANOVA. The anterior and posterior PINV are marked with arrows.

early evoked potentials compared with passable doors, particularly at the frontocentral and occipital sites. Thus, environmental affordances, in terms of being able to program a trajectory to transit spaces, yield a significant measurable effect on early cortical potentials best pronounced over frontal and occipital sites at ~200 ms after the first view of the environment.

Considering the affordance-specific pattern observed for the early P1-N1 complex, previous studies have shown that this visual-evoked potential complex reflects attentional processes associated with spatial or feature-based aspects of stimuli (41–45). Attended stimuli elicit larger P1-N1 amplitudes than unattended stimuli. Based on these findings, our results suggest that passable transitions were associated with increased attentional processing. Keeping this in mind, when viewing the affordance-specific pattern of the P1-N1 complex in light of active inferences (46), the difference confirms the assumption that perceptual processes covary with environmental affordances. In this sense, the amplitude difference might be credited to the process of actively inferring whether the body can move and

transit at all, implying that visual attention is also guided by action-related properties of the environment. Similar to HAC (23) and active inference (22, 47), these findings are in line with parallel cortical processes integrating sensory information to specify currently available affordances.

How one might act on the environment is an ongoing process of resolving affordances, taking place as early as perceptual processes, which situates actions in an intimate position with perception. Such early processes are deeply involved in the conception

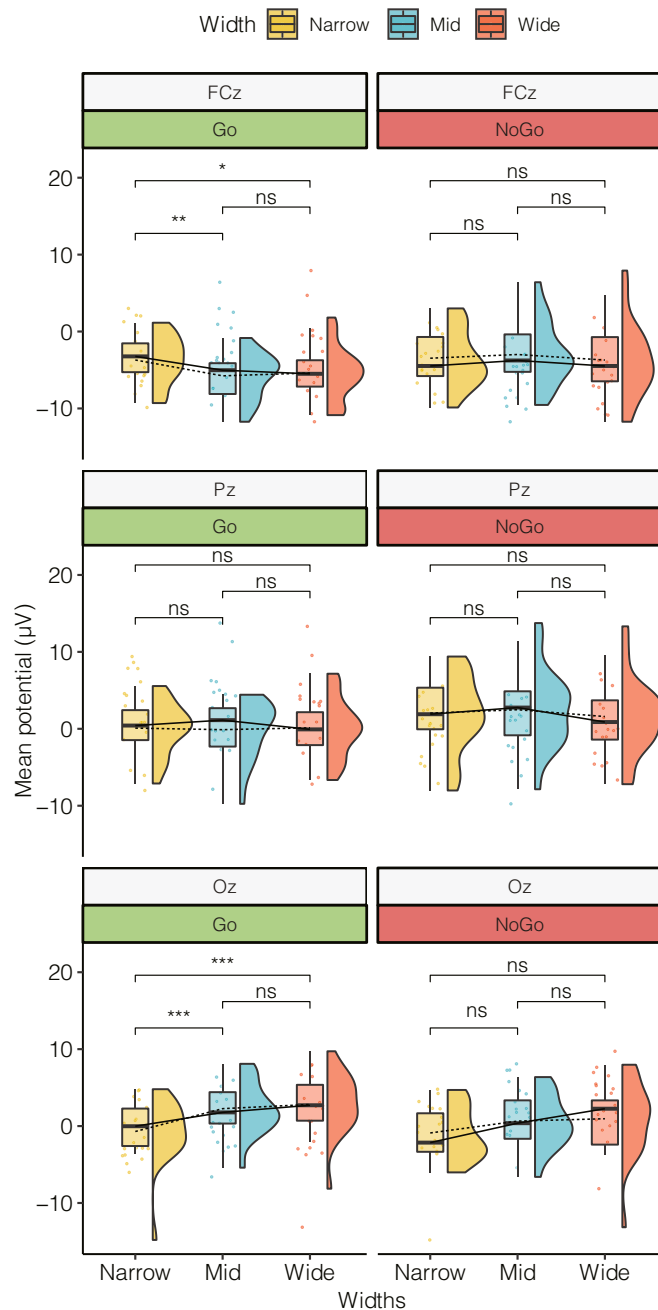


Fig. 8. Raincloud plots of mean amplitude of negative development in the time-locked event of *Go/NoGo* in the time range of 600–800 ms for *FCz*, *Pz*, and *Oz*. Means are indicated by dashed lines; medians, by solid lines. The Tukey HSD contrast revealed differences only in *FCz* and *Oz*, between *Narrow* \times *Mid* for *FCz* ($P = 0.0059$) and *Oz* ($P < 0.0001$) and between *Narrow* \times *Wide* for *FCz* ($P = 0.0323$) and *Oz* ($P < 0.0001$). No differences were observed for *NoGo*. * $P < 0.05$, ** $P < 0.01$, and *** $P < 0.001$; ns, not significant.

and articulation of the environment for an agent, pointing toward the importance of movement in cognition, and of how an agent continuously enacts the world.

Motor-related potentials. Although the ERP plots indicated an affordance trend of the EPIC, statistical tests revealed no significant differences. However, the *Narrow* door width elicited the greatest amplitude in both anterior positivity and posterior negativity. The increased amplitude associated with *Narrow* transitions can be interpreted as a reflection of the body simply not fitting, producing a prediction error because one is forced to interact with the transition. The nature of the PINV component has not been as well investigated as other ERP components, limiting the reliability of an interpretation. Some studies treat this component as modality-unspecific “electrocortical correlate of a cognitive state” (48). Gauthier and Gottesmann (49) hypothesized that the PINV, similar to affordances, acts as a marker of change in the psychophysiological state. Subsequently, the PINV has been used to investigate depression, schizophrenia, learned helplessness, and loss of control (32–34, 50, 51). Depressive and schizophrenic individuals exhibit an increased PINV that is explained as an increased vulnerability for loss of control, as well as increased anticipation for future events (32, 34, 40). It must be emphasized that affordances reflect actions directed toward the future. If an increased PINV reflects increased vulnerability for future events, as we observed for impassable doors, then the component might shed new light on the intentionality in affordances. Given the intention to pass, yet being deprived of doing so, seems to be reflected in the PINV. Casement et al. (34) suggested the PINV depends on lack of control as the state of having no influence, depriving the potential to act. This could explain the difference in the *Narrow* condition, as participants were instructed to attempt to pass at all times until failure, even for impassable openings, leading to a sense of loss of control.

Only in cases of *Go* did we observe a difference in the PINV component, which varied with the environmental affordances. Amplitudes of the component for *Narrow* doors differed significantly different from those for *Mid* and *Wide* doors, while the passable conditions did not differ among the doors. Furthermore, there were no significant differences in the PINV component in cases of *NoGo*, emphasizing the importance of the motor execution itself in evoking the PINV component. These results point toward the PINV component as an expression of the readiness to interact with the designed environment (i.e., less negative for passable doors and more negative for impassable doors), thus serving as a potential marker for the readiness to act given environmental affordances. Our results are also consistent with the observed increase in activity over frontocentral sites reported by Bozzacchi et al. (31), who concluded that the meaning of the action and awareness of being able to act—affordances—affect action preparation, which is here understood as the motor-related potential before movement onset. We argue that the PINV component might reflect a readiness aspect of affordances. This would mean that the PINV is not modulated by the perception that the door is different visual information, but reveals something about the readiness to act. For this reason, we find significant differences in cases of *Go* but not in cases of *NoGo*, and also for passable compared with impassable.

In light of HAC (23), a potential explanation for the absence of differences in the *NoGo* trials is related to the immediate action selection, which in all cases (*Narrow*, *Mid*, and *Wide*) is a simple turn to answer the questionnaire, and thus the task presents the participant with identical affordances. When instead given a *Go*, cortical processes require an action selection related to the anticipated motor trajectory, which differs according to the affordances of the door width. HAC suggests the higher levels bias the lower-level competitions, which operate at the level of action itself, through a cascade of expected next affordances. The lower levels have a continuous competition of how

to satisfy the higher expectations. Action selection, executed while unfolding the planned movements in a continuous manner, depends on the expectation of next affordances.

Of note, regarding architectural experience, because the PINV component was expressed only in the *Go* condition (i.e., forced interaction with the environment), these findings support the importance of movement for architectural experience, in a sense that action or even only the perception of action possibilities alters brain activity. Visually guiding and propelling the body in space greatly influences the continuous emerging of affordances, which in turn affect the human experience. We found differences in frontocentral and occipital areas before movement through space, with the post-imperative negative-going waveform most pronounced over *FCz* indicating involvement of the supplementary motor area (SMA), as reported by Bozzacchi et al. (31). Previous studies showed involvement of the SMA in visually guided actions (52), which is the essence of active inferences. The PINV can be generated independently from the reafferent signal, which is, in terms of active inference, understood as ascending (bottom-up) proprioceptive prediction errors (53). This suggests that the PINV component might reflect descending (top-down) predictions, making the SMA an essential area of the action–perception loop and thus crucial for processing continuous affordances. This might resolve the finding of frontocentral differences in *Go* trials only. The SMA is anatomically bridging the frontal cortex with the motor cortex—perhaps also functionally, as argued by Adams et al. (53), because this anatomical nature fits with the proposed hierarchical characteristics of forward and backward projections in active inferences.

Using VR to investigate cortical processes has its natural limitations, such as the absence of a physical body. Regarding the sense of body, which is at stake in the present study, it has been suggested that VR “may offer new embodied ways for assessing the functioning of the brain by directly targeting the processes behind real-world behaviors” (54), which is remarkably valid for the present study. Riva et al. (54) argued that the brain’s predictive capability immerses the body, and thus related processes, if the visual perception is in line with the body’s actions, for instance, by head movements and wandering. Through a process of trial and error, the brain and body adjust to VR. Furthermore, in terms of architecture, VR as a head-mounted display (55) and as a CAVE system (56) has been integrated into studies with bodily and environmental interests, yielding comparable results. However, VR in combination with neuroscientific methods remains a newer technique and thus must be used with care. It must be emphasized that the purpose of VR in the current experimental setup was to isolate and control the factor of interest. Future studies using MoBI in real-world environments are needed to investigate whether the results from VR can be generalized to the real world.

Conclusion

The present study provides strong evidence for affordances to be processed as early as perceptual processes, linking action and perception in a similar manner to active inference. The results point toward a conception of the brain that seems to deal with “how can I act” while in parallel processes referring to “what do I perceive” take place. These results thus support the assumption that perception of the environment is influenced by affordances and action itself, and thus affordances and action can influence the experience of an environment. Given the importance of affordances and action for brain dynamics, this further emphasizes and qualifies the general idea of enactivism as a holistic approach to investigating cognition. We do not claim that architectural affordances are directly represented as a specific ERP component; however, we provide evidence for an action–perception account of

cognition, which systematically differentiates according to the definition of affordances.

As a note for architects, the fact that we are mobile and predictive beings suggests that architects should take the temporal aspect as seriously as the spatial aspect, given that the predictive process of unfolding bodily movement can alter the perception of space. Moving and transitioning in space is to continuously construct a prediction of a world that we perceive as dependent on our action potentials, which informs brain, body, and mind. Al-

tering perception would ultimately lead spaces to have a potential physiological impact on users. Much remains to be uncovered in architectural cognition.

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