

The psychophysics of compositionality: Relational scene perception occurs in a canonical order

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Abstract

We see not only objects and their features (e.g., glass vases or wooden tables) but also relations between them (e.g., a vase on a table). An emerging view accounts for such relational representations by positing that visual perception is compositional: Much like language, where words combine to form phrases and sentences, many visual representations contain discrete constituents that combine in systematic ways. This perspective raises a fundamental question: What principles guide the compositional process for relational representations, and how are these representations built over time? Here, we tested the hypothesis that the mind constructs relational representations in a canonical order. Inspired by a distinction from the cognitive linguistics tradition, we predicted that ‘reference’ objects (large, stable, or physically controlling objects; e.g., tables) take precedence over ‘figure’ objects (e.g., vases) during scene composition. In Experiment 1, participants who were instructed to arrange items to match linguistic descriptions (e.g., “The vase is on the table”, “The table is supporting the vase”) consistently placed reference objects first (e.g., table, then vase). Experiments 2–5 extended these findings to visual recognition itself: Participants were faster to verify a description when the reference object appeared before the figure object in the scene, rather than vice versa. This Reference-first advantage emerged rapidly (within 100 ms), persisted in a purely visual task, and could not be explained by differences in object size or shape. Together, our findings reveal psychophysical principles underlying compositionality for object relations in visual processing: the mind builds relational representations sequentially, respecting each element’s role.

Keywords: relations; scene perception; intuitive physics; anchor objects; language of thought

1 INTRODUCTION

Look at the image in Figure 1A. What do you see? Certainly you see colors, textures, edges, and countless other visual features—the deep green of a plant, the glossiness of a vase, the grain of a wooden table, and the wicker of a square basket. However, beyond these properties, you may also appreciate something about how the objects *relate* to one another: The plant is sitting *on* the table and the basket is resting *below* it (Fig. 1B). Relational representations are a core topic of study in many domains of higher-level cognition, such as analogical reasoning (Gattis, 2004; Goldwater & Gentner, 2015; Jamrozik & Gentner, 2015; Webb, Fu, Bihl, Holyoak, & Lu, 2023), linguistic reference (Johannes, Wilson, & Landau, 2016; Landau & Jackendoff, 1993; Levinson, 2003; Talmy, 1983; Webb, Holyoak, & Lu, 2023), and causal ascription (Gerstenberg, Peterson, Goodman, Lagnado, & Tenenbaum, 2017; Kominsky et al., 2017; Wolff & Song, 2003).

However, a growing body of empirical evidence suggests that more basic processes of visual perception also encode sophisticated relations such as those depicted in Figure 1B (Hafri, Bonner, Landau, & Firestone, 2024; Hafri & Firestone, 2021; Hafri, Trueswell, &

16 Strickland, 2018; Lovett & Franconeri, 2017). Moreover, this evidence suggests that these
 17 relations are not merely detected by visual processing but also represented by the visual
 18 system in ways that are structured and systematic, as though such representations have
 19 parts that combine into wholes; in other words, they are represented *compositionally* (Hafri,
 20 Green, & Firestone, 2023). On this scheme, for example, when seeing a plant on a table, the
 21 mind represents the scene not as an undifferentiated collection of pixels or textures, but
 22 rather in terms of the discrete constituents *plant*, *table*, and *ON*.

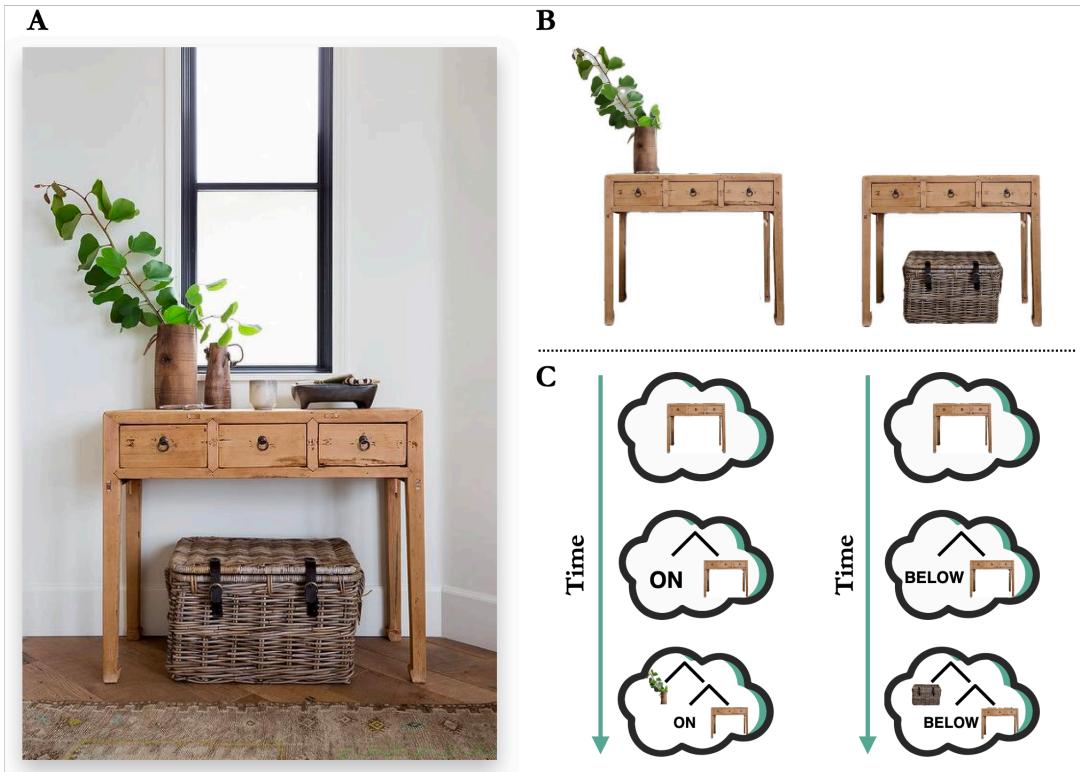


Figure 1: The world contains not only objects and features, but also relations holding between them. (A) We see the plant and its greenness, the table and its size, and so on. But we also appreciate that the plant is sitting *on* the table, and the basket is resting *below* it. (B) A growing literature suggests that visual perception rapidly and spontaneously forms representations of such relations. How does it do so? (C) Despite the fact that the objects are presented to the eyes simultaneously, here we explore the possibility that the mind adopts a sequential order to combine discrete elements into relational representations, with reference objects (e.g., the table) serving as the primary element or ‘scaffold.’

23 This perspective raises a question: How does the mind join these parts together in
 24 forming sophisticated relational representations? Here, we explore the nature of this visual
 25 compositional process, using as a case study the sort of relational representations depicted
 26 in Figure 1A.

27 1.1 Compositionality in visual perception

28 Compositionality refers to a form of representation in which complex representations are
 29 systematically constructed by combining their constituent parts (Fodor & Pylyshyn, 1988).

30 In this sense, it is not particularly controversial to suggest that compositionality can be ob-
31 served in some forms of basic visual representation. For example, recent theoretical work
32 has examined how principles of compositionality apply to visual representations of the
33 bounding contours of objects, suggesting that perceiving contours depends on different
34 modes of composition, such as combining features to fragments and fragments to contours
35 (Lande, 2023). Compositionality is also a fundamental principle in influential theories of
36 object recognition. For example, the Recognition-by-Components theory proposes that ob-
37 jects are represented by freely combining a set of basic components ('geons') (Biederman,
38 1987) into more sophisticated structures; and more recent accounts describe shapes via
39 their parts' intrinsic axes and connections (their 'skeletons'), often in a hierarchical tree
40 format (Feldman & Singh, 2006). Ample empirical evidence supports the psychological
41 reality of such compositional representations (Ayzenberg & Lourenco, 2022; Bonnen, Wag-
42 her, & Yamins, 2023; Firestone & Scholl, 2014; Lewis & Frank, 2016; Lowet, Firestone, &
43 Scholl, 2018; Sun & Firestone, 2021; Van Tonder, Lyons, & Ejima, 2002; Wilder, Feldman, &
44 Singh, 2011).

45 While at first this focus on composition might seem to apply only *within* objects, more
46 recent theoretical work has made the case that this form of representation extends beyond
47 single objects and their contours, to representations that hold *between* objects. Cavanagh
48 (2021) recently proposed a 'language of vision,' whereby visual processing separates im-
49 ages into language-like components, including components such as 'visual nouns' (ob-
50 jects), 'visual verbs' (actions), and 'visual prepositions' (spatial relations), and structurally
51 combines them into 'sentences' (descriptions). In this framework, such 'visual sentences'
52 embody the principle of compositionality in vision, since they capture the relations that
53 combines discrete components into a holistic representation. However, while we can ex-
54 tract information about such entities *based* on visual observation, it is an open question
55 how the mind does so. Are such representations constructed via deliberative reasoning
56 processes (after more basic elements are extracted), or automatically extracted and repre-
57 sented in visual processing itself?

58 Recent work lends support the latter possibility. First, there is evidence that relations
59 are represented abstractly and categorically in visual perception. Observers performing
60 rapid target-recognition tasks 'confuse' scenes depicting the same relation with one an-
61 other, even when the participating objects share few visual features in common, e.g., a
62 phone in a basket being mistaken for a knife in a cup (Hafri et al., 2024; Vettori, Hochmann,
63 & Papeo, 2024). Confusion errors of this sort occur because representing the relation ab-
64 stracted away from its constituent objects makes distinct scenes appear similar. Further-
65 more, even spatial relations that appear to vary from instance to instance are represented
66 in an 'all-or-none,' categorical fashion. When a small ring approaches from above and
67 passes a big ring, the spatial relation between the two is perceived in terms of discrete cat-
68 egories such as *above*, *touching*, *overlapping*, and *containing*. The visual system is especially
69 sensitive to metric changes that cross category boundaries (e.g., from overlapping to con-
70 taining) than those that do not (e.g., from less overlapping to more overlapping) (Lovett &

71 Franconeri, 2017).

72 Second, there is also evidence that visual processing binds representations of arbitrary
73 entities to distinct roles in a relational *structure*. For example, in a relational scene where
74 a girl is pushing a boy, the girl fills the role of *Agent* (i.e., the initiator of the action) and
75 the boy the role of *Patient* (the recipient of the action). Recent work has shown that this
76 binding happens rapidly and spontaneously: When observers repeatedly reported the lo-
77 cation of a target individual (e.g., boy) in a stream of action photographs (e.g., girl-kicking-
78 boy, boy-pushing-girl), they were slower when the target individual's role (Agent/Patient)
79 switched (e.g., pusher on trial $n + 1$ but kickee on trial n) (Hafri et al., 2018). Thus, visual
80 processing is sensitive to this structure, such that changes to this structure produce re-
81 sponse costs even when observers are engaged in orthogonal tasks (Hafri, Papafragou, &
82 Trueswell, 2013; Hafri et al., 2018; Vettori, Odin, Hochmann, & Papeo, 2023).

83 The above work suggests that visual processing represents relations in ways that pre-
84 serve the identities of both the entities themselves and the relations in which they par-
85 ticipate. This representational scheme is often called *role-filler independence* (Quilty-Dunn,
86 Porot, & Mandelbaum, 2023), and it makes such representations compositional: Just as
87 individual words compose together to form phrases and sentences in language, relational
88 representations contain discrete constituents that combine in systematic ways (Hafri, Green,
89 & Firestone, 2023). The existence of visual relational representations with this property
90 suggests that some aspects of perception may exhibit core properties of a 'Language of
91 Thought (LoT)' (Fodor, 1975; Quilty-Dunn et al., 2023), a format of representation that can
92 readily accommodate compositionality in ways that other formats that are more tradition-
93 ally associated with visual perception may not (i.e., iconic or 'picture-like' formats; Block,
94 2023; Burge, 2022; Carey, 2009; Kosslyn, Thompson, & Ganis, 2006).

95 1.2 Our question: What constraints govern the process of visual composition?

96 While compositionality is traditionally discussed in terms of representational *format*, the
97 existence of LoT-like representations in visual perception raises an intriguing question
98 about the compositional *process* itself: How are such representations composed by the
99 mind from their constituent parts? In other domains such as speech processing, struc-
100 tured representations are constructed incrementally, as the speech signal is dynamic and
101 temporally extended, unfolding over time (Christiansen & Chater, 2016). By contrast, in
102 visual processing, relational content (objects and their visual features) is in principle im-
103 mediately available from an image. Despite this, might visual relational representations
104 also be 'built' sequentially by the mind (Figure 1C)?

105 Classic research in visual cognition offers clues about the dynamic nature of relational
106 processing in vision. Ullman (1987) proposed the concept of *visual routines*—sequences
107 of spatial operations executed to extract simple relations, such as whether one object is
108 INSIDE, ON, or COLLINEAR with another. One well-known example of such visual rou-
109 tines is *curve-tracing*, in which the visual system systematically follows a curve's contour
110 in order to judge whether two points lie on the same or different curve (Jolicoeur, Ullman,

& Mackay, 1986, 1991). These routines occur dynamically in both space and time, and are generally considered to require effort and intentional initiation (though see Wong & Scholl, 2024). However, this work primarily examines simple geometric features such as points, lines, and curves. It remains unclear how—or whether—such routines extend to more sophisticated relations between real-world objects (e.g., those shown in Figure 1A). These relations are fundamentally different: rather than being defined purely by spatial properties like continuity or particular distances, they often involve abstract roles or physical forces (e.g., Support, Containment). As such, they cannot be traced or followed in any literal sense. Instead, visual processing for such relations may rely on a different kind of routine—one that sequentially combines objects into a structured mental representation. If so, what principles govern this compositional process?

Insights from event cognition suggest that the *roles* of participants such as Agent and Patient are crucial for determining how relational scenes are processed in time. A range of ‘Agent advantages’ have been reported in the literature: relative to Patients, Agents are prioritized in visual search, recognition, and attention (Segalowitz, 1982); allow for better predictions about upcoming events (Cohn & Paczynski, 2013); facilitate the processing of actions (Cohn & Paczynski, 2013); and elicit stronger neural responses (Cohn, Paczynski, & Kutas, 2017). The primary function of Agent is even more evident in psycholinguistic research: When the Agent comes before the Patient in an English sentence (i.e., Agent as grammatical subject), readers find it easier to grasp the sentence’s meaning. For example, a sentence like *the dog bit the man* is processed more readily than its passive counterpart, *the man was bitten by the dog* (Ferreira, 2003). On one hand, this Agent advantage seems intuitive: as the initiator of an action, the Agent often carries more information and ranks early as the subject in the sentence. Thus, it is possible that such advantages would only apply to Agent-Patient relationships and not extend beyond them, to spatial or physical relations involving inanimate entities. On the other hand, the ease of processing Agent-first order in event relations might reflect a more general mechanism in relational processing that applies to relations of all types. How, then, are other types of visual relations represented?

Here, we extend research on Agent-Patient events into more fundamental forms of relational processing, by testing the hypothesis that the mind builds spatial and physical relations sequentially, according to the roles of the participating objects. The objects in a relational scene can assume different roles (often called ‘thematic roles’ in linguistics). In spatial or physical relations such as ON or BELOW, these roles are known as ‘Reference’ (sometimes called ‘Ground’) and ‘Figure’. Reference objects are generally those that are large, stable, and/or physically ‘control’ other objects, while figure objects are those that are small and/or mobile (Gleitman, Gleitman, Miller, & Ostrin, 1996; Landau & Jackendoff, 1993; Talmy, 1975).

Of note, there is a systematic relationship between the non-linguistic construal of such entities and the structural positions in which they are encoded in linguistic utterances, with the reference object placed farther down in the syntactic structure of the utterance than the figure (Landau & Gleitman, 2015). For example, in the sentence ‘The bike is to

the left of the garage', the grammatical subject (bike) is figure, and the grammatical object (garage) is reference. Interestingly, despite the fact that figure objects are often the subject of a sentence and thus (in English) appear first in sequence, the cognitive representation of such relations appears to be such that the reference object is primary. According to a tradition known as cognitive linguistics [Miller and Johnson-Laird (1976); Talmy (1975)], the reference object defines the spatial or physical reference frame relative to which the figure object is located (see also [Gleitman et al. (1996); Landau & Jackendoff (1993)]). Indeed, this fundamental asymmetry is precisely why the term 'Reference' is used in the first place. The asymmetric placement of entities in a linguistic utterance can even influence reference and figure assignment. Consider an utterance such as "The garage is next to the bike"; here, the garage and bike are in syntactic roles usually associated with figure and reference objects, respectively. This shift in structure either makes the sentence sound unnatural (or even imbues these entities with the properties corresponding to the roles associated with these syntactic positions; as if, e.g., there a small garage on wheels moving around a giant, stationary bike statue; [Gleitman et al. (1996)]). This suggests that syntactic structure not only reflects, but can actively shape, relational interpretation in language. (As we show later, this linguistic influence may play a role in scene composition when visual properties alone do not provide sufficient cues.)

Vision research also hints at the primary role of reference objects in constructing relational scenes. For example, studies on complex scene perception suggest that visual processing takes advantage of the unique role of 'anchor' objects to guide search and recognition ([Võ, Boettcher, & Draschkow, 2019]). In cluttered scenes (e.g., a classroom), observers were slower to find the target object (eraser) if the anchor object (chalkboard) was swapped with an irrelevant object (map) ([Boettcher, Draschkow, Dienhart, & Võ, 2018]). Studies with simpler geometric stimuli also suggest that figure and reference objects play different roles in the temporal construction of relations. Indeed, prior work has found that the perception of certain spatial relations requires serial processing ([Holcombe, Linares, & Vaziri-Pashkam, 2011]). Evidence from attentional cuing also indicates that previewing the location of one object in a spatial relation can influence processing speed. [Roth and Franconeri (2012)] showed that observers responded faster to questions such as "Is the red disc above the blue disc?" when they previewed the linguistic subject (red disc) before seeing the full relation between two objects ([Roth & Franconeri, 2012]).

Developmental research also supports a structured composition process in representing certain spatial and physical relations. When children acted out a relational scene with real-world objects according to a statement (e.g., "The green block is on the top of the pink block."), they made fewer errors when they placed the figure object (green block) relative to a fixed reference object (pink block) than vice versa ([Huttenlocher & Straus, 1968a]). In similar work using both active and passive statements (e.g., "The red truck is pushing the blue truck"; "The blue truck is pulled by the red truck", etc.), children reconstructed relational scenes with real-world objects more quickly when placing the Agent (the red truck) with respect to the Patient, regardless of which entity served as the grammatical subject

193 ([Huttenlocher, Eisenberg, & Strauss, 1968b](#)). Although the above work is suggestive, the
194 hypothesis that relational perception follows a canonical order has not yet been tested in
195 psychophysical experiments. It remains unclear whether the visual system builds relations
196 in an order that respects the objects' roles.

197 **1.3 The present experiments: How to build a scene**

198 Our work takes inspiration from the cognitive linguistics literature in hypothesizing that
199 reference objects (e.g., tables, shelves, etc.) rather than figure objects (e.g., vases, laptops,
200 etc.) serve as the scaffold for relational representations in visual perception ([Gleitman](#)
201 et al., [1996](#); [Landau & Jackendoff, 1993](#); [Talmy, 1975](#)). To test this hypothesis, we asked
202 whether participants would construct relational scenes following a 'Reference-first' or-
203 der, and whether participants' visual processing of relational scenes would be facilitated
204 when they have visual access to the reference object before the figure object (as opposed
205 to vice versa). We created relational scenes from various household objects (e.g., laptop,
206 desk, lamp, nightstand), encompassing physical relations (e.g., desk supporting laptop)
207 and spatial relations (e.g., laptop below desk). In Experiment 1, participants read linguis-
208 tic descriptions and manually arranged objects on-screen to match the described relation,
209 allowing us to replicate previous findings of a Reference-first advantage in manual con-
210 struction tasks ([Huttenlocher et al., 1968b](#); [Huttenlocher & Straus, 1968a](#)). Experiments
211 2–5 used a recognition paradigm in which participants matched visual scenes to linguis-
212 tic or pictorial probes viewed just beforehand. This paradigm builds on the classic sen-
213 tence–picture verification task, which has traditionally been used to study how people
214 determine whether a linguistic statement accurately describes a picture ([Clark & Chase,](#)
215 [1972](#), [1974](#)). In the standard version of the task, participants read a sentence and view a
216 scene, and then verify whether the sentence correctly describes the image. We introduced
217 a crucial change: instead of presenting the full relational scene at once, we presented the
218 objects asynchronously, either with the reference or figure object appearing first.

219 To foreshadow the key results, we found a Reference-object advantage across all five
220 experiments: Participants employed a reference-first order in composing relational scenes,
221 and they were faster to recognize a visual scene when the reference object appeared before
222 the figure object rather than vice-versa. We further found that when visual and physical
223 differences between objects were eliminated, the linguistic structure of the probe sentence
224 influenced the order of composition. Taken together, these results suggest that even though
225 visual scene information is in principle available to the observer all at once, the mind com-
226 poses relational representations sequentially, in ways that respect the role of each element
227 in the relation.

228 **2 EXPERIMENT 1 – Manual Construction of Relational Scenes**

229 How do people construct relational representations? Experiment 1 took a literal approach
230 to this question by asking whether participants have a preference for the order in which

they place constituent objects to compose relational scenes. Our approach was inspired by early studies of sentence comprehension conducted by Huttenlocher and colleagues, in which children placed one object with respect to a fixed object according to a statement (Huttenlocher et al., 1968b; Huttenlocher & Straus, 1968a). Children had an easier time doing so when the reference object was fixed in place first. The current study aimed to extend these previous findings by employing a similar paradigm, but with key differences: While previous studies fixed an object for children and asked them to place another object relative to it, our task allowed adult participants to place objects in whatever order they wanted. On each trial, participants were asked to compose a scene to match a pre-specified linguistic description by dragging objects into a framed workspace. While they could do so in whatever order they liked, we predicted that they would move the reference object first, even without external pressure to do so.

2.1 Method

2.1.1 Open Science Practices

An archive of the data, code, stimuli, experiment preregistrations, and other relevant materials is available at:

https://osf.io/vzxdq/?view_only=5b194f89f42542429f6c69b64b2e3630.

For each experiment, we preregistered the sample size, experimental design, and analyses (including exclusion criteria and some secondary analyses). Demos of the experiments can be viewed at <https://palresearch.org/buildingrelations>, so readers can experience them as participants did.

2.1.2 Participants

We recruited 40 participants for this experiment from the online platform Prolific (<https://www.prolific.co/>). (For a discussion of this participant pool's reliability, see Peer, Brandimarte, Samat, & Acquisti, 2017) This sample size was determined based on a smaller pilot study. Participants were prescreened for a minimum approval rate of 85%, at least 50 prior submissions, normal or corrected-to-normal vision, native English proficiency, and U.S. nationality. Sample sizes were preregistered for this and all other experiments. All studies were approved by the Johns Hopkins University Institutional Review Board.

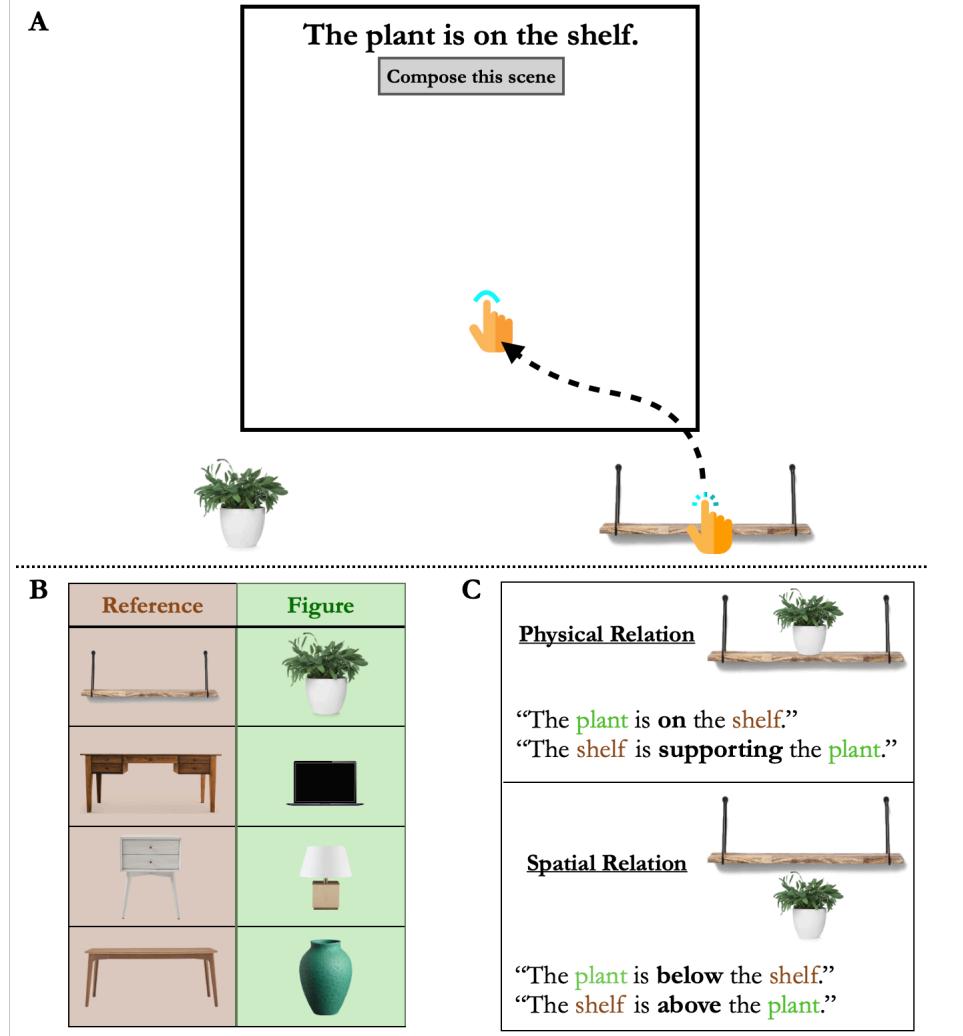


Figure 2: Illustration of the manual-composition task used in Experiment 1. (A) On each trial, participants were asked to compose a scene by moving two objects into the workspace, according to a sentence that described a relational scene. (B) The objects used to compose the scenes were four pairs of objects, with one reference object and one figure object in each pair. (C) The relational scenes were divided into two categories: physical relations and spatial relations. Each scene was described in two ways: either the figure object or the reference object was the grammatical subject of the descriptive sentence, and thus was the first mentioned entity in the sentence.

2.1.3 Stimuli

Eight colored images were used in the experiment, grouped into four pairs: vase/table, laptop/desk, lamp/nightstand, and plant/shelf (Figure 2B). Each pair consisted of a reference object that was relatively large and stable (e.g., table, desk, nightstand, and shelf) and a figure object that was relatively small and mobile (e.g., vase, laptop, lamp, and plant).

Participants were provided with sentences that described a relation between the two objects in each pair. These sentences varied along two dimensions: (1) whether the depicted relation was physical or (merely) spatial (e.g., "the plant is on the shelf" describes a physical relation, and "the plant is below the shelf" describes a spatial relation); and (2) whether the reference object or the figure object was the grammatical subject of the sen-

271 tence or not (e.g., in “the plant is on the shelf,” the figure object, ‘plant,’ is the subject, while
272 in “the shelf is supporting the plant,” the reference object, ‘shelf,’ is the subject). See Figure
273 2C for a complete list of sentences for one of the object pairs.

274 Images ranged in size from 120×68 pixels to 417×186 pixels and were presented in the
275 participant’s Web browser. The workspace was presented at 600×600 pixels, with a white
276 background. Because of the nature of online studies, we could not know the exact viewing
277 distance, screen size, and luminance (etc.) of these stimuli as they appeared to participants.
278 However, any distortions introduced by a given participant’s viewing distance or monitor
279 settings would have been equated across all stimuli and conditions.

280 2.1.4 Procedure

281 The experimental task is depicted in Figure 2A. On each trial, participants first read a state-
282 ment describing a relation between two objects, and then they clicked a button to indicate
283 they were ready to compose the scene. Immediately after the click, the two mentioned
284 objects appeared beneath the workspace (one on the right and the other on the left). Par-
285 ticipants dragged each object into the workspace in whatever way they chose to compose a
286 scene that correctly reflected the statement. Once both objects were inside the workspace,
287 participants were able to click on a button to proceed to the next trial.

288 Overall, the experiment consisted of 32 trials. The four object pairs were combined
289 with the four types of sentences, which made 16 unique trials. Each combination of object
290 pair and sentence appeared twice: once with the figure object on the left side (below the
291 workspace), once with the figure object on the right side. Trial order was randomized
292 across participants.

293 2.1.5 Exclusions

294 As specified in the preregistration, we planned to exclude trials where the composed visual
295 scenes did not accurately match the linguistic descriptions. To do so, we preregistered
296 scene-specific boundaries within which each object should be placed in the particular scene
297 in order to be considered accurate (with these boundaries detailed in the preregistration).
298 We also planned to exclude any participant who had low overall accuracy (< 90%), lacked
299 at least one trial in each combination of the key factors (Relation Type, Sentence Structure,
300 and Object Side), or failed to provide a complete dataset.

301 2.2 Results

302 One participant was excluded for failing to submit a complete dataset. As expected, sub-
303 jects had little difficulty completing the task, with a mean accuracy of 98.7% in composing
304 the visual scenes to match the linguistic descriptions given.

305 Crucially, participants overwhelmingly placed the reference object first (e.g., shelf be-
306 fore plant), doing so on 96.7% of trials across participants, $t(38) = 48.13, p < 2.2 \times 10^{-16}$.
307 This pattern held regardless of sentence order (i.e., whether the reference object was the

grammatical subject or object of the sentence), object identity (all four pairs of images), and relation type (physical or spatial). Moreover, this Reference-first preference emerged at the very beginning of the task: Even in the very first trial of the experiment, a majority of participants (34 out of 39) moved the reference object before moving the figure object (binomial test, $p = 2.43 \times 10^{-16}$). These effects held even when including all trials (including those coded as incorrect, $t(38) = 48.66$, $p < 2.2 \times 10^{-16}$), and they generalized across the different object pairs (t -test across object-pair means, $ts(38) \geq 35.90$, $ps < 2.2 \times 10^{-16}$). We also conducted a repeated-measures ANOVA across subject means of Reference-first proportion, with the factors of interest as Relation, Sentence Structure, and Object Side, and found that these factors did not significantly modulate the Reference-first effect ($Fs(1, 38) \leq 3.65$, $ps \geq 0.064$)).

These results provide initial evidence that the mind applies a canonical routine in constructing relational scenes. They also raise the possibility that the mind adopts this Reference-first routine in representing visual relations more broadly, including in visual recognition. We explore this possibility in the remaining experiments of this paper.

3 EXPERIMENT 2 – Visual Recognition of Relational Scenes

Experiment 1 revealed a predominant order in relational scene ‘production’: participants preferred to move the reference object into the scene first and position the figure object relative to it. Does this pattern extend beyond mere preferences and drive visual processing itself? Experiment 2 asked whether relational representations are built according to a similar compositional routine in visual .

To do so, we used a variant of the ‘sentence-picture verification task’ initially developed by [Clark and Chase] (1972, 1974) to investigate how people decide whether or not a linguistic statement accurately describes a picture. In such a task, participants read a sentence (e.g., “The star is above the plus sign”) and are asked to verify whether it is a correct description of an image that appears soon after. We made one crucial change to the standard paradigm: Instead of showing the image all at once, sometimes the reference object (e.g., the table) appeared a half-second before the figure object (e.g., the laptop), or vice versa.

Manipulating this display order allowed us to ask whether there is a privileged order for visual recognition. We reasoned that if the visual system builds relational representations sequentially, following the Reference-first order pattern observed in Experiment 1, then participants would be faster to verify the target sentence when they saw the reference object right before the figure object, because this presentation order would ‘match’ the order in which the mind constructs such representations.

343 **3.1 Method**

344 **3.1.1 Participants**

345 As per our preregistration, 40 participants were recruited through Prolific. This sample
346 size was determined by a power analysis on a small pilot study that produced similar
347 results and indicated a 99% probability of detecting the effect of interest. Participants in
348 this and subsequent experiments had to pass the same pre-screening criteria as in Exper-
349 iment 1, and they could participate only if they had not previously completed a related
350 experiment in this series.

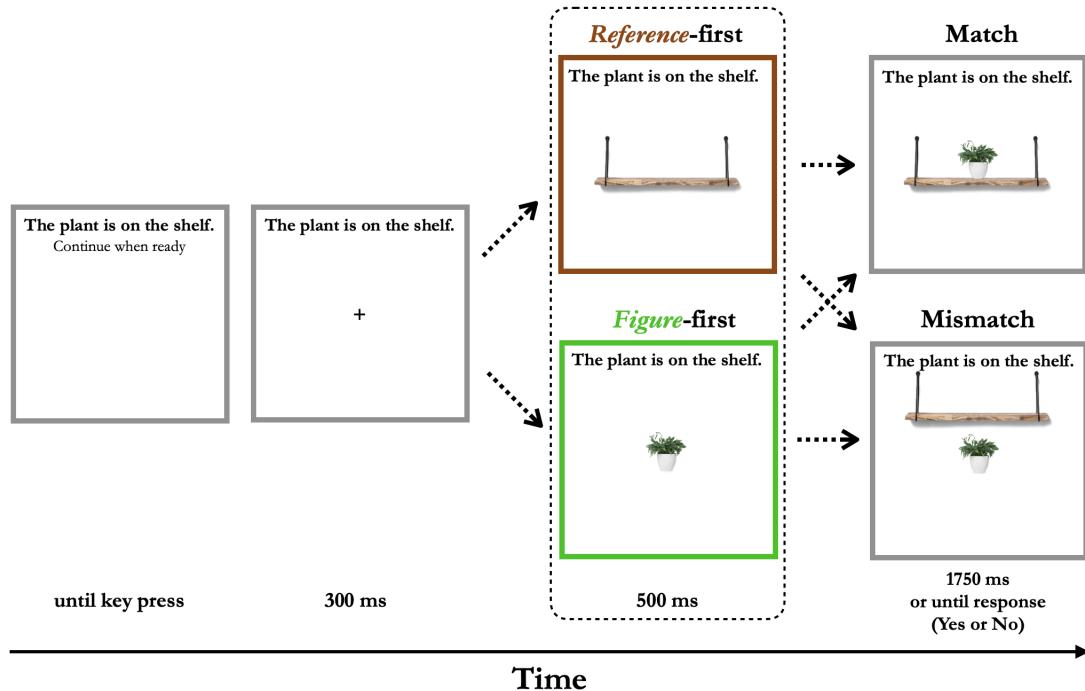


Figure 3: Illustration of the visual recognition task. At the beginning of each trial, participants were given a sentence describing a relational scene. Once they pressed a key to proceed, a fixation cross appeared in the center of the frame for 300 ms, and then either the reference or figure object appeared at the center. After 500 ms, this was followed by the other object, which completed the scene. (On some trials, the objects appeared simultaneously.) Participants indicated whether the resultant scene matched or mismatched the sentence description as rapidly and accurately as possible, before the trial timed out. (For simplicity, the 500-ms blank display before fixation is omitted in this figure.)

351 **3.1.2 Design and Procedure**

352 Participants were instructed that on each trial, they would read a sentence and then have
353 to verify whether a subsequently presented visual scene matched the sentence. Figure 3
354 illustrates the trial sequence. All sentences and scenes were presented on a white back-
355 ground within a 600 × 600 pixel frame. At the beginning of each trial, participants read a
356 sentence that described a relational scene (e.g., "The laptop is on the table"), presented at
357 the top of the frame. After reading, they pressed the space bar to indicate that they were
358 ready to continue (but had to remain on the sentence screen for at least 500 ms). Once

359 participants pressed the key, a blank screen appeared for 500 ms, followed by a fixation
360 cross at the center of the scene for 300 ms. Then the to-be-verified scene was presented.
361 The sentence remained at the top of the image frame throughout the trial.

362 Each visual scene was presented in one of three object-order conditions: (1) Reference-
363 first, in which the reference object was presented first and then the figure object appeared
364 500 ms later; (2) Figure-first, in which the figure object was presented first, followed by
365 the reference object 500 ms later; and (3) Simultaneous, in which both reference and figure
366 objects appeared at the same time, right after the fixation cross disappeared. After the full
367 scene was displayed, participants judged whether the complete scene matched the pre-
368 specified sentential description as fast as possible without sacrificing accuracy, by pressing
369 either Y for a match or N for a mismatch. Trials timed out if no response was given within
370 1750 ms. Of note, in the two 500-ms delay conditions (Reference-first and Figure-first), the
371 first presented object always appeared at central fixation, so its location was not predictive
372 of the correct response. (In the Simultaneous condition, the center object could be either
373 Figure or Reference, as described below.)

374 All four pairs of objects and 16 sentences used in Experiment 1 (Figure 2) were also
375 used in this experiment. Several factors were fully crossed within participants: (a) Relation
376 Type (spatial or physical), (b) Second Image Delay (0 ms or 500 ms), (c) Sentence Structure
377 (Figure-as-subject or Reference-as-subject), (d) Trial Type (match or mismatch), and (e)
378 Center Object Type (either the reference or figure object), yielding 128 test trials ($2 \times 2 \times$
379 $2 \times 2 \times 2 \times 4$ object pairs). Object images were the same size as in Experiment 1.

380 The key factor of interest was the presentation order of objects, determined by the
381 combination of the factors Second Image Delay and Center Object. In particular, at the
382 0-ms second image delay (i.e., the Simultaneous condition), the object presented centrally
383 in the visual scene was either Reference or Figure; while at the 500-ms delay, the object
384 presented first (Reference or Figure) would always appear at the center. Six practice trials
385 preceded the test trials and contained objects not used in the main study (a book and
386 bookshelf), all in the Simultaneous object-order condition. This resulted in 134 trials in
387 total. Test trial order was randomized for each participant.

388 3.1.3 Analysis

389 As stated in our preregistration, the dependent variable was the participant's response
390 time (RT) on each trial, measured from the onset of the second image. (For the Simulta-
391 neous condition, the onset of the second image was also the onset of the full scene.) Only
392 Match trials in which the visual scene matched the sentence description were analyzed.
393 We also excluded trials that timed out before a response was given, error trials, and trials
394 in which RTs were extraordinarily fast (< 200 ms). Additionally, we excluded participants
395 who met any of the following criteria: (a) Low overall accuracy (< 90%), (b) too many time-
396 outs (> 5% of trials), (c) too many extraordinarily fast RTs on test trials (> 5% of RTs < 200
397 ms), (d) after trial exclusion, not having at least one trial in a cell for each combination of
398 the factors of interest (i.e., Relation Type, Object Order, and Sentence Structure), or (e) fail-

399 ing to contribute a complete dataset. However, note that none of the results reported here
 400 or in subsequent experiments were dependent on the particular exclusion criteria used for
 401 RT or accuracy; in other words, the effects were significant in the same direction regardless
 402 of whether the exclusion criteria were applied or not.

403 We conducted a repeated-measures analysis of variance (ANOVA) on participant means
 404 of inverse-transformed response times (-1000/RT, correctly answered Match trials only) to
 405 examine the main effects of Object Order, Relation Type, and Sentence Structure, as well
 406 as their interactions. Our primary question in this study concerned Object Order: In par-
 407 ticular, we expected to observe shorter RTs when the reference object was presented right
 408 before the figure object in the scene, as compared to vice versa. We expected this to hold re-
 409 gardless of relation type (spatial or physical). Crucially, we predicted that this would also
 410 hold regardless of sentence structure (i.e., whether the figure or reference object appeared
 411 as the grammatical subject of the sentence), which would indicate that any object-order
 412 advantage observed was not simply driven by the order in which objects were mentioned
 413 in the sentence.

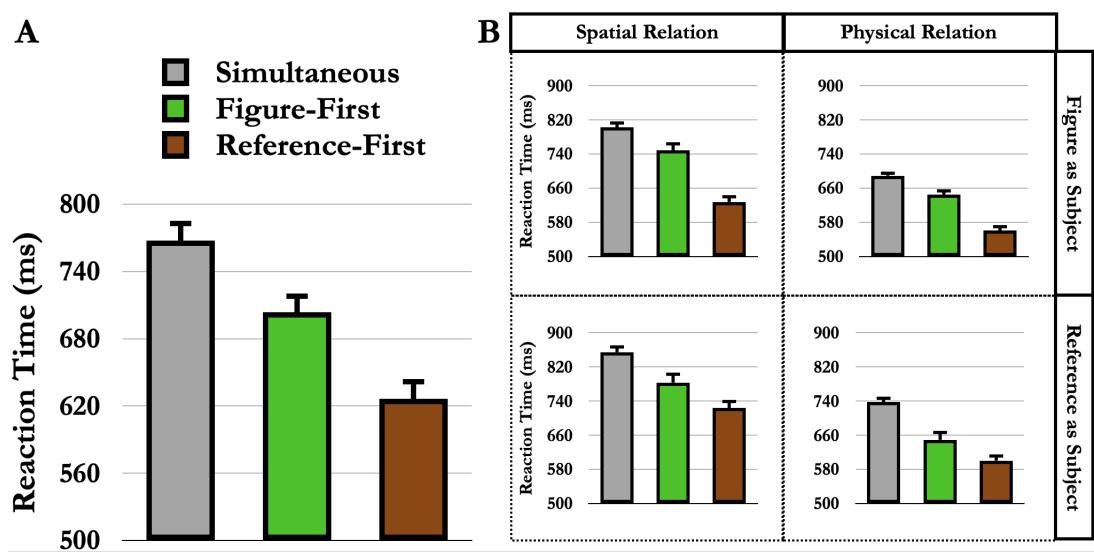


Figure 4: Results of Experiment 2. (A) Participants were faster verifying the relational scene when the reference object appeared first and the figure object second, compared to when the order was reversed or when both objects appeared simultaneously. (B) This Reference-first advantage in relational recognition emerged for both spatial (“above”/“below”) and physical relations (“on”/“support”), regardless of which object was mentioned first in the sentence. Bars reflect mean response times across participants (computed from correct trials only), and error bars reflect within-participant 95% confidence intervals.

414 3.2 Results

415 Three participants were excluded based on preregistered criteria, leaving 37 participants
 416 for further analysis. As expected, participants had little difficulty completing the task, with
 417 a mean accuracy of 97% and a mean response time (across all conditions) of 741 ms.

418 The ANOVA revealed a significant main effect of Object Order, $F(2, 72) = 77.98, p =$
 419 $2 \times 10^{-16}, \eta^2 = 0.68$. As shown in Figure 4A, the mean RT for the Reference-first condition

420 was shorter than Figure-first and Simultaneous conditions, indicating that participants
421 were faster to recognize the visual scene and verify that it matched the sentence when the
422 reference object appeared before the figure object. Subsequent Holm-Bonferroni-corrected
423 paired-samples *t*-tests revealed significant pairwise differences between the three Object-
424 Order conditions (Reference-first vs. Figure-first, Reference-first vs. Simultaneous, and
425 Figure-first vs. Simultaneous, all $ts(36) \geq 5.44$, $ps \leq 0.0001$, $ds \geq 0.25$). We note that
426 the Simultaneous condition was slower than both the other two conditions. While we
427 did not have strong predictions with respect to this condition, one possible explanation
428 is that sequential presentation, even in the Figure-first condition, helps pre-segment the
429 objects, whereas in the Simultaneous condition, participants can only begin to perform
430 this segmentation once all objects have appeared on the display.

431 Figure 4B illustrates the effect of Object Order, split by Relation Type and Sentence
432 Structure. As predicted, the ‘Reference-object advantage’ arose for both physical and spa-
433 tial relational scenes, and it held no matter the order of elements mentioned in the sentence
434 descriptions. Indeed, while there was a significant interaction of Object Order and Sen-
435 tence Structure ($F(2, 72) = 7.0$, $p = 0.0017$, $\eta^2 = 0.16$), Holm-Bonferroni-corrected paired-
436 samples *t*-tests confirmed significant pairwise differences between the three Object-Order
437 conditions at each level of Sentence Structure (all $ts(36) \geq 3.22$, $ps \leq 0.005$, $ds \geq 0.54$).
438 In addition, Relation Type and Sentence Structure both emerged as significant main ef-
439 fects (Relation Type: $F(1, 36) = 129.80$, $p = 1.7 \times 10^{-13}$, $\eta^2 = 0.78$; Sentence Structure:
440 $F(1, 36) = 11.89$, $p = 0.0015$, $\eta^2 = 0.25$). However, there was no significant three-way
441 interaction among the three factors ($F(2, 72) = 1.40$, $p = 0.25$, $\eta^2 = 0.037$).¹

442 We also conducted secondary preregistered analyses to test whether the effect of Ob-
443 ject Order generalized across the different object pairs. A repeated-measures ANOVA for
444 mean RTs across object pairs showed a significant main effect of Object Order, $F(2, 6) =$
445 55.8 , $p = 1.33 \times 10^{-14}$, $\eta^2 = 0.95$. Subsequent Holm-Bonferroni-corrected paired-samples
446 *t*-tests revealed significant pairwise differences of the three Object-Order conditions, all
447 $ts(3) \geq 3.85$, $ps \leq 0.033$, $ds \geq 1.92$.

448 4 EXPERIMENT 3 – Timing of the Compositional Process

449 Experiment 2 revealed a Reference-first advantage in the recognition of relational visual
450 scenes: Participants were faster to match the scene to the pre-specified linguistic descrip-
451 tion when the reference object appeared just a half-second before the figure object, rather
452 than vice versa. This result raises a natural question: How quickly does this effect emerge?
453 Here we probed the timing of the compositional process in detail. We ran the same study
454 as before, but with one change: We systematically varied the onset of the second object in

¹It is worth noting that one prompt version differs slightly from the others: the Reference-first physical-relation sentence probe used constructions such as “The table is supporting the vase,” which employs a verb (“support”) rather than a preposition (as in “on,” “above,” or “below”). However, no appreciable slowdown was observed in this condition compared to the others (see Figure 4B). Moreover, the absence of a significant three-way interaction suggests that the object-order effect for this condition was largely consistent in both direction and magnitude with the other conditions.

455 the scene, from very early (100 ms) to late (1000 ms). This manipulation also allowed us to
 456 distinguish between different possible underlying mental processes driving the Reference-
 457 first advantage. One possibility is that the effect arises from deliberate, cognitive expect-
 458 ations: Participants may form predictions about what will appear next based on the first
 459 object's identity. In that case, the effect might take time to emerge and possibly strengthen
 460 as more time is available for reasoning between presentation of the two objects. Alter-
 461 natively, visual processing itself might enforce a Reference-first order when constructing
 462 relational representations. In this case, we would expect the effect to emerge rapidly, even
 463 with minimal delays between the first and second object. By systematically varying the
 464 presentation timing, we aimed to distinguish between these possibilities.

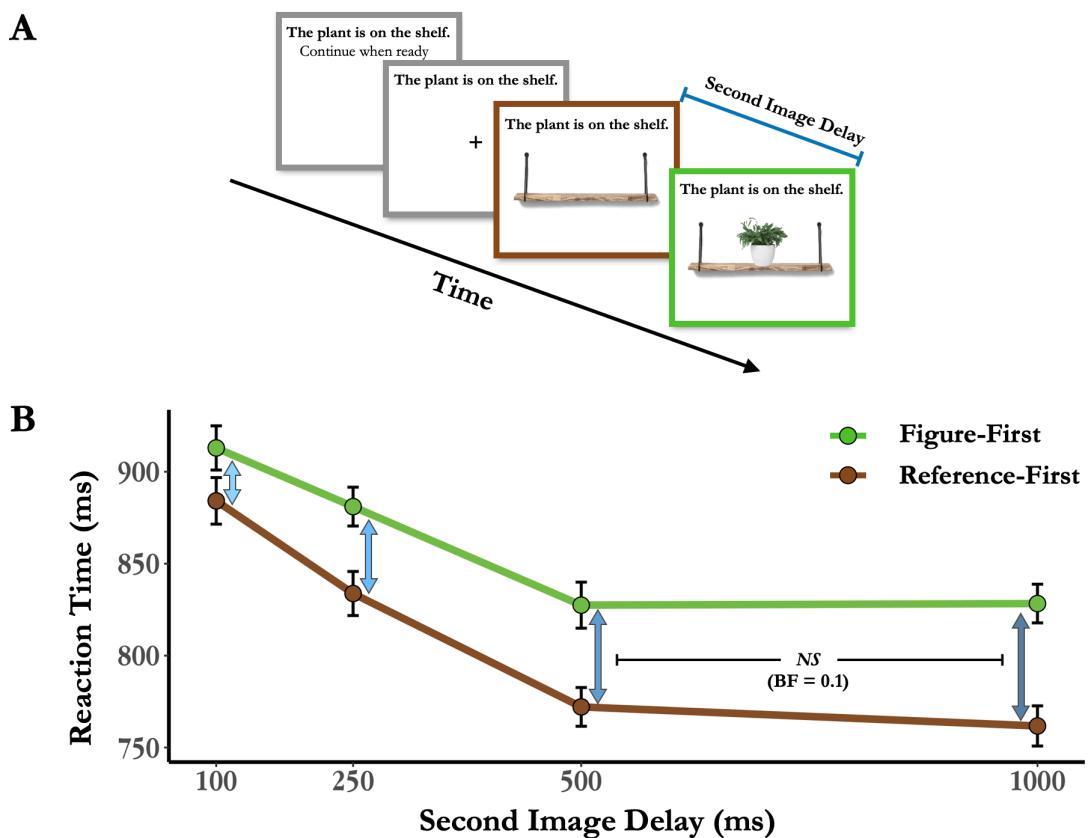


Figure 5: Trial structure and results of Experiment 3. (A) The delay between figure object and reference object presentation was systematically varied from 100 ms to 1000 ms. (For simplicity, blank displays during the trial are omitted from this figure.) (B) The Reference-first advantage (i.e., faster RTs when the reference object appeared first) emerged as early as 100 ms, increased to a peak at 500 ms, and then plateaued. Filled circles reflect mean response times across participants (computed from correct trials only), and error bars reflect within-participant 95% confidence intervals.

465 4.1 Method

466 4.1.1 Participants

467 One-hundred fifty participants were recruited through Prolific. This sample size was cho-
 468 sen based on a power analysis of a small pilot study and was preregistered.

469 **4.1.2 Procedure**

470 Stimuli, procedures, and exclusion criteria were identical to those in Experiment 2, except
471 that the onset delay of the second object varied across trials: 100 ms, 250 ms, 500 ms, and
472 1000 ms (Figure 5A). There was no Simultaneous (i.e., 0 ms) condition in this experiment.
473 Otherwise, the factors in the experiment were the same as in Experiment 2: 4 (Second Im-
474 age Delay) \times 2 (Object Order: Reference-first or Figure-first) \times 2 (Relation Type: Physical
475 or Spatial) \times 2 (Trial Type: Match or Mismatch) \times 2 (Sentence Structure: Figure-as-subject
476 or Reference-as-subject), yielding 64 trials. The four pairs of objects were fully distributed
477 among three primary factors—Second Image Delay, Object Order, and Relation Type—and
478 randomly assigned among other factors (Trial Type [Match or Mismatch] and Sentence
479 Structure [Reference-as-subject or Figure-as-subject]). Participants were given six practice
480 trials at the beginning of the experiment to become familiar with the task (all with 0-ms
481 delay between the first and second object).

482 **4.2 Results**

483 Twenty-four participants were excluded based on preregistered criteria, leaving 126 par-
484 ticipants for further analysis. The remaining participants had little difficulty completing
485 the study, with a mean accuracy of 97% and a mean response time (across all conditions)
486 of 865 ms.

487 We first conducted a repeated-measures ANOVA on participant means of inverse-
488 transformed response times (-1000/RT, correctly answered Match trials only) to examine
489 the main effects of interest in this experiment—Second Image Delay, Object Order, Relation
490 Type—as well as their interactions. Consistent with Experiment 2, the ANOVA revealed
491 a significant main effect of Object Order, $F(1, 125) = 85.89, p = 7.02 \times 10^{-16}, \eta^2 = 0.41$,
492 confirming an overall Reference-first RT advantage.

493 The ANOVA also revealed a significant interaction between Object Order and Second
494 Image Delay, $F(3, 375) = 4.35, p < 0.01, \eta^2 = 0.033$, suggesting that the magnitude of
495 Reference-first advantage differed depending on the onset delay of the second object. To
496 examine this further, we first computed the mean Reference-first RT advantage at each
497 level of delay (collapsing over Relation Type) by subtracting Reference-first from Figure-
498 first inverse RTs. One-sample t -tests showed a significant Reference-first advantage at all
499 delay conditions (100 ms: $t(125) = 2.5, p = 0.013, d = 0.22$; 250 ms: $t(125) = 4.7, p =$
500 $1.29 \times 10^{-5}, d = 0.22$; 500 ms: $t(125) = 5.8, p = 1.31 \times 10^{-7}, d = 0.22$; 1000 ms: $t(125) = 6.4,$
501 $p = 9.60 \times 10^{-9}, d = 0.22$). That is, participants were faster to verify relational scenes when
502 the reference object appeared right before the figure object, even with a minimal 100-ms
503 delay (Figure 5B).

504 To further explore how this effect unfolded over time, we conducted a series of paired-
505 sample t -tests to compare the Reference-first advantage between each pair of delay condi-
506 tions. The Reference-first advantage significantly increased from the 100-ms delay to the
507 500-ms delay ($t(125) = 3.14, p < 0.01, d = 0.28$), but then showed no further increase in
508 magnitude from 500 ms to 1000 ms, $t(125) = 0.25, p = 0.80$. An exploratory Bayesian

509 paired-sample *t*-test provided further evidence in favor of no difference between the 500-
510 ms and 1000-ms conditions ($BF_{01} = 0.1$, using the default Cauchy prior with scale $\sqrt{2}/2$).
511 In other words, the Reference-first RT advantage peaked at the 500-ms delay and then
512 plateaued (See Figure 5B).

513 Overall, the Reference-first advantage emerged even when the figure object appeared
514 just a brief moment after the reference object (just 100 ms). This effect quickly increased
515 and then plateaued after a half-second delay. These results suggest that the compositional
516 process for building relational representations from visual scenes is rapid and does not
517 rely on slow, deliberate reasoning processes, such as predicting what should happen next
518 after seeing a given object.

519 **5 EXPERIMENT 4 - Image-Only Composition**

520 The previous experiments found that relational scenes were matched more quickly to their
521 corresponding linguistic descriptions when the reference object appeared before the figure
522 object, illuminating the compositional process for between-object relations in visual recog-
523 nition. However, the task involved both visual and linguistic components: participants
524 compared a visual representation to a linguistic one (i.e., the previously presented sen-
525 tence). Considering the compositional nature of natural language, it is conceivable that
526 the linguistic probe could have influenced how the recognition process unfolded, without
527 any corresponding compositional process within recognition itself.

528 The present experiment minimized that potential influence by running the original
529 recognition study with one crucial change: we replaced the linguistic probe with a visual
530 one. This allowed us to test whether evidence of compositional visual processing emerges
531 independent of the format of the probe stimulus.

532 **5.1 Method**

533 **5.1.1 Participants**

534 Forty participants were recruited for this experiment. This sample size was the same as
535 Experiments 1 and 2.

536 **5.1.2 Procedure**

537 The same stimulus set from Experiments 2 and 3 was used in this recognition task. The
538 paradigm was similar to that used in Experiments 2 and 3, with one key difference: instead
539 of using linguistic descriptions as probes, each trial began with a pictorial probe showing
540 a target relational scene. As shown in Figure 6A, the probe image was presented for 350
541 ms, followed by a mask image for 350 ms. The mask was a box-scrambled version of all
542 object images (chosen randomly with replacement from 16 masks, made up of 22×22
543 blocks), and was included to interrupt the formation of afterimages and iconic memory
544 for the probe. After the mask, a blank screen of 100 ms and a fixation cross of 300 ms were

545 displayed. Then, either the figure object or the reference object was displayed, followed
 546 by the second object 500 ms later. At this point, participants responded whether the scene
 547 matched the probe image or did not. The experiment followed a 2 (Relation Type: Physical
 548 vs. Spatial) \times 2 (Object Order: Figure-first or Reference-first) \times 2 (Trial Type: Match or
 549 Mismatch) \times 4 (Object Pairs) design resulted in 32 unique trials. Each unique trial repeated
 550 twice, and there were also 6 practice trials at the beginning of the session, resulting in a
 551 total of 70 trials. Practice trials always featured a 500-ms delay between objects. The probe
 552 was presented centered on a frame with a gray background and dashed border, scaled to
 553 75% of the size of the target scene in order to avoid overlap in image features between
 554 probe and target.

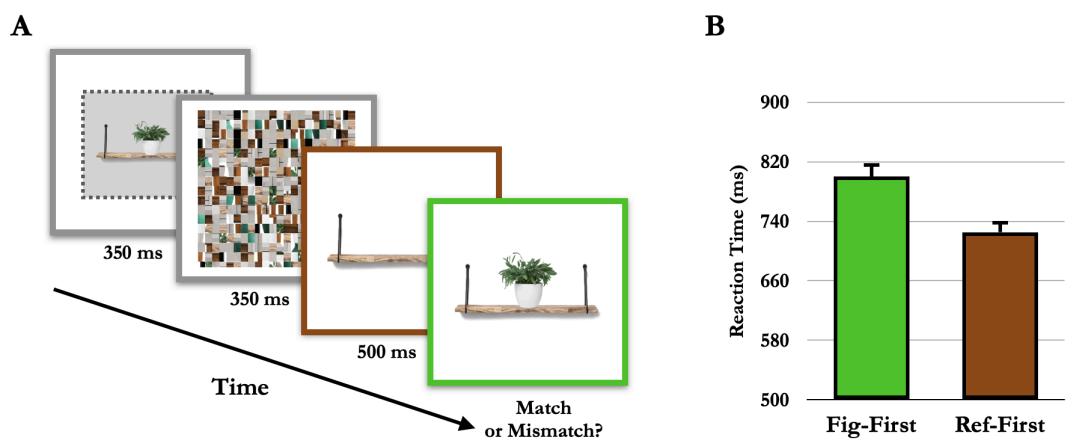


Figure 6: Trial structure and results of Experiment 4. (A) Participants viewed a relational scene; after being replaced by a mask, its constituent objects reappeared either in a Reference-first or Figure-first sequence (e.g., the shelf appearing before the plant). Participants then judged whether the second scene matched the first. (For simplicity, the blank display and fixation cross during the trial are omitted from this figure.) (B) Responses were faster when the reference object was displayed first. Bars reflect mean response times across participants (computed from correct trials only), and error bars reflect within-participant 95% confidence intervals.

555 5.2 Results

556 Six participants were excluded based on our preregistered exclusion criteria, leaving 34
 557 participants for further analysis. The remaining participants had little difficulty complet-
 558 ing the study, with a mean accuracy of 96% and a mean response time (across all condi-
 559 tions) of 773 ms.

560 As shown in Figure 6B, a Reference-first RT advantage was again observed: Partici-
 561 pants were faster to match the visual scene with the pictorial probe when the reference
 562 object appeared right before the figure object in the scene, rather than vice versa (725
 563 ms vs. 800 ms). A 2×2 repeated-measures ANOVA on participant means of inverse-
 564 transformed response times (-1000/RT, correctly answered Match trials only) that included
 565 Object Order and Relation Type as factors confirmed a significant main effect of Object Or-
 566 der ($F(1, 33) = 44.61, p = 1.33 \times 10^{-7}, \eta^2 = 0.57$). There was also a significant effect of
 567 Relation Type ($F(1, 33) = 112.4, p = 3.67 \times 10^{-12}, \eta^2 = 0.77$), with responses to physi-

568 cal relations being faster than those to spatial relations (690 ms vs. 836 ms). There was
569 no significant interaction between Object Order and Relation Type ($F(1, 33) = 0.51, p =$
570 $0.48, \eta^2 = 0.015$).

571 These findings demonstrate a Reference-first advantage independent of the format of
572 the probe stimulus. This suggests that the compositional process for building relational
573 representations can arise in visual processing alone, rather than arising only from a cogni-
574 tive comparison between representations formed from linguistic and visual modalities.

575 6 EXPERIMENT 5 - Identical-Object Composition

576 Our previous experiments consistently revealed a canonical order for visual recognition of
577 relational scenes, suggesting that reference objects, rather than figure objects, take prece-
578 dence in forming relational representations. However, in those experiments, the roles of
579 the objects were confounded with certain visual properties. As shown in Figure 2B, the
580 reference objects used (i.e., table, desk, nightstand, and shelf) were typically large, recti-
581 linear, and stable, while the figure objects (i.e., vase, laptop, lamp and plant) were smaller,
582 more mobile, and contained more rounded features. While an object's role in a relational
583 scene is often predictable from properties such as size, shape, or other mid-level visual fea-
584 tures, these properties are not essential for determining role. Instead, what often defines
585 Reference-hood, particularly for relations such as support-from-below ("on"), is whether
586 one object physically *controls* another (Landau & Gleitman, 2015; Talmy, 1975). These 'hid-
587 den' forces underlie many sophisticated relations between objects, transcending the lower-
588 level visual properties of individual objects, which are typically extracted efficiently in the
589 course of visual processing (Long, Konkle, Cohen, & Alvarez, 2016; Long, Yu, & Konkle,
590 2018). Thus, it remains unclear whether the Reference-first effect is driven solely by low-
591 or mid-level visual differences between objects or also by the more abstract relationships
592 between them: relations like ON, SUPPORT, ABOVE, or BELOW. The previous experiments
593 could not distinguish between these two possibilities.

594 To do so, we ran a final recognition experiment using a completely new set of object
595 stimuli. Instead of the previous objects, which differed in their visual properties (i.e., the
596 table, laptop, etc.), we used identical objects differing only in color: a red book and a blue
597 book. Consider the support relation depicted in Figure 7A. Here, the red book *supports* the
598 blue book by exerting a stabilizing force (against gravity). This 'hidden' force establishes
599 the red book as the reference object, while the blue book, being physically controlled, is the
600 figure object.

601 The situation differs for spatial relations without physical control, such as ABOVE.
602 Imagine a blue book above a red book; what determines role assignment here? Talmy
603 and others (Gleitman et al., 1996; Landau & Jackendoff, 1993; Talmy, 1975) noted that Fig-
604 ure and Reference roles parallel grammatical roles in language, with figure objects (e.g.,
605 a cat) often mapped to Subject position and reference objects (e.g., a mat) to Complement
606 position—the deeper position in syntactic structure (see Figure 7B). Crucially, this map-

607 ping is bidirectional: syntactic structure can shape how objects are interpreted as Figure
608 or Reference. For instance, the sentences *The bike is next to the garage* and *The garage is next*
609 *to the bike* differ in their implied interpretations. While the second sentence is unusual,
610 it becomes plausible in a context where the garage is mobile and the bike is stationary—
611 precisely the properties that figure and reference objects, respectively, often possess. This
612 suggests that syntax can guide relational interpretation in the absence of visual cues and
613 even ‘imbue’ participating objects with the typical properties of their assigned roles.

614 To test this, we reintroduced the sentence-picture verification task here. For physical
615 relations such as SUPPORT, we hypothesized that sentence structure would again not influ-
616 ence the Reference-first advantage, as Reference roles are determined by physical control
617 (Figure 7A). However, for spatial relations like ABOVE, we predicted that sentence struc-
618 ture would play a key role. For instance, in Figure 7B, where the sentence is *The blue book*
619 *is above the red book*, the blue book occupies Subject position and the red book Complement
620 position. Based on prior theoretical work (Gleitman et al., 1996; Landau & Jackendoff, 1993;
621 Talmy, 1975), we hypothesized that the Complement object—in this case, the red book—
622 would gain a response-time advantage, consistent with the Reference-first effect. Notably,
623 this does not correspond to a simple mapping between sentence order and image order.
624 For example, in Figure 7C, the Complement object (red book) is the *second* item mentioned
625 in the sentence but is expected to afford a processing advantage when viewed *first* in the
626 scene due to its role as Reference (determined by syntactic position in the sentence probe).²

627 By designing the experiment in this way, with new controlled object stimuli and a
628 sentence-verification task, we aimed to investigate the interplay between linguistic and
629 visual influences on the compositional process.

630 6.1 Method

631 6.1.1 Participants

632 In line with our preregistration, 100 participants were recruited through Prolific. This sam-
633 ple size was chosen based on a power analysis of a small pilot study.

634 6.1.2 Stimuli and Procedure

635 The relational scenes in this experiment consisted of two identical objects differing only
636 in color: a red book and blue book. As in previous experiments, we tested two types of
637 relations: physical and spatial. For physical relations, Reference and Figure roles were
638 determined by physical control, with the supporting object designated as the Reference

²We also piloted an image-only version of the task using the book stimuli, where participants viewed spatial relations (e.g., red book above blue book) without linguistic prompts. Preliminary data appeared to show a *bottom-first* RT advantage: Participants responded faster when the object positioned at the bottom of the scene (e.g., the blue book) appeared first. While we initially expected no preference for spatial relations, this result suggests that, in the absence of linguistic or physical-control cues (i.e., support-from-below), a bottom-first strategy may emerge as a default visual bias (see Langley & McBeath, 2023 for evidence that lower regions of scenes are perceptually salient). However, as will be shown, Experiment 5 revealed that the bottom-first bias for visual relations is flexible: Linguistic structure effectively overrides it for spatial relations like ABOVE, while physical control dominates role assignment for ON relations regardless of order in the linguistic probe.

639 (Figure 7A). For spatial relations, however, Reference and Figure roles were determined
 640 by the grammatical structure of the sentence probe (Figure 7B). For example, when the
 641 sentence “The red book is below the blue book” served as the linguistic probe, the gram-
 642 matical complement (i.e., blue book) functioned as Reference, and the grammatical subject
 643 (i.e., red book) functioned as Figure.

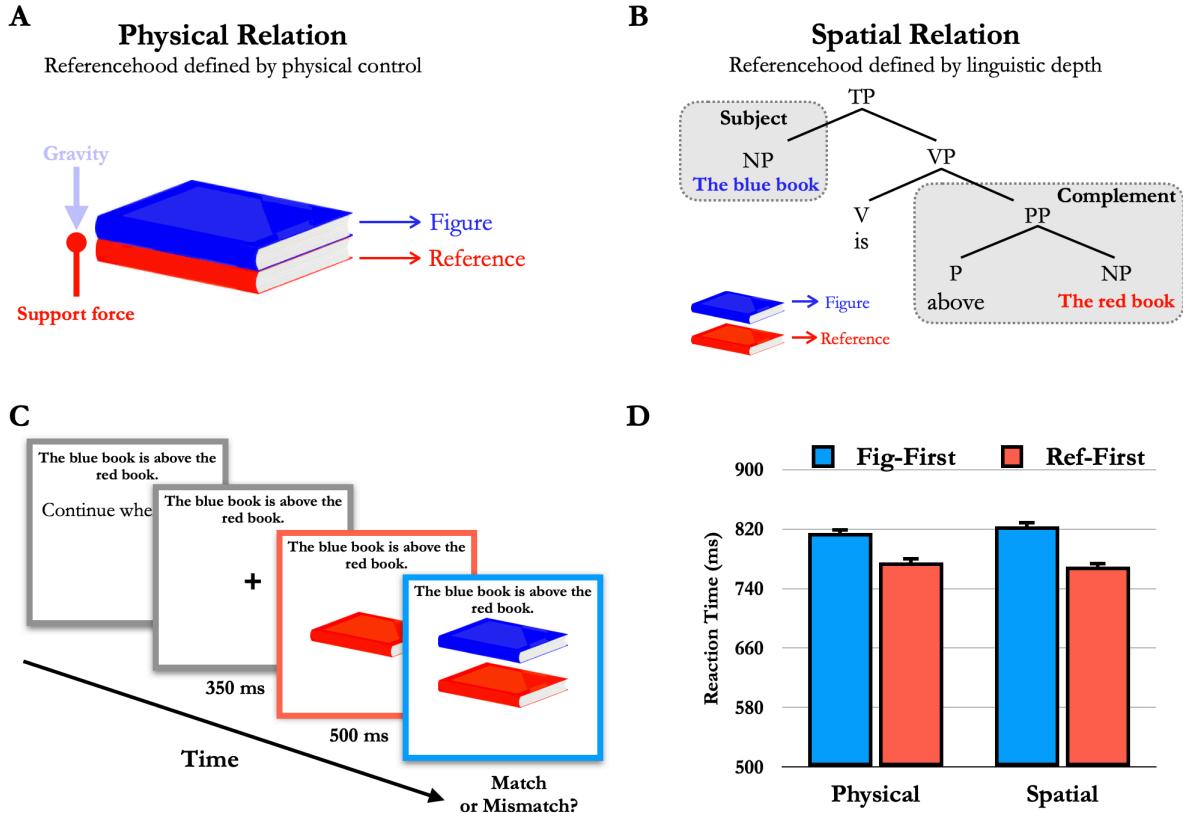


Figure 7: Illustration of the identical-object task in Experiment 5. Since the two objects differed only in color, their roles were determined by either physical control or linguistic structure. In the experiment, the colors of figure and reference object were counterbalanced. (A) For physical relations, the reference object provided support against the force of gravity for the figure object. (B) For spatial relations, the reference object was the entity embedded deeper in the linguistic structure than the figure object. (C) In the task, participants read a sentence description at the beginning of each trial and then verified whether the subsequent scene matched the description. (The 500-ms blank between sentence and fixation is omitted from the figure for simplicity.) (D) Results showed that participants were faster in verifying scenes when the reference object appeared before the figure object, for both physical and spatial relations. Bars reflect mean response times across participants (computed from correct trials only), and error bars reflect within-participant 95% confidence intervals.

644 As in previous experiments, there were several factors fully crossed within participant:
 645 (a) Relation Type (Physical or Spatial); (b) Book Order (Red or Blue book appearing first
 646 in the visual scene); (c) Sentence Structure (Red or Blue book as grammatical subject of
 647 the linguistic probe); and (d) Role Assignment (Red or Blue book serving as Reference in
 648 the visual relation). This resulted in 16 unique trials ($2 \times 2 \times 2 \times 2$), with each repeated
 649 4 times, yielding 64 Match trials. In addition, there were 64 Mismatch trials, consisting of
 650 3 types: (i) Role Mismatch (16 trials): The Reference and Figure roles were swapped (e.g.,

651 if the sentence stated, “The red book is on the blue book”, the visual scene depicted a *blue*
652 book on a *red* book); (ii) Relation Mismatch (16 trials): The relational category was changed
653 (e.g., if the sentence stated, “The red book is on the blue book”, the visual scene depicted
654 a blue book *above* a red book); and (iii) Full Mismatch (32 trials): Both role assignment and
655 relational category were inconsistent with the sentence. The full experiment consisted of
656 136 trials (8 practice trials, 64 Match trials, and 64 Mismatch trials). The 128 test trials were
657 presented in a fully randomized order for each participant.

658 For analyses, we created a new Object Order factor (with Reference-First and Figure-
659 First conditions), where Reference and Figure were defined differently based on the Rela-
660 tion Type. For physical relations, the supporting object was Reference and the supported
661 object was Figure. By contrast, for spatial relations, these were based on grammatical posi-
662 tion in the sentence probe: the grammatical object served as Reference and the grammatical
663 subject as Figure.

664 6.2 Results

665 In accordance with our preregistered exclusion criteria, 16 participants were excluded,
666 leaving 84 participants for further analysis. The remaining participants had little difficulty
667 completing the study, with a mean accuracy of 97% and a mean response time (across all
668 conditions) of 846 ms.

669 As shown in Figure 7D, even though the figure and reference objects were identical
670 in size and shape, participants were still faster verifying the relational scene when the
671 reference object appeared before the figure object than vice versa (772 ms vs. 820 ms).
672 This Reference-first RT advantage was confirmed by a 2×2 repeated-measures ANOVA,
673 which revealed a main effect of Object Order ($F(1, 83) = 57.40, p = 4.51 \times 10^{-11}, \eta^2 =$
674 0.41). There was also a marginally significant interaction between Relation Type and Object
675 Order ($F(1, 83) = 3.77, p = 0.056, \eta^2 = 0.043$), but no significant main effect of Relation
676 Type ($F(1, 83) = 0.12, p = 0.74, \eta^2 = 0.0014$).

677 To further examine the effect of Object Order within each Relation Type, we conducted
678 Holm-Bonferroni-corrected paired-samples *t*-tests separately for physical relations and spa-
679 tial relations. Both analyses confirmed the advantage for the Reference-first order (Phys-
680 ical: ($t(83) = 7.70, p_{corrected} = 5.20 \times 10^{-11}, d = 0.84$; Spatial: $t(83) = 4.03, p_{corrected} =$
681 $1.24 \times 10^{-4}, d = 0.44$). Finally, to determine whether the Object-Order effect was stronger
682 for certain relational words over others, we conducted two separate paired *t*-tests on the
683 Reference-first RT advantage (computed as in Experiment 3): one for physical relations
684 (“on” vs. “supporting”) and one for spatial relations (“below” vs. “above”). Neither test
685 was significant ($t_s(83) < 1.37$, corrected $ps > 0.34$).

686 This experiment disentangled objects’ relational roles from relevant visual properties
687 (e.g., size, shape), allowing us to draw several conclusions. First, the persistence of the
688 Reference-first RT advantage suggests that the Reference-first advantage may be driven
689 by objects’ abstract roles as Reference and Figure, rather than solely from their visual fea-
690 tures. Second, when the asymmetry between objects is not given by their visual differ-

691 ences intrinsic to the objects (e.g., size, shape) nor by their physical relationship (e.g., via
692 physical control), the syntactic structure of a linguistic description can guide Figure and
693 Reference assignments, shaping the order in which relational representations are mentally
694 constructed.³

695 7 General Discussion

696 The present work investigated how everyday visual relations (such as one object supporting
697 another) are represented in the mind, revealing key principles governing how they are
698 composed in time. In a manual construction task (Experiment 1), participants assembled
699 relational scenes by placing reference objects (e.g., tables, desks) first, followed by figure
700 objects (e.g., vases, laptops). Similarly, in a series of visual recognition tasks (Experiments
701 2–5), participants demonstrated a Reference-first advantage, responding faster when the
702 reference object appeared before the figure object. This Reference-first advantage emerged
703 rapidly—within just 100 ms—and plateaued by 500 ms (Experiment 3), suggesting that
704 it does not rely on deliberative expectations about object order. Notably, this effect per-
705 sisted even without linguistic prompts, suggesting that the visual system independently
706 employs a Reference-first compositional routine (Experiment 4). Finally, we found that
707 relational roles (Reference or Figure) are assigned based on different sources of informa-
708 tion. While visual properties or physical control appear to be primary drivers of Figure
709 and Reference role assignments, linguistic descriptions can guide role assignments in the
710 absence of large asymmetries in these cues, and thus dictate the order in which to build
711 relational representations (Experiment 5). Taken together, our findings suggest that the
712 mind automatically employs a sequential routine for composing relational representations,
713 respecting each object’s role in the relation: Reference first, Figure second.

714 7.1 Understanding perception as a compositional process

715 Our work aligns with a growing body of research exploring compositional representations
716 in visual perception. Classic theories of object representation propose that objects are not
717 processed as undifferentiated wholes, but rather as hierarchically decomposed structures
718 consisting of parts and subparts (Biederman, 1987; Feldman & Singh, 2006; Marr & Nishi-
719 hara, 1978). Such hierarchical structures not only define the relations among constituent
720 parts but also establish a principle of primacy, whereby certain components carry greater
721 weight than others in the representation. For example, the main skeletal axis of a shape is

³To further test whether the linguistic effects in Experiment 5 were driven by syntactic structure rather than surface word order, we piloted a version of the task using *it*-cleft sentences (e.g., *It is the red book that is below the blue book* vs. *It is the red book that the blue book is below*). These sentences reverse the order in which the objects are mentioned while preserving the deeper syntactic structure (i.e., which entity serves as subject vs. complement). Preliminary results trended in the predicted direction—consistent with syntactic structure driving the effect—but the task proved challenging for participants due to the processing complexity of *it*-cleft sentences, leading to longer and more variable response times. For this reason, we did not pursue a full version of this study, though the pilot findings support the conclusion that syntactic structure, rather than linear word order, determines relational composition in such cases.

722 given greater representational priority than the ‘ribs’ that define the peripheral parts of a
723 shape, reflecting an intrinsic ordering principle in object perception (El-Gaaly, Froyen, El-
724 gammal, Feldman, & Singh, 2015; Feldman & Singh, 2006). This hierarchical precedence is
725 even reflected in how object representations are ‘grown’ or generated in the mind: Mount-
726 ing empirical evidence suggests that such representations are composed in a systematic
727 order, where main axes emerge before subordinate branches, reinforcing the psychological
728 reality of these processes (Ayzenberg & Lourenco, 2022; Destler, Singh, & Feldman, 2023;
729 Sun & Firestone, 2022).

730 Our findings extend this principle beyond individual objects to between-object rela-
731 tions: Just as certain object parts take precedence in forming object representations, certain
732 relational constituents (here, reference objects) take precedence in forming relational rep-
733 resentations. In doing so, our work contributes to growing scientific attention on visual
734 relations as a fundamental unit of perception and cognition (for reviews and discussion,
735 see Cavanagh, 2021; Hafri & Firestone, 2021; Hafri, Green, & Firestone, 2023; Hafri & Pa-
736 peo, in press; Hochmann & Papeo, 2021; Hummel & Holyoak, 2003; Kaiser, Quek, Cichy, &
737 Peelen, 2019; Miller & Johnson-Laird, 1976; Papeo, 2020; Peelen, Berlot, & de Lange, 2024;
738 Quilty-Dunn et al., 2023; Võ et al., 2019).

739 The fact that compositional representations exist both within and between objects may
740 suggest a fundamental and general kind of compositional process that extends across di-
741 verse relational domains. One key example comes from social cognition, where event
742 representations are sensitive to certain compositional orders for relational roles like Agent
743 and Patient. In particular, Agents are prioritized in perception, recognition, and predic-
744 tion of events, substantiating their primary role in relational encoding (Brocard, Wilson,
745 Berton, Zuberbühler, & Bickel, 2024; Cohn & Paczynski, 2013; Cohn et al., 2017; Hafri et
746 al., 2018; Sauppe & Flecken, 2021; Ünal, Wilson, Trueswell, & Papafragou, 2024; Wilson,
747 Zuberbühler, & Bickel, 2022). Recent work shows that these event-based relational repre-
748 sentations exist even in the minds of preverbal infants, indicating that such compositional
749 processes do not rely on natural language but are rooted in early-emerging non-linguistic
750 cognitive capacities (Hafri, 2024; Papeo et al., 2024). Our findings extend these insights
751 from social relations to physical and spatial ones, demonstrating that compositional prin-
752 ciples govern not only event structure but also the perception of objects in space. This is
753 consistent with the idea that relational structure, rather than individual objects, form the
754 backbone of visual scene perception writ large.

755 In this way, the present work also dovetails with a renewed interest in the ‘Language-
756 of-Thought’ (LoT) hypothesis (Fodor, 1975), which has re-emerged in debates over the
757 format of mental representation (see Quilty-Dunn et al., 2023, and commentaries therein).
758 The LoT framework proposes that certain types of mental representations are composed of
759 discrete, combinable elements—a principle originally formulated to explain systematicity
760 and productivity in language and high-level reasoning. Recent extensions of this frame-
761 work suggest that LoT-like representations may exist across many cognitive systems, in-
762 cluding in visual perception (Mandelbaum et al., 2022). In a recent article (Hafri, Green, &

763 Firestone, 2023), we built on this idea, arguing that visual perception exhibits core LoT-like
764 properties, where discrete perceptual units (e.g., object parts or whole objects) combine
765 systematically.

766 The current work extends this proposal not only by reinforcing the notion that visual
767 perception involves compositional representations, but also by revealing the *processes* by
768 which they are constructed—in other words, the ‘psychophysics’ of relational composition,
769 or the timing and ordering of how relational structures are built from their parts. Our
770 results provide evidence for a sequential combinatorial process that operates over abstract
771 relational roles and generalizes beyond event structure to spatial and physical relations,
772 advancing beyond previous work in this area (Boettcher et al., 2018; Franconeri, Scimeca,
773 Roth, Helseth, & Kahn, 2012; Holcombe et al., 2011). Our findings suggest that visual
774 composition is not an instantaneous process, but one that unfolds in a structured, role-
775 sensitive manner, akin to how linguistic syntax forms a structural basis for interpreting
776 utterances.

777 7.2 The interface between vision and language

778 Our findings highlight a novel aspect of how visual and linguistic systems might interact
779 during scene perception. On the one hand, the exact word order of the linguistic prompts
780 we employed did not generally alter the order by which relational representations were
781 constructed (indexed by RT differences between Reference- and Figure-first orders). On
782 the other hand, Experiment 5 demonstrated that linguistic structure can guide the assign-
783 ment of Figure and Reference roles when strong visual cues to asymmetry are absent (e.g.,
784 in visually symmetric relations such as *above* or *below*; Gleitman et al., 1996; Talmi, 1975).

785 More broadly, our findings suggest that while the visual and linguistic systems may
786 operate largely independently, they converge under conditions of perceptual ambiguity:
787 Linguistic descriptions typically exert little influence on scene processing (at least of the
788 kind studied here), but when perceptual cues are ambiguous, language can ‘imbue’ objects
789 with relational roles, influencing the compositional process. However, we suspect that this
790 influence of language does not directly alter visual representations themselves (e.g., by
791 changing the objects’ appearance in the scene; Firestone & Scholl, 2016). Instead, it may
792 function as a kind of ‘cognitive instruction,’ guiding how observers attend to and encode
793 an upcoming image (Knowlton et al., 2021)—albeit in a manner that may not reach explicit
794 awareness. Crucially, whatever the precise mechanism by which language influences scene
795 processing here, its effect is not determined merely by the probe’s surface word order
796 (where the reference object appears second) but by its syntactic structure—that is, by what
797 occupies the deeper, hierarchically primary position.

798 This pattern joins classic and more recent literature detailing the interactions between
799 linguistic and visual systems (Cavanagh, 2021; Jackendoff, 1987; Miller & Johnson-Laird,
800 1976; Strickland, 2017). For example, in recent work, we found that the mind is sensi-
801 tive to correspondences between linguistic and visual notions of symmetry (Hafri, Gleit-
802 man, Landau, & Trueswell, 2023). Despite the striking differences between a butterfly’s

803 appearance and a sentence like *Mary and Bill marry*, both share an abstract symmetry—
804 an invariance to transformation. In images, this is evident in the bilateral symmetry of a
805 butterfly; in language, it is reflected in flexible argument order for certain predicates like
806 *marry* (e.g., *Mary marries Bill* vs. *Bill marries Mary*). In cross-modal matching tasks, we
807 observed surprising correspondences across such stimuli, providing evidence for these in-
808 tuitive psychological connections between symmetry in vision and language. These and
809 other findings (e.g., [De Freitas & Alvarez, 2018]) suggest that the mind employs common
810 formats and principles across cognitive systems—allowing perceptual representations of
811 relations to be readily accessed by higher-level processes ([Hafri, Green, & Firestone, 2023;
812 Quilty-Dunn, 2020]) and, in some cases, enabling linguistic representations to guide atten-
813 tional patterns in scene perception.

814 7.3 Open questions and future directions

815 These connections across cognitive systems raise a broader question: Why do syntactic
816 roles in language appear to align so strongly with perceptual compositional order? In
817 syntactic structures, complements—where reference objects often reside—occupy deeper
818 positions in the hierarchy than subjects (where figure objects often end up). In many
819 syntactic frameworks, linguistic structures are assumed to be derived bottom-up ([Chom-
820 sky, 1995]). Likewise, psycholinguistic evidence shows that planning and parsing unfold
821 incrementally—often anticipating material that is syntactically deeper, even if it appears
822 earlier in the linear sequence ([Momma & Phillips, 2018]). This pattern mirrors our find-
823 ings and may reflect a general cognitive principle in which structure is built incrementally
824 from the most foundational elements (e.g., reference objects) upward. While speculative,
825 this suggests a striking parallel between visual composition and linguistic derivation that
826 warrants further investigation.

827 Our findings also highlight an intriguing dissociation between event and non-event
828 relational representations in the alignment of compositional order and grammatical or-
829 der. In event representations, Agents—typically dynamic entities that move or initiate
830 change—are psychologically primary and usually mapped to grammatical subject posi-
831 tion. In contrast, for physical and spatial relations like those studied here, reference objects
832 are primary yet are generally mapped to grammatical *complement* positions, occupying
833 deeper levels in hierarchical syntactic structure. While speculative, we suggest that this
834 difference in mapping may reflect underlying differences in the functional properties of
835 Agents and reference objects. Dowty’s (1991) proto-role theory aims to formalize how pro-
836 totypically Agent-like properties (such as movement or initiating change) determine the
837 mappings between event participants (arguments of verbs like *kick* or *fear*) and grammati-
838 cal roles (subject and object); however, this framework does not straightforwardly extend
839 to non-event relations. Reference objects illustrate this tension clearly: although capable
840 of initiating change (an “Agent-like” property—e.g., a table moving can cause a vase atop
841 it to move as well), they typically function as stable, stationary anchors (a more “Patient-
842 like” property). Thus, a complementary theoretical framework may be needed to capture

843 generalizations across non-event relations like those studied here.

844 Our work also opens important avenues for future research. One critical question con-
845 cerns how the mind learns to identify reference objects as such, as well as their capacity to
846 control other objects in physical or spatial relations. Experiments 1–4 showed that visual
847 properties such as size, stability, and rectilinearity often determine Reference role assign-
848 ment, consistent with findings on “anchor” objects in scene perception (Boettcher et al.,
849 2018; Võ et al., 2019). However, the results of Experiment 5 reveal that Reference role
850 assignment can also emerge from more abstract principles, such as intuitive physics (see
851 also Firestone & Scholl, 2017; Little & Firestone, 2021) or even linguistic structure. Devel-
852 opmental research suggests that pre-linguistic infants are sensitive to relational concepts
853 like support and containment (Baillargeon et al., 2012; Hespos & Spelke, 2004), raising
854 the possibility that early interactions with physical forces inform the mind’s sensitivity to
855 Reference-hood. How these early experiences integrate with visual properties and higher-
856 level conceptual principles remains an open question.

857 Finally, another important open question is what additional visual routines under-
858 lie relational composition beyond those documented for simple geometric relations (Joli-
859 coeur et al., 1986, 1991; Ullman, 1987; Wong & Scholl, 2024). While our work uncovered
860 a Reference-first order for spatial and physical relations, other routines may contribute to
861 the construction of everyday relational scenes. Moreover, questions remain about how
862 more complex relations are represented compositionally, beyond the simpler dyadic cases
863 studied here. For instance, transfer events (e.g., *giving*) or caused motion (e.g., hitting a
864 ball with a racket into the net) involve at least three roles (Tatone & Csibra, 2024; Ünal
865 et al., 2024). Likewise, relational structures may be embedded. Consider a cat on a mat
866 that is in a box. Is this situation represented as two independent relations (cat-on-mat
867 and mat-in-box) or as a single embedded relation (cat-on-mat in box)? Our approach, us-
868 ing a sentence–picture verification task with delayed presentations, may prove fruitful in
869 addressing these questions.

870 7.4 Conclusions

871 In sum, this work reveals a fundamental principle of relational composition: the mind con-
872 structs relational representations sequentially, respecting the roles of elements in the rela-
873 tion. By uncovering a Reference-first compositional routine, we show that visual percep-
874 tion employs a structured process akin to compositional principles observed in language
875 and event cognition. More broadly, these findings provide new insights into the nature of
876 visual representation, the underlying construction algorithms, and the deep connections
877 between perception, language, and intuitive physics.

878 Acknowledgments

879 For helpful discussion and/or comments on drafts of this article, the authors thank Re-
880 becca Tolland and members of the University of Delaware Perception & Language Labora-

881 tory. This work was supported by National Science Foundation Grant No. BCS-2021053
882 awarded to C. Firestone and National Science Foundation Directorate for Social, Behav-
883 ioral and Economic Sciences (SBE) Postdoctoral Research Fellowship No. SMA-2105228
884 awarded to A. Hafri.

885 **CRediT authorship contribution statement**

886 **Zekun Sun:** Conceptualization, Methodology, Software, Investigation, Formal analysis,
887 Writing - Original Draft, Writing - Review & Editing, Visualization. **Chaz Firestone:** Con-
888 ceptualization, Methodology, Writing - Review & Editing, Supervision, Funding acquisi-
889 tion. **Alon Hafri:** Conceptualization, Methodology, Software, Investigation, Formal anal-
890 ysis, Writing - Original Draft, Writing - Review & Editing, Visualization, Supervision.

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