

Tangled physics: Knots strain intuitive physical reasoning

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Abstract

Whereas decades of research have cataloged striking errors in physical reasoning, a resurgence of interest in intuitive physics has revealed humans' remarkable ability to successfully predict the unfolding of physical scenes. A leading interpretation intended to resolve these opposing results is that physical reasoning recruits a general-purpose mechanism that reliably models physical scenarios (explaining recent successes), but overly contrived tasks or impoverished and ecologically invalid stimuli can produce poor performance (accounting for earlier failures). But might there be tasks that persistently strain physical understanding, even in naturalistic contexts? Here, we explore this question by introducing a new intuitive physics task: evaluating the strength of knots and tangles. Knots are ubiquitous across cultures and time-periods, and evaluating them correctly often spells the difference between safety and peril. Despite this, 5 experiments show that observers fail to discern even very large differences in strength between knots. In a series of two-alternative forced-choice tasks, observers viewed a variety of simple "bends" (knots joining two pieces of thread) and decided which would require more force to undo. Though the strength of these knots is well-documented, observers' judgments completely failed to reflect these distinctions, across naturalistic photographs (E1), idealized renderings (E2), dynamic videos (E3), and even when accompanied by schematic diagrams of the knots' structures (E4). Moreover, these failures persisted despite accurate identification of the topological differences between the knots (E5); in other words, even when observers correctly perceived the underlying structure of the knot, they failed to correctly judge its strength. These results expose a blindspot in physical reasoning, placing new constraints on general-purpose theories of scene understanding.

Keywords:

intuitive physics, visual perception, simulation

Significance Statement

Intuitive physics research has largely focused on rigid-body objects and systems, with recent work revealing strikingly successful reasoning about their physical behavior. The present study introduces a novel stimulus class to this domain of research: knots. Despite being pervasive in everyday life, from tying our shoes to rock climbing, little is known about how well intuitions about the physical properties of knots, such as their resistance to pulling force, map onto their known physical properties. Remarkably, 5 experiments demonstrate that observers fail to produce correct judgments about the strength of very simple knots, revealing a blindspot in theories of physical reasoning. This work may not only prompt further exploration of knots in intuitive physics research (and beyond), but also testifies to the importance of ordinary everyday phenomena that are often overlooked when studying psychological processes.

1 Introduction

2 Look at the images in Figure 1A. One of the knots depicted there is a staple of
3 sailing and scouting practice, widely used across different cultures and historical eras
4 to secure belongings, join lengths of string, and otherwise fasten and bind materials.
5 The other is essentially a ‘trick’ knot; it is so insecure that it often comes apart on its
6 own, and relying on it for anything practical would invite disastrous consequences
7 (whether for your safety or the security of your belongings). Can you tell which
8 is which? In other words, which knot seems like it would remain intact if pulled
9 strongly at both ends, and which would easily capsize?

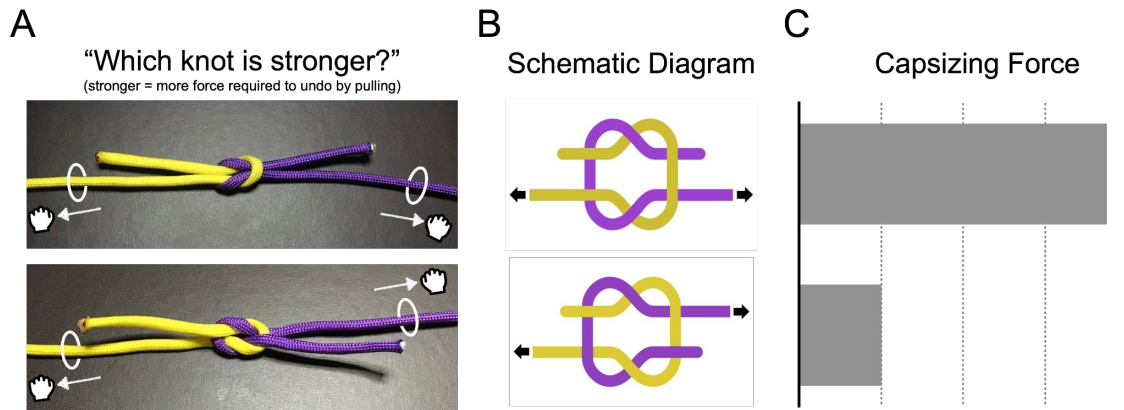


Figure 1: (a) Imagine pulling the longer ends of the two knots displayed here. Can you guess which one withstands the most pulling force? (The answer is revealed later in this caption.) (b) Schematic diagram showing the topological organization of each knot from panel A. Notice, for example, the relative placement of the two pulled strands (i.e., those with arrows on them); in the top knot, the two pulled ends are on the same side as one another (yellow and purple both below), whereas in the bottom knot, the two pulled ends are opposite one another (yellow is below and purple is above). (c) Despite minimal topological differences, the reef knot (top) is substantially stronger than the reef knot (bottom), as measured by the force required for it to capsize (i.e., collapse or come apart). Readers can see this for themselves at <https://perceptionresearch.org/knots>, which features a video of author S.C. attempting to undo each of them.

10 Judgments about physical scenarios and events pervade our daily lives, from de-
11 ciding whether the stack of dishes in our sink can withstand another plate, to choosing
12 how hard to push a child on a swing. However, the nature and accuracy of these
13 judgments has been the subject of debate across different approaches and research
14 traditions in psychology. Early work investigating physical reasoning cataloged many
15 striking and surprising contexts in which physical intuitions sharply deviate from the
16 principles of Newtonian physics. For example, when asked to predict the trajectory
17 of an object dropped from an airplane, or to trace the path of a ball exiting a spiral
18 tube, even highly educated college students (including those with formal physics ed-
19 ucation) make odd and persistent errors, such as believing that objects always fall
20 straight down rather than maintaining their lateral momentum (McCloskey et al.,
21 1980; McCloskey, 1983; Cook and Breedin, 1994; Gilden and Proffitt, 1994). These
22 and other errors motivated theories of physical reasoning as a heterogeneous and
23 inconsistent set of heuristics that are employed in specific contexts, with varying
24 degrees of (in)accuracy (for a review, see Kubricht et al., 2017).

25 However, a different perspective has emerged more recently, driven by newer
26 results that highlight surprisingly successful physical reasoning. For example, ob-
27 servers can correctly and rapidly predict whether and how a tower of blocks will
28 fall (Battaglia et al., 2013; Firestone and Scholl, 2016, 2017), the relative masses of
29 objects participating in collisions (Hamrick et al., 2016), and even the proportion of
30 a poured liquid that will end up on either side of a partition (Bates et al., 2019).
31 These and other successes have motivated a different account, in which physical intu-
32 itions derive from a rich, probabilistic, generative model of the world and its physical

33 laws, rather than the application of rough and ready heuristics. One especially in-
34 triguing hypothesis in this domain is that such models and simulations resemble the
35 software architectures used in gaming environments (Battaglia et al., 2013; Ullman
36 et al., 2017). According to this view, observers infer the future state of the world
37 by running simulations in a mental “intuitive physics engine” (IPE), and treat the
38 outputs of this engine (which may be subject to perceptual noise and uncertainty)
39 as statistical samples from which to make physical inferences. These features of the
40 IPE allow for sufficiently accurate predictions in most everyday scenarios (though
41 they may also be subject to occasional illusions and biases, perhaps as a result of
42 limited cognitive resources). More generally, accounts of this sort tend to embrace
43 general-purpose approaches to physical reasoning, on which the mind applies roughly
44 the same principles and architecture to a wide variety of physical reasoning tasks.

45 *Reconciling successes and failure: Naturalism and context*

46 These two research traditions, one older and one more recent, offer conflicting
47 perspectives on the nature and accuracy of intuitive physical reasoning. How do
48 the more recent views emphasizing success account for the many failures observed
49 earlier?

50 A leading approach has been to explain away earlier failures by appealing to the
51 contrived or impoverished nature of the stimuli and tasks used in previous studies.
52 For example, whereas early work reported striking errors when subjects must use a
53 pen to trace the future trajectory of a weight cut from a swinging pendulum (Car-
54 mazza et al., 1981), more recent work discovered that if the pendulum is animated
55 and subjects must move a cup to catch the weight, they behave much more accu-

56 rately (Smith et al., 2018). Indeed, many other intuitive misconceptions reported
 57 in early research may be ameliorated or abolished by the use of more naturalistic
 58 and dynamic stimuli and tasks, such as rich, animated scenes (Kaiser et al., 1992),
 59 more familiar and ecologically valid tasks and contexts (Kaiser et al., 1986), and
 60 measures that prompt simulated or imagined actions (Schwartz and Black, 1999);

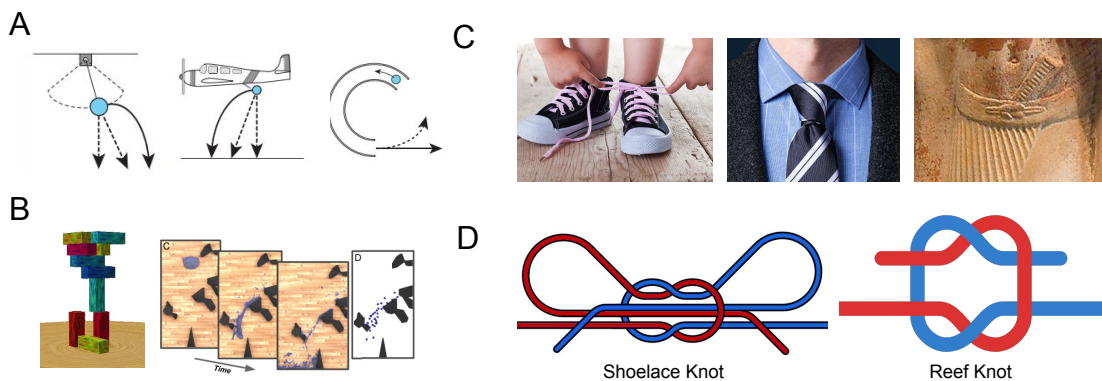


Figure 2: Example stimuli from intuitive physics research. (a) Early studies of intuitive physics revealed systematic errors in judgment. When participants are instructed to identify the trajectory of the blue target object, they are reliably inaccurate. For example, participants predict that a ball cut from a swinging pendulum or dropped from a moving plane will take a straight path to the ground rather than a curved one. Conversely, naive participants tend to believe that a ball exiting a spiral tube will continue on a curved trajectory rather than exiting on a straight path. (Adapted from Kubricht et al., 2017.) (b) More recent intuitive physics research has revealed more accurate and reliable judgments. When participants are instructed to judge the stability of a block tower or the flow of a poured liquid over obstacles, they demonstrate subtle and reliable understanding of these physical scenarios. This evidence has been taken to support a general-purpose mechanism for simulating the unfolding of physical scenes, especially when using naturalistic stimuli (as compared to earlier studies using diagrams). (Adapted from Hamrick et al., 2016; Bates et al., 2019.) (c) The present work explores intuitive judgments about knots. Knots are used in a wide variety of contexts, ranging from specialized activities such as sailing, rock climbing and survivalism to more mundane activities such as tying one’s shoelaces or a necktie. The rightmost image shows a reef knot (the same kind of knot seen in Figure 1A) around the belt of a figure in an Ancient Egyptian sculpture ca. 2350 BCE — evidence that these knots have been in use across cultures and time periods. (d) As shown in schematic diagrams, a typical shoelace knot is far more complex than the reef knot (and its variations) that we study here, and indeed even ‘contains’ a reef knot at its core.

61 see also discussion in [Fischer and Mahon \(2021\)](#), who propose that “first-person”
62 or user-oriented tasks produce better physical judgments than third-person problem
63 solving. In light of these and other results, it has more recently been proposed that
64 “the contrast between rich and calibrated versus poor and inaccurate patterns of
65 physical reasoning exists as a result of using different systems of knowledge across
66 tasks” ([Smith et al., 2018](#)), and that “when using more-realistic displays and actions,
67 our intuitions actually closely match Newtonian dynamics” ([Ullman et al., 2017](#)).

68 Thus, intuitive physics research has expanded to include more familiar and eco-
69 logically valid physical reasoning tasks, and there is evidence that this addition of
70 richness and context may account for certain failures observed earlier. However,
71 there are many physical systems and behaviors that are part of our everyday lives
72 but have remained almost completely unexplored in this literature. Might any of
73 those domains put pressure on the above consensus? In other words, might there be
74 a class of stimuli and tasks that both (a) are naturalistic, familiar, and intertwined
75 with daily life, and yet (b) dramatically strain human physical scene understanding?

76 Identifying such cases is important because it may reveal boundary conditions or
77 constraints on the general-purpose nature of physical reasoning mechanisms. Dis-
78 covering which stimuli and tasks are easy and which are difficult may serve as crucial
79 data to ultimately inform a complete theory of physical scene understanding (since
80 any such theory will have to account for both successes and failures).

81 *Introducing knots to the study of physical reasoning*

82 Here, we introduce such a stimulus class to the study of intuitive physics, by
83 exploring human judgments about *knots*. Knots are naturalistic stimuli that appear

84 across cultures and time periods. For example, art from Ancient Egypt (ca. 2350
85 BC) depicts the classic “reef” knot around a person’s waist (Louvre, 1938), and there
86 is similar evidence from Ancient Greece, Ancient and Imperial China, and even pre-
87 historic societies that engaged in sewing and other clothwork (d’Errico et al., 2018;
88 Leroi-Gourhan, 1982). It is often thought that knots predate human use of both
89 fire and the wheel (Turner and van de Griend, 1996), and there is also evidence of
90 cordage production among Neanderthals (Hardy et al., 2020); even non-human ani-
91 mals employ tangled structures in nest-building, predation, and other practices (for
92 example, see Herzfeld and Lestel, 2005, for a fascinating ethnographic study of an
93 orangutan who can tie “true” knots using her hands, feet, and mouth). Moreover,
94 knots are widely used both in mundane scenarios (e.g., tying one’s shoelaces or the
95 drawstring of a bag) and in more technical applications where one’s knot selection
96 and skill can spell the difference between safety and peril (e.g., sailing or rock climb-
97 ing). We’re also often tasked with *untying* knots, such as when headphone cords or
98 necklaces become tangled in one’s pocket.

99 Knots can also be depicted in a variety of styles and representations, including
100 naturalistic images and animations, as well as abstract idealizations and diagrams
101 (i.e., in both of the formats popular in previous intuitive physics research). More-
102 over, their physical properties can be precisely characterized. For example, recent
103 research in the domains of topology and applied physics has simulated and exper-
104 imentally investigated the physical mechanics of many popular knots (Patil et al.,
105 2020), allowing for a ground-truth baseline against which to test human intuition.
106 However, knots remain almost completely unexplored in intuitive physics research,

107 despite suggestions that they may form a rich and promising domain for investigation
108 ([Santos et al., 2019](#)).

109 The present work enters this new domain by examining the ability of naive human
110 subjects to evaluate the strength of various knots and tangles. As a case study, we
111 focus on a series of 2-tangle knots that join lengths of string, known as the “reef”,
112 “thief”, “granny” and “grief” series. These knots, depicted in [Figure 3](#), are quite
113 visually similar, and yet they vary widely in their stability, which is operationalized
114 as the amount of force required for them to capsize: Reef knots (one of the most
115 prevalent and recognizable knots in the world) are much stronger than thief knots;
116 similarly, granny knots are much stronger than grief knots. This is true not only
117 according to the cultural knowledge and practices of the communities that use (or
118 avoid) these knots (such as sailors and scouts), but also according to recent scientific
119 studies of them. For example, [Patil et al. \(2020\)](#) specifically examined the mechanics
120 of this series of knots and concluded through computer simulations and real-world
121 experiments that the received wisdom about these knots is accurately reflected in
122 their physical behavior.

123 Surprisingly, the knots in this series are often distinguished only by the position of
124 a single thread, and yet they differ dramatically in strength. In fact, the uppermost
125 knot in [Figure 1A](#) (a reef knot) is many times stronger than the lowermost knot
126 (a grief knot), despite their relatively minimal visual and topological differences.
127 (Indeed, the *Ashley Book of Knots*, an authoritative and widely referenced source
128 on knotcraft, calls the grief knot “hardly a practical knot” and instead considers it
129 merely “an interesting trick”; [Ashley, 1944](#).)

130 Importantly, the knots mentioned here are (a) among the simplest knots that can
131 be tied with two lengths of string, and (b) quite prevalent in daily life (even if they
132 may not initially seem that way). For example, the standard “shoelace knot” that
133 many of us tie every morning contains within it a reef knot (such that the reef knot
134 is, by definition, *simpler* than the shoelace knot). And a granny knot is simply two
135 half knots tied one after the other. Thus, chances are that *you* have frequently tied
136 this knot without realizing it (e.g., to secure sweatpants or a bag, or simply in the
137 course of tying your shoelaces; [Skwarecki, 2023](#)). Thus, if it turns out that ordinary
138 people *cannot* easily intuit the strength of these simple and pervasive knots, then it
139 is quite likely that even less familiar and/or more complicated knots (e.g., complex
140 knots that take these knots as constituents, or entirely separate patterns of tangles)
141 would be all the more challenging.

142 ***The present experiments: Evaluating the strength of knots and tangles***

143 The tightly controlled nature of this group of knots, combined with the estab-
144 lished hierarchy of their physical strength, makes them well suited to the present
145 research question and easy to adapt to a psychophysical paradigm. Here, we present
146 5 experiments examining people’s intuitions about the physical dynamics of knots.
147 Participants viewed images of these knots in various formats and presentation condi-
148 tions (including photographs of the physical knots, digital renders from simulations,
149 dynamic videos, and schematic diagrams) and were simply asked to evaluate their
150 relative strengths under forced-choice conditions.

151 If performance on intuitive physics tasks derives from a general-purpose physical
152 reasoning mechanism that approximates Newtonian physics (at least in naturalistic

153 settings), then we might expect participants to reliably select the stronger knots,
154 in line with their hierarchical organization. For example, reef knots should tend to
155 be judged as stronger than the other three knots in the series, grief knots should
156 be judged as weaker, and so on. However, if participants instead fail to appreciate
157 these differences in knot strength (despite their naturalistic presentation and con-
158 text), then this might reflect broader limits on physical reasoning. To foreshadow our
159 key results: Across all experiments and presentations, participants failed to produce
160 strength judgments consistent with Newtonian physics (Experiments 1–4), despite
161 demonstrating accurate visual and topological understanding of the knots they were
162 viewing (Experiment 5). Indeed, participants often gave *actively incorrect* rankings
163 of the knot hierarchy within a given experiment (such that the findings do not merely
164 reflect null results or chance performance). We suggest that these results put pres-
165 sure on general-purpose accounts of physical scene understanding, and place new
166 constraints on theories of how we reason about the physical world.

167 **Experiment 1: Naturalistic judgments of knot** 168 **strength**

169 Can naive human observers intuit the strength of visually similar but mechani-
170 cally dissimilar knots? Experiment 1 investigated this question as described above,
171 by evaluating whether observers could accurately judge the strength of reef, thief,
172 granny, and grief knots. Since previous failures in physical reasoning have been at-
173 tributed to contrived stimuli or a lack of context, we maximized naturalism in our

174 stimuli by simply taking photographs of real knots tied with nylon rope. (Later
175 experiments further enhance and probe both the naturalism and precision of this
176 setup.)

177 ***Method***

178 *Open Science Practices*

179 All data and materials supporting this experiment (and all others reported in this
180 paper) are available at <https://osf.io/xyq4h/>. This study was not preregistered.

181 *Participants*

182 50 participants were recruited online using Prolific and were compensated at an
183 average rate of \$10.50 per hour for their time. All participants were located in the
184 United States. One participant was excluded from analysis due to failed attention
185 checks (see below for more information).

186 *Stimuli*

187 Stimuli consisted of photographs of the reef, thief, granny, and grief knots (here-
188 after RTGG), tied (by author S.C.) using 4mm nylon rope. Each knot was tied
189 in three separate colorways (red/green, yellow/purple, and orange/blue) and pho-
190 tographed from two different perspectives (front and back views of the knot), result-
191 ing in 24 total images. Each knot was roughly pulled taut, and tied to maximize
192 visual similarity using the length of the bitter ends (the section of a rope that is tied
193 off) as a reference. Each knot was photographed lying flat against a dark background
194 and lit with neutral lighting. (In addition, two “catch” knots were created using a
195 similar method; see below for more detail.)

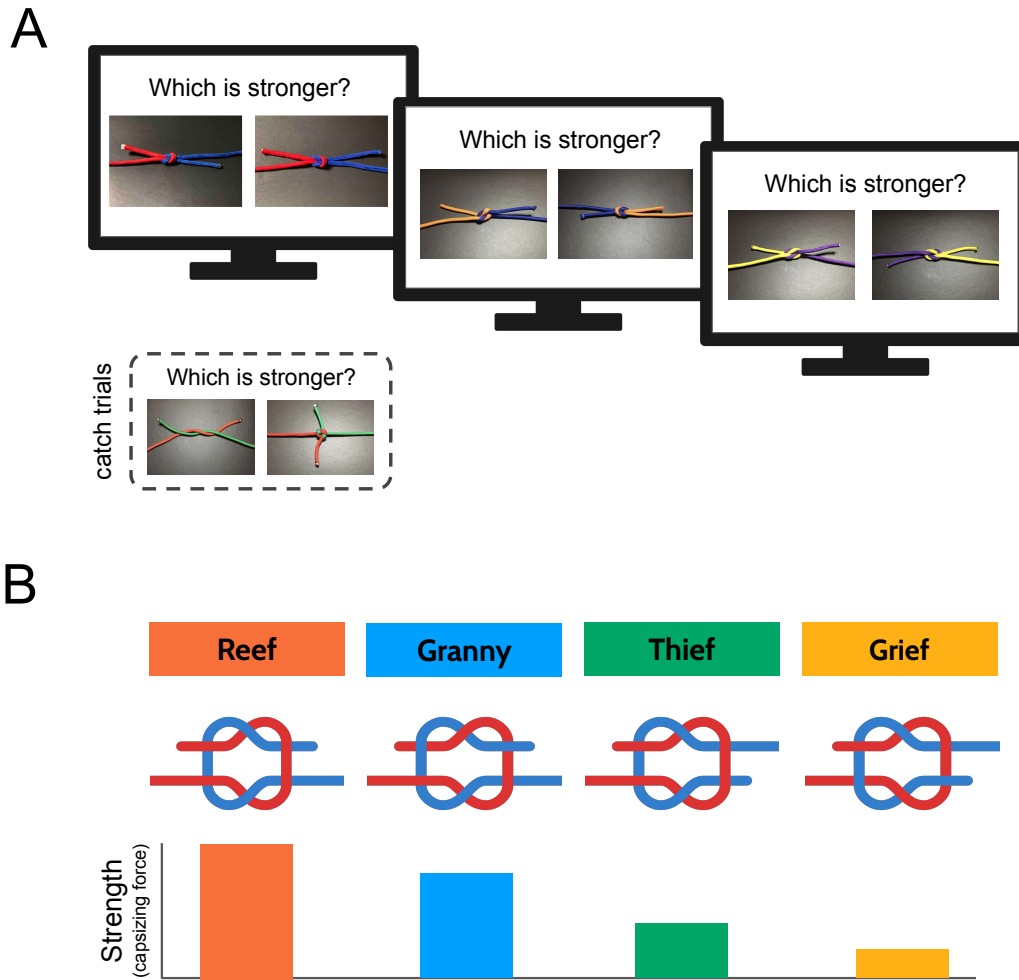


Figure 3: Design and predictions of Experiment 1. (a) Each monitor shows a sample trial of Experiment 1, which presents two knots on each trial. Participants simply answered which was stronger, using the criteria described in the main text and illustrated earlier in Figure 1. (Inset: Catch trials, which depicted a trivially easy strength contrast.) (b) Bar chart displaying the relative strengths of each knot in the RTGG knot series. If naive participants are sensitive to how the topological differences map onto differences in strength, then reef knots should be selected as the strongest in pairwise comparisons, grief knots least often, and so on for the other comparisons. Readers can experience this task for themselves at <https://perceptionresearch.org/knots>.

196 *Procedure*

197 Participants were told that their task was to evaluate knot strength, which was
198 defined (and visually depicted) as being unlikely to come undone if you were to pull
199 on the two long strands extending off-screen. (To ensure that these instructions were
200 clear, participants had to pass a practice trial in which a very secure knot appeared
201 next to loosely woven strings.) On each experimental trial, participants saw pho-
202 tographs of two knots at a time and were prompted to select the knot that appeared
203 to be stronger by clicking on it. Feedback was not given. Since every trial only
204 displayed two knots, each trial had either a correct or an incorrect answer, though
205 some trials showed knots with greater strength differences than others. Participants
206 saw every combination of the four knots possible, crossed with color and perspective
207 (either the front or back of the knot), totaling 144 experimental trials. Additionally,
208 four catch trials were dispersed at random through the task (these were the same
209 images as the practice trials), and later used as exclusion criteria. Finally, subjects
210 were also given a post-experiment survey in which they described any strategies they
211 used to complete the task.

212 Readers can experience this task for themselves at [https://perceptionresearch.](https://perceptionresearch.org/knots)
213 [org/knots](https://perceptionresearch.org/knots).

214 ***Results***

215 One participant failed to answer all catch trials correctly, and so was excluded
216 from further analysis, leaving 49 participants. (However, no result reported in this
217 paper depends on these sorts of exclusions; in other words, all significant findings
218 remain significant, in the same direction, even when no subjects are excluded at all.)

219 We evaluate performance by examining how often a given knot is chosen relative
220 to the others, across all trials. If intuitions about the relative stability of knots map
221 on to their ground truth relative stability, then we should see a pattern that looks
222 roughly like Figure 3B. Reef knots are the strongest of the four, so they should be
223 selected the most often during the experiment, followed by granny, thief and finally
224 grief knots, which are the weakest, and should rarely (if ever) be selected during the
225 experiment.

226 However, as can be seen in Figure 4A, performance did not at all capture this
227 hierarchy; in fact, performance was below chance. Participants selected the stronger
228 knot on only 42.1% of trials (where chance is 50%; $t(48) = 4.87, p < 0.001; d = 0.70$),
229 despite having demonstrated that they understood the instructions and correctly
230 answered the catch trials. Breaking this performance down by knot type: Reef knots
231 were chosen on 34% of the trials where they were shown (where chance is 50%), or on
232 17% of trials overall (where chance would be 25%). Granny knots were chosen 68%
233 of the time (34% overall), Thief knots 32% (16%), and Grief knots 67% (33.3%).
234 In other words, subjects showed little to no sensitivity to the large differences in
235 strength between these visually similar knots.

236 To appreciate this pattern more precisely, consider how judgments of Reef knots
237 (the strongest knots shown) compare to judgments of the other knots. First, Reef
238 knots were chosen at almost identical rates as Thief knots, despite being quite dif-
239 ferent in strength. These two knots differ only in the placement of the bitter ends
240 (Reef - same side; Thief - different side); even though this subtle difference has major
241 consequences for knot strength, subjects evidently did not appreciate these conse-

242 quences. Perhaps even more strikingly, however, Reef knots were consistently chosen
243 as weaker than Granny and Grief knots, despite being substantially stronger than
244 both of them. Indeed, Griefs (the weakest knot) were chosen 67% of the time they
245 were shown (i.e., 33.3% overall), compared to Reefs (the strongest knot), which were
246 chosen 34% of the time they were shown (i.e., 17% overall) — precisely the opposite
247 of their actual relationship.

248 Moreover, using a computational approach developed for computing dominance
249 hierarchies (e.g., the probability that competitor A beats competitor B, C, and so on)
250 from a series of pairwise competitions (Fujii et al., 2014), we can calculate a knot
251 rank hierarchy for each subject based on the outcomes of their pairwise strength
252 judgments. Of the included subjects, the most popular rank order was *granny* >
253 *grief* > *reef* > *thief* (33% of subjects), followed by *grief* > *granny* > *reef* >
254 *thief* (27% of subjects), and then *granny* > *grief* > *thief* > *reef* (12% of subjects).
255 (Notably, none of these rankings is correct, nor even particularly close.) Furthermore,
256 not a single subject expressed the correct rank order.

257 Furthermore, this poor overall performance did not reflect random or unsystem-
258 atic responding. To analyze the consistency of participants’ judgments, we assigned
259 each participant and each knot pair a “consistency score”, corresponding to the
260 proportion of trials where a participant picked the same knot in a given pairwise
261 comparison. For example, on trials where participants saw a Reef and a Grief knot
262 (24 trials total per participant), a participant who always answered Reef (i.e., 100%
263 accuracy) received a consistency score of 1, and a participant who always answered
264 Grief (i.e., 0% accuracy) also received a consistency score of 1. By contrast, a

265 participant who answered Reef on 50% of Reef-Grief trials and Grief on 50% of Reef-
266 Grief trials received a consistency score of 0 (with intermediate values calculated
267 according to the formula $consistencyScore = 2|proportionCorrect - 0.5|$). This
268 analysis revealed consistency scores well above 0 on all pairs, though consistency
269 was much lower for Reef-Thief (mean consistency score = 0.22) and Granny-Grief
270 pairs (mean consistency score = 0.23), which share most of their overall topology
271 and differ only in the position of a single strand. Consistency was much higher for
272 Reef-Granny (mean consistency score = 0.81), Reef-Grief (mean consistency score =
273 0.77), Granny-Thief (mean consistency score = 0.78), Thief-Grief (mean consistency
274 score = 0.79). Thus, even though participants showed that they could discriminate
275 between the knots (since they didn't simply pick each knot with the same frequency)
276 and understand what it means for a knot to be strong (since they passed the catch
277 trials), they failed to grasp the relationship between the visual appearance of the
278 knots and their strength. These results thereby provide initial evidence that knots
279 strain physical reasoning.

280 **Experiments 2–4: Increasing precision, richness and** 281 **naturalism**

282 Experiment 1 provided initial evidence that knots pose a challenge to physical
283 reasoning: When shown natural photographs of knots that vary greatly in strength,
284 subjects failed to distinguish strong knots from weak ones. However, as with the clas-
285 sical physical reasoning errors reviewed earlier, it is possible that poor performance

286 was driven by auxiliary factors that prevented subjects from accessing or demon-
287 strating subtler and more accurate physical knowledge. For example: (1) Although
288 the knots were hand-tied to maximize naturalism and ecological validity, this may
289 have come at the cost of (inadvertent) inconsistencies across colorways, perspectives,
290 and even knot type that may have biased strength evaluations; (2) As static images
291 taken from only one perspective (per image) and only two orientations (per knot),
292 the stimuli may have lacked the full context that would be available when viewing a
293 knot under real-world conditions (which permit dynamic sampling of different view-
294 points, double-checking key perspectives and angles, etc.), in ways that may matter
295 for engaging the operations of a mental physics engine; (3) It is unclear whether sub-
296 jects could even recover the topological structures of the knots, perhaps due to one
297 or more of the above-mentioned reasons, but perhaps due to the inherent difficulty
298 of extracting topological organization from images.

299 Experiments 2–4 addressed each of these weaknesses directly. To ensure that
300 the knots shown to subjects were accurate with respect to their physical properties,
301 Experiment 2 used digital renders from software specifically designed to simulate
302 knots under realistic physical conditions (including pulling force). To ensure that
303 subjects could leverage dynamic information from many viewpoints, Experiment 3
304 presented subjects with scrollable videos of the knots rotating 360° in space. And to
305 ensure that subjects had access to the underlying topology of each knot, Experiment
306 4 included schematic diagrams that make this topology explicit and unambiguous.
307 If subjects continue to fail to appreciate knot strength even under these very ac-

308 commodating conditions, this would be especially strong evidence that knots strain
309 physical reasoning.

310 *Methods*

311 All three experiments used a similar design to Experiment 1: A two-alternative
312 forced-choice task between members of the RTGG series evaluated for strength. Each
313 experiment recruited a new sample: Experiment 2 recruited 50 subjects to mirror
314 Experiment 1, and Experiments 3 and 4 recruited 100 subjects each to increase
315 statistical power. Of these, zero participants were excluded in Experiment 2 (for a
316 total of 50 subjects), 16 subjects were excluded in Experiment 3 (for a total of 84
317 subjects) and 4 participants were excluded in Experiment for a total of (96 subjects).
318 What differed primarily was the nature of the stimuli. Participants in each task were
319 compensated at an average rate of \$10.50 per hour for their time.

320 *Stimuli*

321 Experiment 2 depicted the same knot series as Experiment 1, but digitally ren-
322 dered in MATLAB using the procedure developed by [Patil et al. \(2020\)](#). The simu-
323 lated knots had a 4mm diameter, a bending modulus of 0.1 GPa, a Young’s modulus
324 of 1 GPa, a Poisson’s ratio of 0.3, and 15 N of pulling force. The simulation was run
325 to maximize visual similarity of the knots using the length of the bitter ends as a
326 reference. Each knot was rendered against a transparent white background.

327 Experiment 3 used hand-tied knots like Experiment 1; but rather than pho-
328 tographs showing static images of the front and back of each knot, participants
329 viewed interactive videos of each knot rotating 360°. All dynamic knot videos were

330 recorded using an iPhone 11 and converted into a sequence of 126 frames each using
331 kdenlive (<https://kdenlive.org/>). Each frame displayed a knot rotating along the
332 z axis until it completed a full 360° rotation, working out to about 3° of rotation per
333 frame. Participants could dynamically scroll through the video frames by dragging a
334 scroll bar under each video. The frame displayed for each knot corresponded to the
335 participant-initiated position of the scroll bar (i.e., if the scroll bar was in position
336 67, the 67th frame of the video would be shown). Participants could not advance to
337 the next trial without at least partially scrolling through both videos.

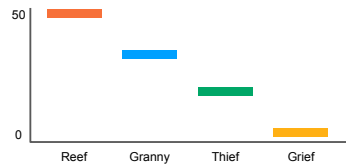
338 Experiment 4 used the same static photographs from Experiment 1, but with
339 the addition of schematic diagrams underneath each of the knot images. Each knot
340 schematic was adapted from public domain images, and altered to match the color-
341 ways depicted in the knot photographs. Arrows were also added to the longer ends
342 of each schematic to indicate the pulling direction participants should imagine when
343 evaluating its strength.

344 ***Results and Discussion***

345 All three experiments failed to reveal accurate evaluations of knot strength, with
346 performance at or below chance. (Note that the distinction between performing at
347 chance vs. below chance is not crucial for our purposes; what matters most is that
348 participants failed to perform *above* chance.)

349 In Experiment 2 (renders), overall performance was 44.8%, which was signifi-
350 cantly different than chance, $t(49) = 2.57, p < 0.05; d = 0.36$. Despite similarly poor
351 performance overall, the pattern differed from Experiment 1 with respect to the cho-
352 sen hierarchy of knots. For example, while subjects in Experiment 1 clearly chose

A Knot Evaluation: Accurate/Ideal Performance



B Knot Evaluation: Actual Performance

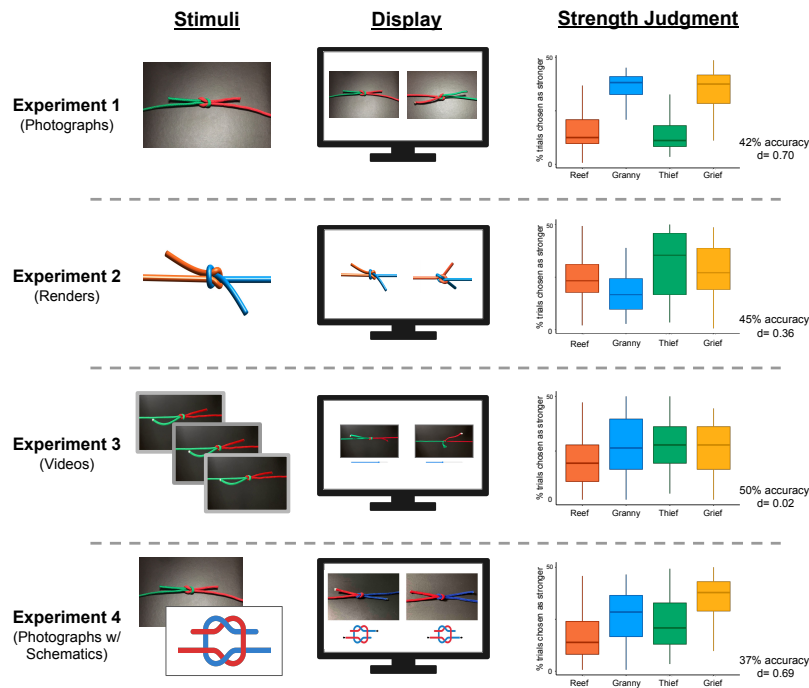


Figure 4: Results of Experiments 1–4. (a) ‘Accurate’ performance for the knot evaluation task. If subjects correctly represent knot strength (even subject to noise or error), the distribution of strength judgments should resemble the depicted ordering. Higher frequencies indicate that a knot won more pairwise comparisons throughout the experiment (i.e., was judged as stronger). (b) In fact, Experiments 1–4 show that participants fail to produce judgments consistent with ground-truth physics. Center line is the median, top and bottom of the boxes represent the interquartile range, and whiskers are minimum and maximum values excluding outliers. Importantly, responses were not merely random: As can be seen across experiments, responses were often quite consistent – just consistently *incorrect*. These results suggest that knots reliably strain physical reasoning.

353 Granny and Grief knots more often than Reef and Thief knots, in Experiment 2 this
354 pattern was more equivocal, though Thief and Grief knots were chosen marginally
355 more often than Granny and Reef knots. Despite these differences, subjects were
356 similarly consistent in their choices as in Experiment 1, with an average consistency
357 score of 0.62 across all pairwise comparisons. Mean consistency scores for each pair-
358 wise comparison were as follows: Reef-Grief: 0.57; Reef-Granny: 0.50; Reef-Thief:
359 0.59; Granny-Grief: 0.55; Granny-Thief: 0.73; Thief-Granny: 0.67.

360 In Experiment 3 (videos), overall performance was 49.6%, which was not signif-
361 icantly different than chance, $t(83) = 0.21, p = 0.83; d = 0.02$. Consistency scores
362 here averaged 0.55, with the following consistency scores for each pairwise compar-
363 ison: Reef-Grief: 0.64; Reef-Granny: 0.62; Reef-Thief: 0.37; Granny-Grief: 0.52;
364 Granny-Thief: 0.59; Thief-Granny: 0.66.

365 In Experiment 4 (schematics), performance was 36.9%, which was significantly
366 lower than chance, $t(95) = 6.76, p < 0.0001; d = 0.69$. The pattern of results mirrors
367 those of Experiment 1, with Grief knots and Granny knots being chosen as stronger
368 more consistently than Reef and Thief knots, despite the diagrams unambiguously
369 showing how the strands overlap. Participants showed an average consistency score
370 of 0.65. Across pairwise comparisons, the mean consistency scores were as follows:
371 Reef-Grief: 0.78; Reef-Granny: 0.70; Reef-Thief: 0.55; Granny-Grief: 0.52; Granny-
372 Thief: 0.66; Thief-Granny: 0.72.

373 In other words, all of these variations not only failed to reveal accurate physical
374 intuitions about knot strength, but in many cases also revealed *inaccurate* physical
375 intuitions. (For full rank-orders for all subjects and all experiments, see our data

376 archive.) These failures are all the more striking given that each experiment added
377 detail intended to give subjects every chance to evaluate the knots accurately (in-
378 cluding variations specifically inspired by complaints about previous intuitive physics
379 tasks), and also included catch trials that all included subjects answered correctly.
380 In other words, subjects understood their task, and demonstrated that they were ca-
381 pable of making at least some minimal evaluation of knot strength (albeit in a fairly
382 trivial case). These results thus continue to suggest that knots pose a particular
383 challenge to human physical reasoning.

384 **Experiment 5: Knot identification vs. knot evalua-** 385 **tion**

386 Experiments 1–4 provide evidence for striking failures in knot strength evaluation,
387 across many variations in presentation. However, it may still be that these results
388 do not reflect failures of physical understanding per se, but rather a more general
389 failure of visual cognition to extract the topology of the knots from the presented
390 images. In other words, perhaps errors reflect impoverished *inputs* to the physical
391 reasoning mechanism, rather than the operation of the physical reasoning mechanism
392 itself. This may be true even for Experiment 4, which presented schematic diagrams
393 alongside the knots; though our intention was that this additional information would
394 facilitate extraction of topology (and thereby enable accurate strength judgments),
395 perhaps these schematics simply failed to achieve this goal.

396 As a check on this possibility, Experiment 5 employed a similar design as Experi-
397 ment 4, but instead of making strength judgments, participants simply matched the
398 knot photographs to their corresponding schematic diagrams. Success in this task
399 is contingent on accurately representing the knots' topologies; so, if subjects can
400 perform well at this task, then failures in early experiments are unlikely to reflect
401 mere input constraints and instead likely to reflect deeper errors in physical scene
402 understanding.

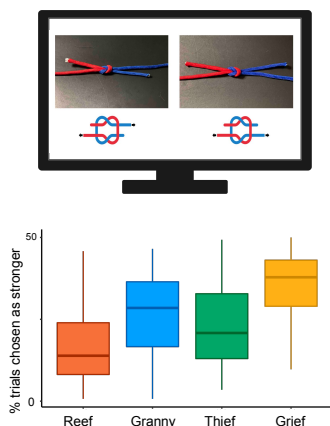
403 *Method*

404 This experiment used the same knot photos from Experiments 1 and 4, and the
405 same knot diagrams from Experiment 4. However, in the present task, participants
406 simply matched a photograph of a knot with its schematic diagram. On each trial, a
407 single knot photograph appeared, and beneath it were each of the four schematic dia-
408 grams (reef, thief, granny, and grief). Participants clicked on the schematic diagram
409 that they believed represented the knot.

410 To ensure that the task was clear, participants had to complete four practice
411 trials before they could proceed to the full experiment, where they matched different
412 versions of each knot in a colorway not shown during the full experiment. In the full
413 experiment, each knot (including front and back views) was displayed twice across
414 the same three colorways used earlier, for 48 test trials. In addition to these test
415 trials, randomly during the experiment participants also completed two catch trials
416 where, instead of a knot photograph appearing, a schematic diagram itself appeared,
417 such that one of the four options was just a copy of the central image; this was to
418 ensure that participants were looking at each diagram closely.

Evaluation

Experiment 4
(which knot is stronger?)



Identification

Experiment 5
(which knot is this?)

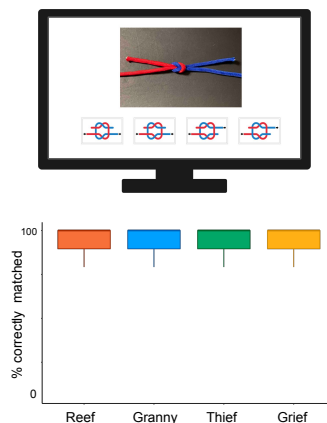


Figure 5: Results of Experiment 5. Whereas evaluations (left; Experiment 4) of knot strength showed striking inaccuracies (failing to match ground-truth physics), knot identification (right; Experiment 5) showed striking *accuracy*, with performance near ceiling. In other words, participants *were* able to tell what kind of knot they were viewing (where such discriminations require parsing finer details of the knots); they were just unable to translate that understanding into accurate evaluations of knot strength – in line with our hypothesis that knots are challenging to reason about physically (even when participants can accurately represent their underlying topology).

419 *Results and Discussion*

420 In principle, this task might have set up participants for worse performance than
421 previous experiments, since the odds of a correct guess on any trial was 1 in 4 rather
422 than 1 in 2. However, performance in this task was exceptional, and indeed even
423 close to ceiling: 92.5% (where chance is 25%), $t(78) = 44.34, p < 0.0001; d = 4.99$.
424 (And even this high average perhaps undersells participants' performance, due to the
425 skewness of this measure; for example, 68% of participants scored above 95%.)

426 This result suggests that observers can extract the topological properties of these
427 knots after all — or, at least, those details that distinguish the knots from one
428 another. And so the failure to do so is unlikely to be the explanation of poor per-
429 formance in Experiments 1-4. Put differently: Participants *were* able to grasp the
430 topological properties of the knots; what they were unable to do was derive from that
431 understanding an accurate sense of the physics that such topology entails. (Of course,
432 participants were not literally perfect; but occasional errors are not a sufficient ex-
433 planation of the results of Experiments 1-4.) The strongest remaining explanation,
434 then, is that human physical reasoning truly is strained by knot-like stimuli.

435 **General Discussion**

436 Whereas recent work documents surprisingly accurate intuitions about a variety
437 of physical phenomena — and uses these successes to posit a general-purpose physi-
438 cal reasoning mechanism — here we have explored a new class of visual stimuli and
439 phenomena that strains physical understanding. Across four experiments, human
440 observers failed to discern even very large differences in the strength of simple knots.
441 Importantly, the errors observed here persisted despite several additions and modifi-
442 cations to the stimuli and task intended to draw out the knots' mechanical properties.
443 These variations include: Naturalistic photographs (Experiment 1), digital renders
444 from physically precise simulations (Experiment 2), dynamic videos (Experiment 3),
445 and schematic diagrams (Experiment 4). Additionally, these failures were not simply
446 due to an inability to visually extract the topological structure of the knots, since
447 performance was near ceiling in a task that required matching photographs of the

448 knots to their respective schematic diagrams (Experiment 5). In other words, par-
449 ticipants were able to discern the structural and topological properties of the knots;
450 what they failed to understand was how this structure translates into corresponding
451 physical and mechanical properties. Moreover, participants were not merely guessing
452 randomly in making their judgments, since many experiments revealed systematic
453 patterns in responding (just not patterns that tracked with the actual strength of
454 the knots). Overall, then, these experiments provide evidence that knots pose a chal-
455 lenge to physical reasoning; and by extension, they place constraints on theorizing
456 about physical scene understanding and the mechanisms underlying it.

457 It is worth being clearer about the nature and significance of these constraints;
458 what implications do these results have for broader theorizing about general-purpose
459 physical reasoning mechanisms? Though there can, in principle, be many general-
460 purpose accounts of physical reasoning, one especially popular theory in recent years
461 is the Intuitive Physics Engine (IPE) hypothesis (for a review, see [Ullman et al.,](#)
462 [2017](#); for an earlier presentation of the core idea, see [Battaglia et al., 2013](#)). This
463 account extrapolates from success in certain domains of physical reasoning — such
464 as judging the stability of a tower of blocks, the behavior of connected gears and
465 pulleys, or the flow of a liquid around obstacles (as in Figure 2A and 2B) — to a
466 general-purpose physical simulation device in the mind. This hypothesized device
467 models the physics of the world (and the objects within it) according to Newtonian
468 laws and principles, with terms for mass, gravity, friction, and other relevant physical
469 parameters; performance on a given physical reasoning task is thus thought to reflect
470 the output of this device and its simulations. Although the IPE is hypothesized to be

471 “noisy” and probabilistic — only approximating scenes and their physics, subject to
472 uncertainty ([Battaglia et al., 2013](#); [Sanborn et al., 2013](#)) — it is nevertheless thought
473 to be sufficient for most commonsense visual judgments.

474 Though our interests here go beyond any particular instance or variation of this
475 hypothesis, the IPE is a useful vehicle for understanding how domain-general physical
476 reasoning might be carried out by the mind – and so is correspondingly useful for
477 thinking through the implications of the present results.

478 If physical reasoning indeed reflects a domain-general process that models the
479 world according to principles of Newtonian mechanics, then a natural question arises
480 as to why participants consistently failed to appreciate the strength of knots in our
481 tasks. Under the IPE hypothesis, for example, failures in physical reasoning are
482 typically thought to emerge when the stimulus is impoverished or presented without
483 sufficient context (e.g., line diagrams rather than naturalistic images or videos), or the
484 task or physical scenario is unnatural or unfamiliar (e.g., tracing the trajectory of an
485 object exiting a spiral tube; [Battaglia et al., 2013](#); [Kubricht et al., 2017](#)). While these
486 factors certainly seem relevant for explaining poor performance in other intuitive
487 physics tasks, it is not clear that they straightforwardly account for the failures we
488 observe here in Experiments 1-4. The stimuli used in our experiments were shown
489 in a variety of presentations designed to maximize both visual context and realism,
490 and Experiment 5 revealed that participants could correctly parse the layout of each
491 knot based on static images. This indicates that the stimuli themselves contained the
492 information that governs differences in their strength, and that participants could

493 access that information in other contexts. What they failed to do, consistently, was
494 translate that information into accurate knowledge of knot strength.

495 *The role of familiarity and experience*

496 A more open question, perhaps, is how ‘familiar’ or ‘natural’ knots are as a
497 stimulus class, and indeed whether one should expect a domain-general physical
498 reasoning mechanism (whether the IPE or any other mechanism) to apply to them
499 in the first place.

500 One concern along these lines is that knots may just seem like an overly spe-
501 cialized domain — a skill of interest to sailors and rock climbers but not ordinary
502 people. However, as discussed previously, knots are actually quite pervasive, cer-
503 tainly in contemporary life (tying shoes, untying tangled headphone cords, etc.),
504 across cultures and time periods (where they have been used for millennia for prac-
505 tical, ritualistic, and decorative purposes; [d’Errico et al., 2018](#); [Leroi-Gourhan, 1982](#);
506 [Turner and van de Griend, 1996](#)), and even in the practices of other species ([Hardy
507 et al., 2020](#); [Herzfeld and Lestel, 2005](#)). Though it is admittedly unclear just how
508 familiar a stimulus must be in order to fall within the purview of a given physical
509 reasoning mechanism (at least under current frameworks), we note that knots seem
510 no *less* familiar than other stimuli that elicit accurate physical intuitions. For ex-
511 ample, previous work has shown that naive subjects succeed at tasks that require
512 them to anticipate the behavior of interlocking gears or systems of connected pul-
513 leys ([Hegarty, 2004](#)). It strikes us that, if naive subjects succeed at *those* (rather
514 unfamiliar) tasks, then unfamiliarity per se may not be a reason to predict failure on
515 knots. (Ask yourself: When was the last time you hoisted an object using a system

516 of interconnected pulleys? And when was the last time you tied your shoes?) And
517 even if our participants were unfamiliar with the specific knots used in our task,
518 these knots are actually *less* complicated than the already rather simple shoelace
519 knot (which in fact contains the reef knot studied in our experiments).

520 Another way in which knots may be distinct from other kinds of physical stimuli
521 we encounter is that they often represent a form of “received wisdom”; some consider-
522 able portion of any individual’s knowledge about knots often comes from instruction,
523 beyond what they may learn from intuitive self-discovery or observation in nature.
524 This aspect of knots raises questions both about the bounds of physical reasoning as
525 well as the role of experience in parsing knots and evaluating their strength. For ex-
526 ample, it is quite plausible that expert sailors or rock-climbers might succeed where
527 our naive participants failed, owing to their expertise in recognizing and evaluating
528 knots. However, from our perspective this observation only *strengthens* the impli-
529 cations our results have for theories of intuitive physical reasoning. The fact (if it
530 is a fact) that expertise is required to correctly evaluate the strength of knots and
531 tangles only further testifies to their counterintuitive nature; by contrast, no similar
532 training or expertise seems needed to predict the behavior of interlocking gears or the
533 path of a flowing liquid around various barriers (Bates et al., 2019; Hegarty, 2004).
534 This suggests all the more that knots do not belong to the same class of phenomena
535 that humans can readily and accurately reason about — in line with our interest in
536 them as a case study of everyday physical phenomena that fall outside the scope of
537 domain-general physical reasoning capacities. To put the point another way: While
538 expertise would surely be required to reason correctly about electromagnetism or

539 quantum physics, knots are decidedly *unlike* those systems: Knots are *not* somehow
540 more complicated or obscure than many of the physical stimuli and systems that
541 have been shown to elicit successful reasoning, and yet they nevertheless strain our
542 physical intuitions.

543 *Rigid-body physics vs Soft-body physics*

544 Another possibility underlying failure in this task is that domain-general physical
545 reasoning may be optimized for (or restricted to) rigid-body objects, and that phys-
546 ical reasoning is strained when making predictions about the kinds of soft, flexible
547 materials knots are typically composed of. For example, if human physical reasoning
548 works similarly to a physics engine — perhaps one that prioritizes speed and general-
549 ity over precision and accuracy — then one might predict difficulties with soft-body
550 objects, as simulating their physical properties is thought to be more computation-
551 ally demanding than simulating the behavior of simpler geometric rigid-body objects
552 such as stacks of blocks (Ullman et al., 2017). Indeed, realistically simulating knots
553 and ropes has long been a challenge in computer graphics (including in the gaming
554 industry), with various computational techniques developed to approximate different
555 properties. For example, Jakobsen (2001) describes a method in which rope can be
556 simulated in a simple 2D environment by creating a set of particles whose positions
557 are updated to mimic deformations due to gravity and tension, and Phillips et al.
558 (2002) introduce an alternative method where ropes are instead represented as splines
559 of linear springs, and knots can be formed in 3D space by tracking collisions of the
560 rope with itself. A particularly detailed simulator developed by Brown et al. (2004)
561 allows users to manipulate rope in real time and construct knots by modeling rope in-

562 stead as a cylinder that deforms and stretches over physically-motivated constraints.
563 Each of these simulation approaches trades off some degree of realism and accuracy
564 for speed or computational efficiency; it is possible that similar tradeoffs arise in
565 human physical reasoning (perhaps depending on the particular task at hand). That
566 said, it seems unlikely that poor performance in our task could be solely attributed
567 to the non-rigid nature of our stimuli, if only because observers have been shown to
568 make rather accurate predictions and judgments about other non-rigid or soft-body
569 stimuli. Such cases include cloth draped over an object (Wong et al., 2023; Yildirim
570 et al., 2024), liquid pouring into containers (Bates et al., 2019; Kubricht et al., 2016)
571 and elastic objects (Paulun and Fleming, 2020; see also Little and Firestone, 2021).
572 Under current models, it is unclear why observers succeed in these contexts yet fail
573 when asked to judge relative differences in strength between knots. Further research
574 adopting the game-engine approach might shed light on the specific computational
575 constraints of simulation in physical reasoning in a way that accounts for failure to
576 judge the strength of knots while preserving success in other tasks involving soft,
577 flexible materials.

578 *Heterogeneity in physical reasoning*

579 If the above explanations are insufficient, then why did our subjects fail? One
580 possibility is that physical reasoning mechanisms are simply more heterogeneous than
581 a pure simulation-based account would imply, and that the mind employs different
582 physical reasoning strategies depending on stimuli and task demands (see, e.g., Smith
583 et al., 2023). It could even be the case that knots and tangles belong to a special
584 class of objects or systems that cannot be processed by a domain-general physical

585 reasoning mechanism. On this interpretation, when simulation fails (due to compu-
586 tational complexity, resource constraints, or other reasons), subjects may be using
587 heuristics to evaluate the strength of knots, and these heuristics may simply fail to
588 track with knot strength (in at least the present scenarios). Importantly, heuristics
589 may account for the patterns of responses here even though the knots most favored
590 by subjects varied by experiment. For example, if the heuristics subjects used were
591 based (even in part) on some factor that was not systematically varied or measured
592 across experiments — such as, e.g., how tightly wound a knot appeared, whether
593 there was a visible gap between any part of the knot and any other, or even more
594 incidental factors such as how it rested on the surface where it was photographed
595 — then responses that seem unsystematic *with respect to knot type* could still arise
596 from heuristic reasoning. An open question remains as to just how much of phys-
597 ical reasoning is captured by one or the other approach (simulation vs. heuristics)
598 — an issue raised by recent critiques of general-purpose simulation as the primary
599 driver of physical predictions (e.g., [Marcus and Davis, 2013](#); [Ludwin-Peery et al.,](#)
600 [2021](#); though see [Bass et al., 2021](#)). Our work here is agnostic about these broader
601 challenges, though it is certainly possible to see the present failures in this more
602 skeptical light.

603 Beyond considerations about the class of stimuli knots may (or may not) belong
604 to, it is also possible that the type of physical judgment used in this task may
605 be beyond the scope of intuitive physical reasoning. While we may quickly and
606 accurately make judgments about properties such as weight, center of mass and
607 projectile motion, perhaps judgments about strength (or at least how much pressure

608 a knot can withstand without capsizing) recruit separate reasoning mechanisms. It
609 has already been demonstrated that, even within the same class of stimuli, physical
610 judgments can converge or diverge with Newtonian predictions. For example, while
611 participants fail to correctly draw the trajectory of a ball on a pendulum once the
612 string has been cut, they can correctly guess its landing location (Smith et al., 2018).
613 This result has been taken to suggest that prediction and explanation of physical
614 scenes may rely on separate mechanisms; the former reflective of a veridical domain
615 general physical world model and the latter heavily biased and prone to error.

616 These experiments also open the door to further questions about how people rep-
617 resent and reason about knots. Outside of the challenge they pose to general-purpose
618 theories of physical intuitions, knots have often been seen as having significant (but
619 mostly unrealized) promise to explore physical reasoning more broadly (Santos et al.,
620 2019). For example, even though subjects in our studies struggled to evaluate knot
621 strength, it seems likely that this ability could be acquired through practice and
622 study (and may be present in knot “experts” such as scouts or sailors). In that
623 case, knots could serve as a testbed for physics “training” — the ability to acquire
624 new physical knowledge that is initially unintuitive. There may also be other knot-
625 related tasks that are easier (or harder) for subjects, such as evaluating whether a
626 given configuration of string would or would not become a knot when pulled taut, or
627 even simply estimating how much string is required to make a given knot (see Figure
628 6).

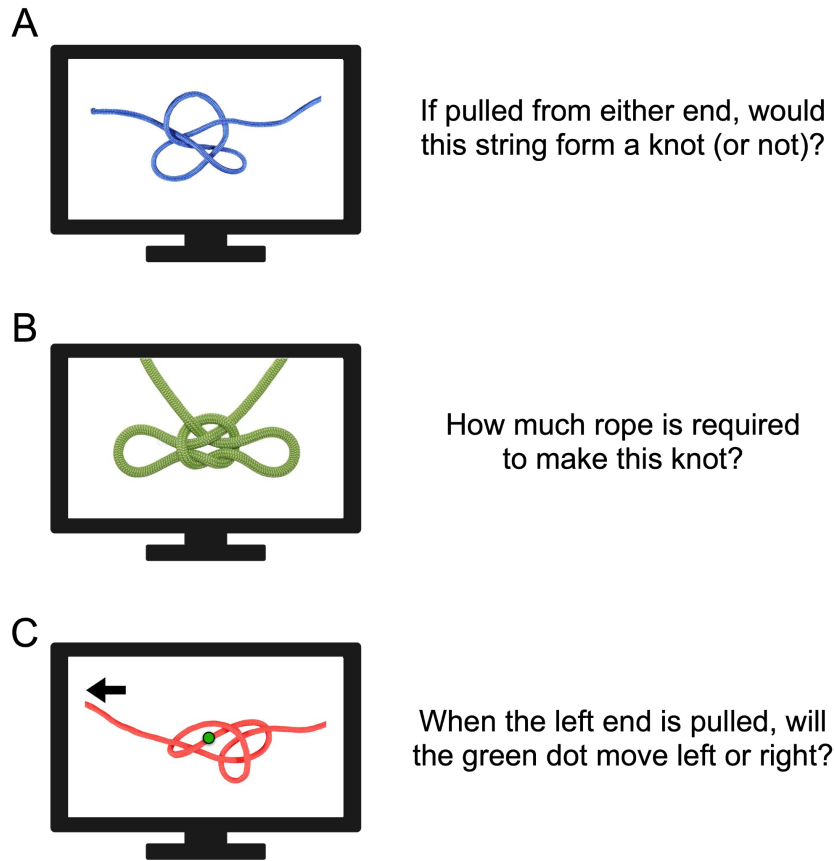


Figure 6: Other tasks exploring intuitive judgments of knots. (a) How easily can naive participants tell when a tangle of string will form a knot. (b) Can we ‘mentally unravel’ bound knots to determine how much string was used to make them? (c) A future set of experiments could ask about following elements of a knot as it is loosened or tightened (cf. [Hegarty 2004](#)).

629 Conclusion

630 Physical judgments about the environment are often reliable and robust; but
 631 the breadth and depth of physical knowledge may still be both under-examined and
 632 under-specified. While relatively unexplored in the domain of intuitive physics, knots
 633 provide useful insight into the nature of physical scene understanding — posing a

634 challenge both to reasoners about knots and perhaps even to theories of physical
635 reasoning.

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737 Appendix A

738 One potential concern with the design of our studies is the use of “reef”, “thief”,
739 “granny” and “grief” knots as *bends* (knots that join two pieces of thread) rather
740 than binding knots (a knot made of just one thread, tied to itself, that may be
741 used to keep a single object or multiple loose objects securely fastened, see Figure
742 7A). As discussed in our main text, the topological mechanics of this knot series
743 has been validated by both optomechanical experiments and computer simulations
744 (Patil et al., 2020, see) even when they are used as bends. However, received wisdom
745 from communities that use these knots (e.g., sailors, rock climbers) sometimes holds
746 that even the strongest of these knots is too weak to justify most practical uses. For
747 example, Clifford Ashley, author of an important manual discussed in our text, goes
748 as far to claim that the misuse of reef knots as bends has caused “more death and
749 injury than all other knots combined” (Ashley, 1944, pg. 18). The scenarios he has
750 in mind are likely cases where someone has used a reef knot to secure a boat to a
751 dock or to hoist a heavy object into the air.

752 Despite not being recommended for such sensitive and high-stakes uses, we chose
753 bends for our physical reasoning experiments because both conceptualizing and eval-
754 uating their strength is relatively simple (one only needs to consider the pulling
755 forces as well as the implied friction from the strings once tied around each other)
756 and because their strength has been validated in previous work (Patil et al. 2020 —
757 which, again, assesses these knots as bends). Finally, we thought that bends better
758 lent themselves to motor simulation processes, since the task given to subjects is to

759 predict what would happen if they physically pulled on the loose ends of each knot
760 (see [Schwartz and Black, 1999](#), and also discussion in ?, who propose that “first-
761 person” or user-oriented tasks produce better physical judgments than third-person
762 problem solving). By contrast, the force applied to binding knots comes from the
763 bound object, rather than a pulling force of the sort that a person could apply.

764 However, to be sure that the results of our main experiments aren’t due to their
765 presentation as bends, we re-ran Experiment 1 with images of the four knots used
766 as binding knots instead. As noted above, binding knots are typically used to fasten
767 objects; a common maritime application, for example, is to keep a sheet of sail rolled
768 up tightly. Importantly, the communities that rely on these knots still consider
769 the same hierarchy to apply (with *reef* > *granny* > *thief* > *grief*). However,
770 to our knowledge this hierarchy has not been physically validated in the same way
771 as it has for bends ([Patil et al., 2020](#)). We thus include this experiment only in
772 this Appendix, because it lacks the kind of ground-truth baseline available for the
773 experiments included in our main text.

774 ***Method***

775 *Stimuli*

776 50 participants were recruited online using Prolific. Each participant was compen-
777 sated monetarily for their participation. One participant was removed from analysis
778 due to a server error in recording their data. None of the participants failed any of
779 the catch trials.

780 *Procedure*

781 This experiment used the same procedure as Experiments 1-4: A two-alternative
782 forced-choice task between members of the RTGG knot series evaluated for strength.
783 Rather than instructing participants to imagine pulling on either end of the knot,
784 participants were instead asked to infer which knot would be “least likely to let the
785 paper towels unravel”, or “more likely to keep the paper towel bound up”.

786 *Stimuli*

787 The same four knots from Experiments 1-5 were depicted as binding knots instead
788 of bends. For this stimulus set we used 4mm nylon rope that was tie-dyed so that
789 participants could easily parse how the rope overlapped with itself. Each knot was
790 pulled roughly taut around a bundle of paper towel lying on a black surface, and
791 tied to maximize visual similarity using the length of the bitter ends (the section of
792 a rope that is tied off) as a reference. There were three separate tie-dye colorways
793 (pink/yellow, purple/pink, and green/purple) for each knot, resulting in 12 total
794 images.

795 ***Results and Discussion***

796 This experiment, despite depicting the knots as binding knots instead of bends,
797 yielded very similar results to Experiment 1 (Figure 7C). Overall performance was
798 46.8% which was not significantly different from chance, $t(48) = 1.53, p = 0.132; d =$
799 0.219. This result suggests that the failure to intuit the strength of these knots, as
800 extensively explored and documented in our main text, generalizes to other presen-
801 tations and does not depend on their depiction as bends.

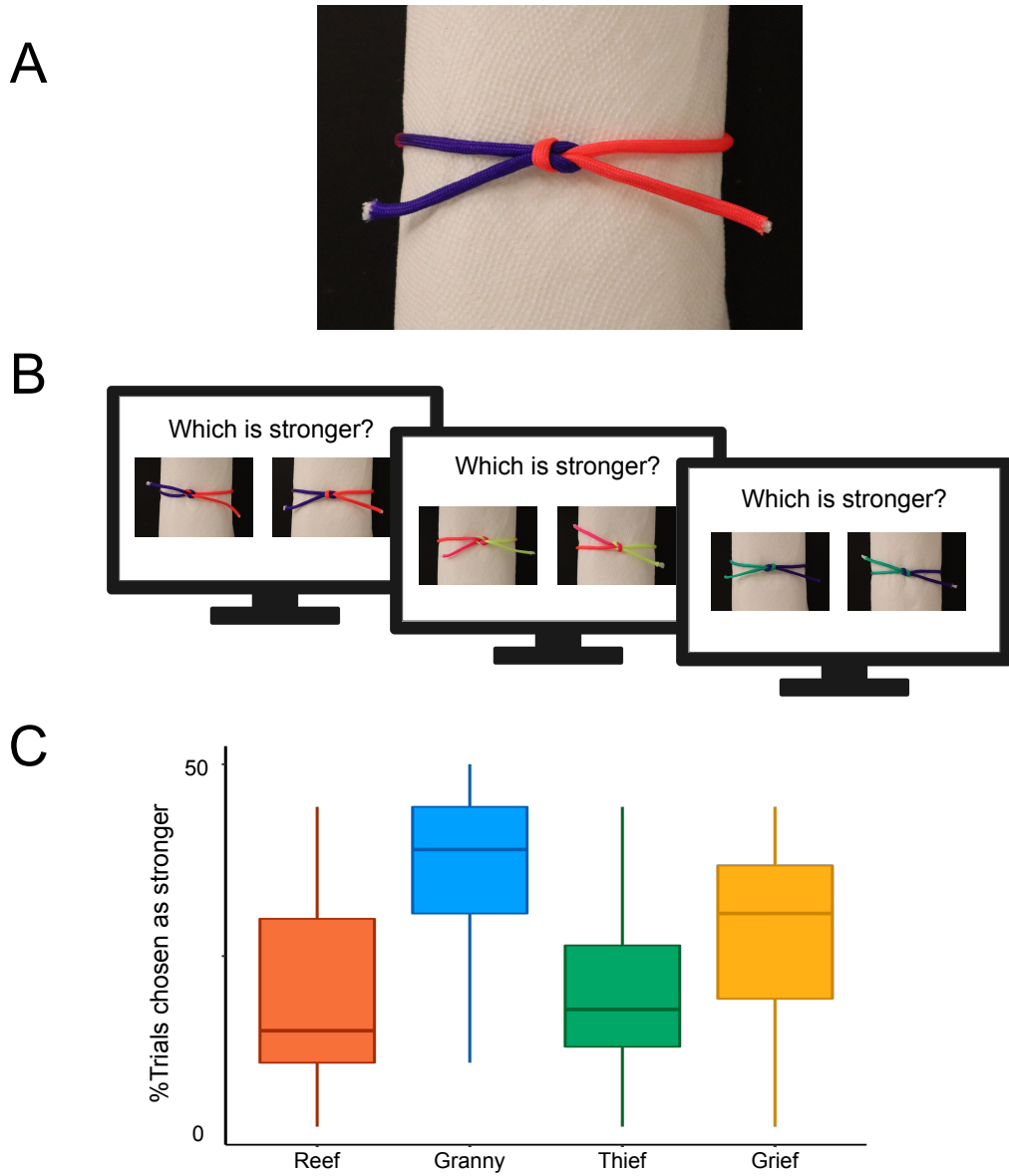


Figure 7: Experiment 6. (a) A sample binding knot used as stimuli. (b). Each monitor shows a sample trial of Experiment 6, which presents two knots on each trial. Participants simply answered which was stronger. (c) Each boxplot represents the distribution of frequencies a knot was chosen as stronger by participants.