

Regularity in Similarity Judgments

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Similarity

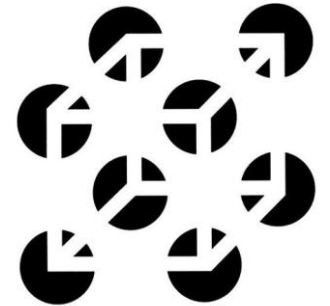
- Similarity is a central concept in a variety judgment and decision-making domains
 - Choice (Tversky, 1972)
 - Probability (Bar-Hillel, 1974)
 - Decision time (Bhatia and Mullett, 2018)
- Large proportion of similarity research has focused on object-to-object, 1-to-1 similarity judgments, not multiple object comparisons.

Hypothesis

- In current research, we focus our attention on how similarity judgments are made among groups of stimuli.
- We believe that, when judging the similarity of a group of objects, the relationships between these objects may play a strong role in how similar the objects seem to each other.
- **Hypothesis: Regularity in differences among the objects will increase similarity judgements of the group when judged as a whole.**

Regularity

- Similarity
 - Strongly dependent on the **psychological representations formed by individuals**
 - (Hahn, 2014; Tversky, 1977; Hahn, Chater, and Richardson, 2003; Shepard, 1962)
- Gestalt psychology
 - Wertheimer (1923)
 - Our mind imposes structure on individual pieces of information we are exposed to, and our perception is inherently affected by this.
- If our brains automatically impose structure on the pieces of information, and it is these mental representations that forms the basis of similarity judgments, **how is structure imposed?**

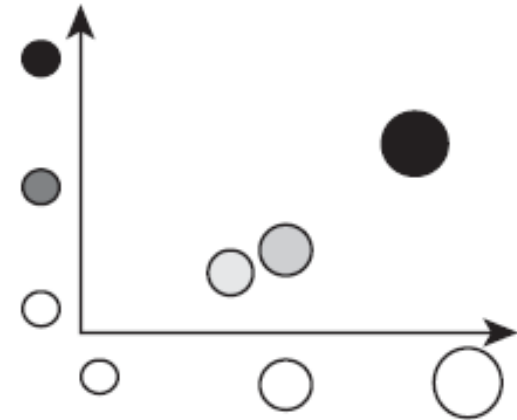


Regularity reduces complexity

- Optimal structure would compress information as much as possible while still maintaining the fundamental characteristics.
- One way to quantify the complexity of a set of information being **Kolmogorov complexity** (Kolmogorov, 1965; Li and Vitanyi, 1997).
- Kolmogorov argued that a way to quantify the complexity of a piece of information is the length of the shortest algorithm necessary to describe it.
 - (1, 3, 5, 7, 9, 11, 13) vs (2, 9, 11, 4)
- Thus, any form of regularity, greatly reduces the complexity, making the information easier to encode (Chater, 1999).
- **Groups that are easier to simplify into a short “algorithm”, that have lower Kolmogorov complexity, will be judged as more similar as compared to groups that do not exhibit this simplicity.**

Hypothesis vs. Spatial Model

- Geometric-distance model (Shepard, 1962)
 - Items we judge are mapped out on a psychological map
 - The longer the distance the items are on this map, the more dissimilar it seems.
 - [2, 2, 0, 6, 4] vs. [7, 14, 21, 28, 35]



Hypothesis vs. Feature-based Model

- Feature Model (Tversky, 1977)
 - Similarity is a function of (shared features)/(distinct features)
 - [2, 2, 0, 6, 4] vs. [7, 14, 21, 28, 35]

$$S(\mathbf{a}, \mathbf{b}) = \frac{f(\mathbf{A} \cap \mathbf{B})}{f(\mathbf{A} \cap \mathbf{B}) + \alpha f(\mathbf{A} - \mathbf{B}) + \beta f(\mathbf{B} - \mathbf{A})},$$

$\alpha, \beta \geq 0,$

Hypothesis vs. Transformation Model

- Transformation model (Hahn, Chater, and Richardson, 2003)
 - More transformations = less similar
 - [2, 2, 0, 6, 4] has 3 number transformations, 10 value transformations
 - [7, 14, 21, 28, 35] has 4 number transformations and, 28 value transformations

Studies

Study 1a: Numbers-within

Study 1b: Numbers-between (MTurk)

Study 1c: Numbers-between (Undergraduate)

Study 1d: Shapes-within

Study 2: Numbers Regularity Factor

Study 1a: Numbers-within

Conditions (N = 50; 56 strings of number)

- 1) All same: [1, 1, 1, 1, 1], [4, 4, 4, 4, 4]
- 2) **Regular Interval: [2, 4, 6, 8, 10], [6, 5, 4, 3, 2], [1, 3, 5, 7, 9]**
- 3) Two Same Three Different: [2, 6, 0, 0, 8], [2, 2, 4, 8, 9], [0, 1, 4, 4, 9]
- 4) Irregular Monotonic: [0, 1, 3, 6, 9], [1, 2, 5, 6, 8], [3, 5, 7, 8, 9]
- 5) Irregular Interval: [6, 4, 8, 7, 1], [0, 6, 4, 8, 1], [7, 0, 5, 1, 9]

 **Simple**

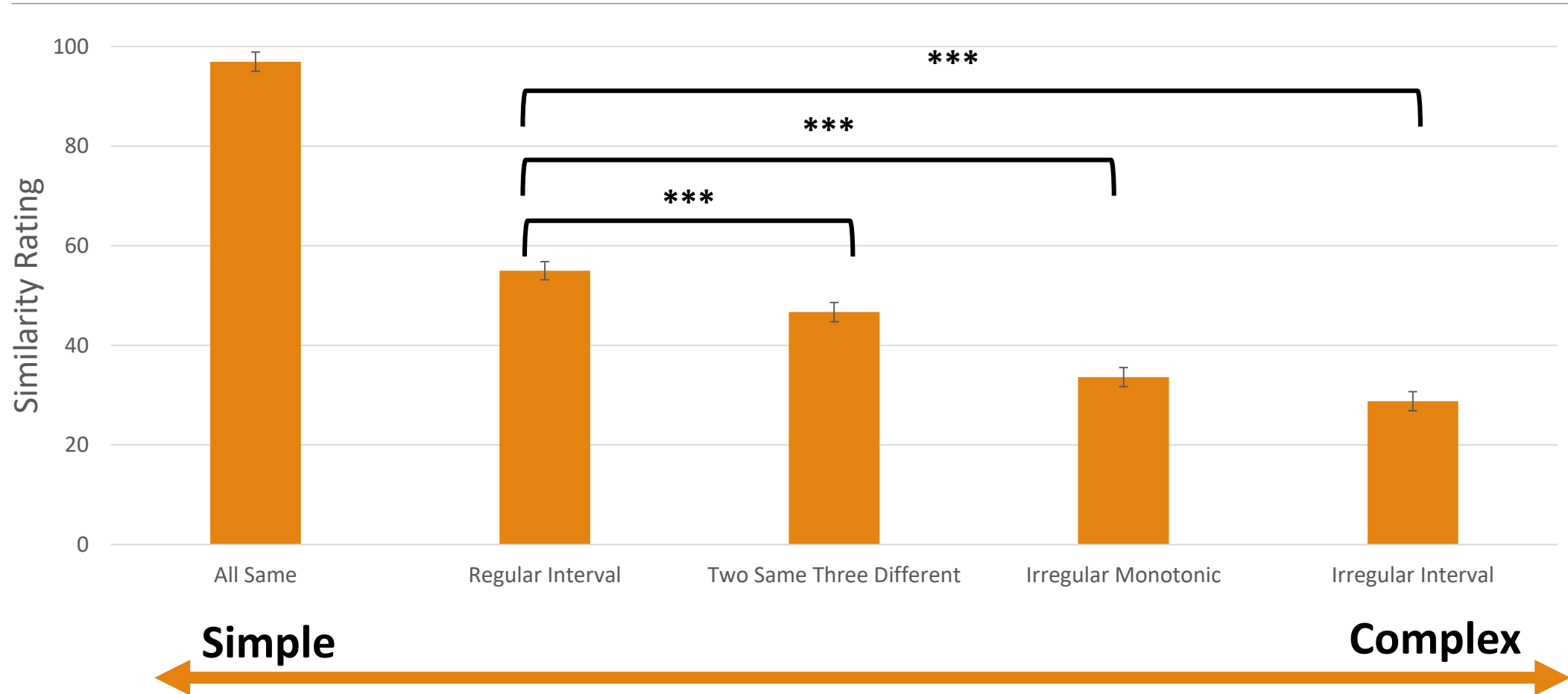
 **Complex**

[2, 4, 6, 8, 10]

“How similar are these numbers to each other?”

0-100 scale, 0 = *Extremely dissimilar*, 100 = *Extremely similar*

Study 1a: Results

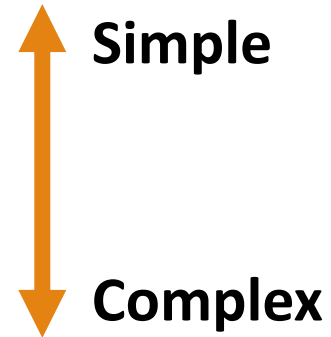


* $p < .05$, ** $p < .01$, *** $p < .001$

Study 1b: Numbers-between (Mturk)

Conditions (N = 185)

- 1) **Regular Interval:** [3, 6, 9, 12, 15]
- 2) Three same Two different: [15, 2, 4, 15, 15]
- 3) Two same Three different: [15, 2, 4, 7, 15]
- 4) Irregular Interval: [8, 2, 4, 7, 15]

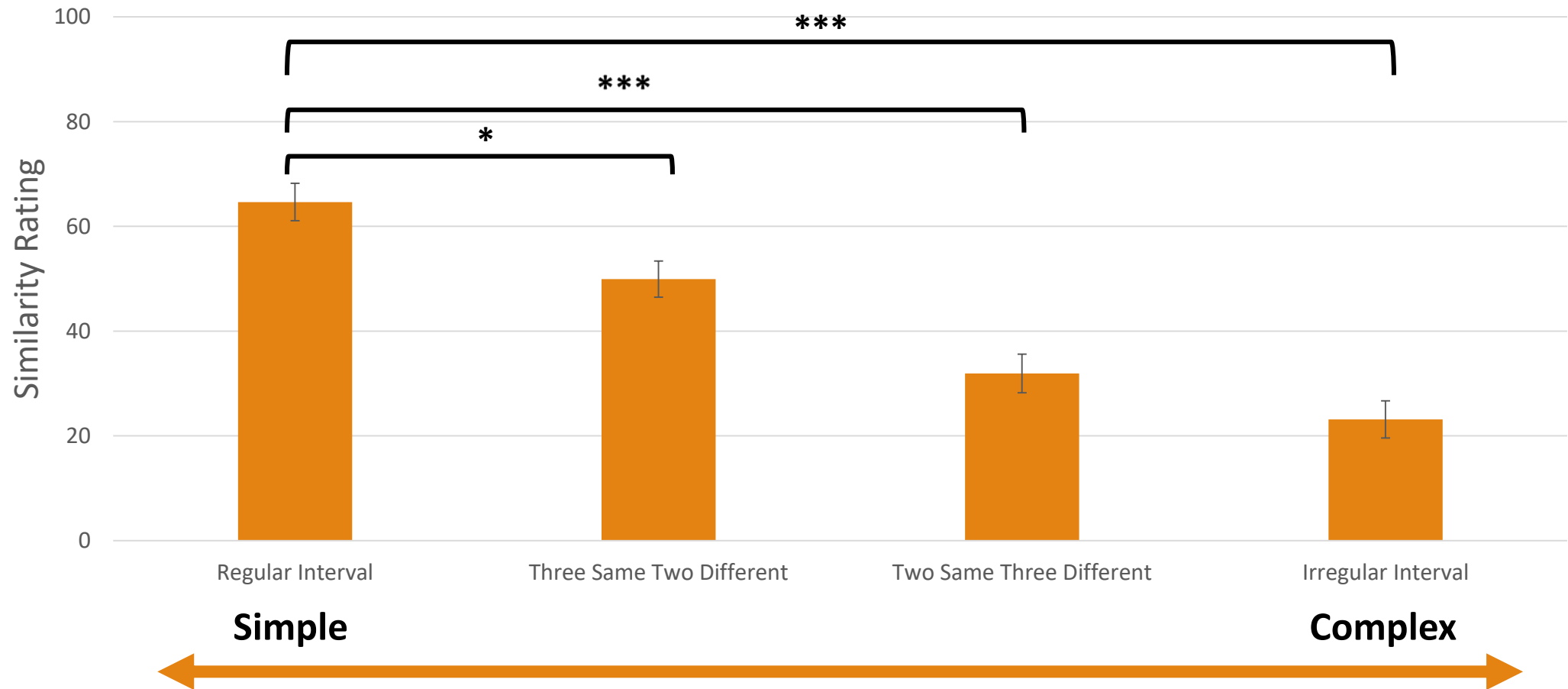


[3, 6, 9, 12, 15]

“How similar are these numbers to each other?”

0-100 scale, 0 = *Extremely dissimilar*, 100 = *Extremely similar*

Study 1b: Results

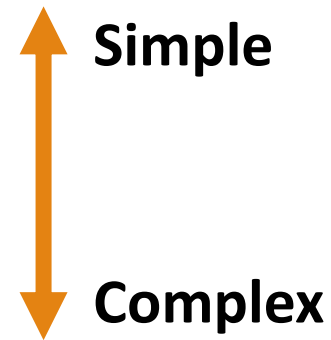


* $p < .05$, ** $p < .01$, *** $p < .001$

Study 1c: Numbers-between (Undergraduates)

Conditions (N = 165)

- 1) **Regular Interval:** [3, 6, 9, 12, 15]
- 2) Three same Two different: [15, 2, 4, 15, 15]
- 3) Two same Three different: [15, 2, 4, 7, 15]
- 4) Irregular Interval: [8, 2, 4, 7, 15]

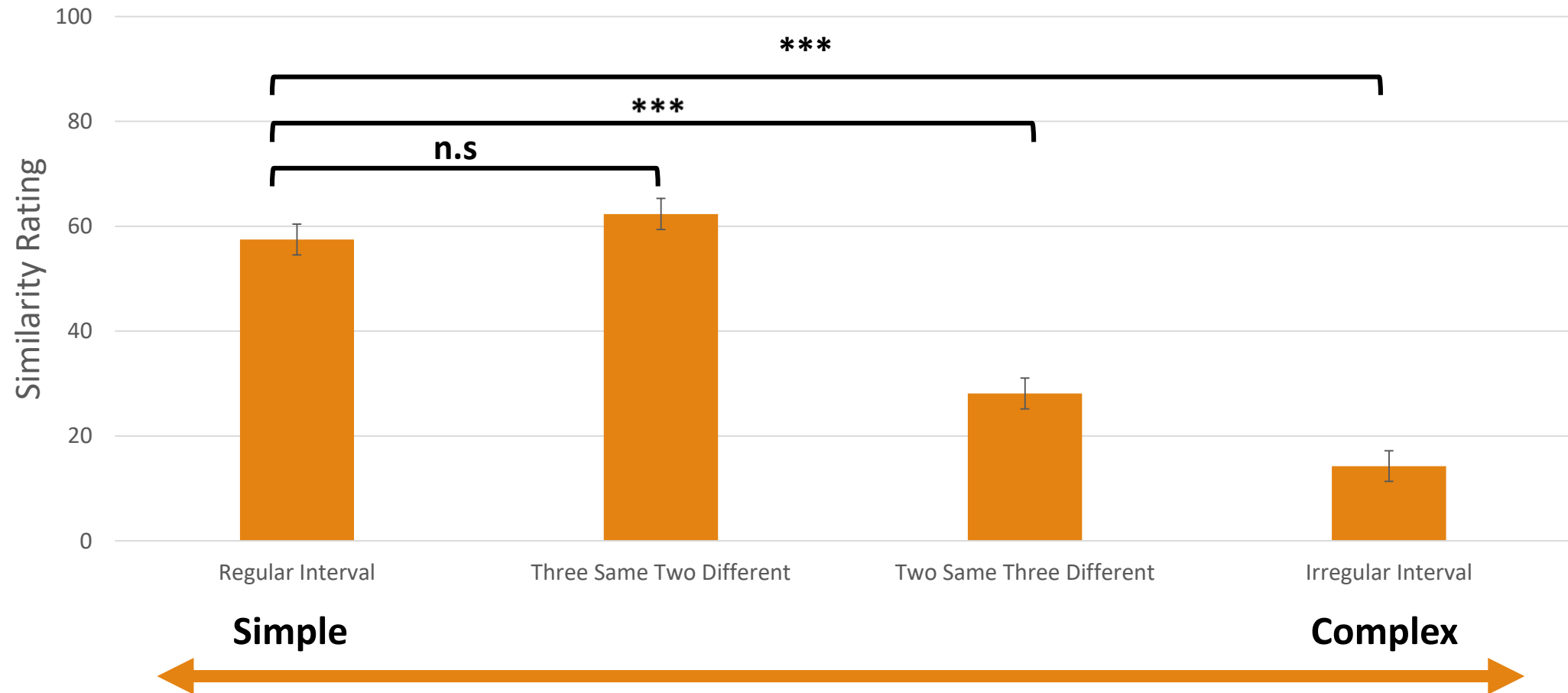


[3, 6, 9, 12, 15]

“How similar are these numbers to each other?”

0-100 scale, 0 = *Extremely dissimilar*, 100 = *Extremely similar*

Study 1c: Results



* $p < .05$, ** $p < .01$, *** $p < .001$

Study 1d: Shapes

Conditions

1) All Same ▲▲▲▲▲

2) Regular Interval ▲▲▲▲▲

3) Irregular Order (Condition 2) ▲▲▲▲▲

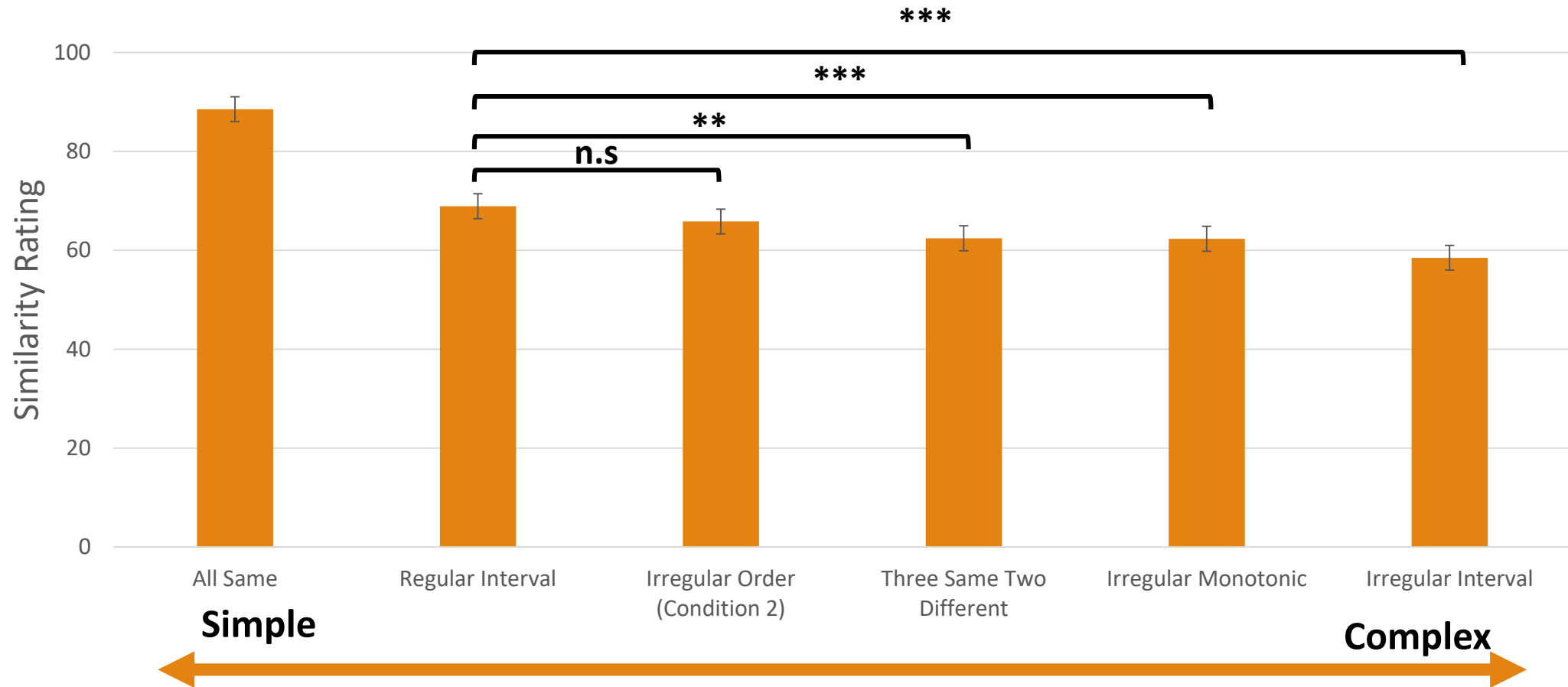
4) Three Same Two Different ▲▲▲▲▲

5) Irregular Monotonic ▲▲▲

6) Unequal Interval ▲▲▲



Study 1d: Results



* $p < .05$, ** $p < .01$, *** $p < .001$



Study 2: Numbers Regularity Factor

Conditions (N = 50; 121 strings of numbers)

1) All same: [1, 1, 1, 1, 1], [4, 4, 4, 4, 4]

2) Regular Interval: [0, 2, 4, 6, 8], [6, 5, 4, 3, 2], [1, 3, 5, 7, 9]

3) Regular Interval (Longer): [0, 2, 4, 6, 8, 10, 12], [3, 6, 9, 12, 15, 18, 21]

4) Two Same Three Different: [2, 6, 0, 0, 8], [2, 2, 4, 8, 9], [0, 1, 4, 4, 9]

5) Irregular Monotonic: [0, 1, 3, 6, 9], [1, 2, 5, 6, 8], [3, 5, 7, 8, 9]

6) Irregular Interval: [6, 4, 8, 7, 1], [0, 6, 4, 8, 1], [7, 0, 5, 1, 9]

Simple

Complex

Study 2: Numbers Regularity Factor

Possible explanations using Tversky's Contrast Model

$$S(a,b) = \frac{f(A \cap B)}{f(A \cap B) + \alpha f(A - B) + \beta f(B - A)},$$

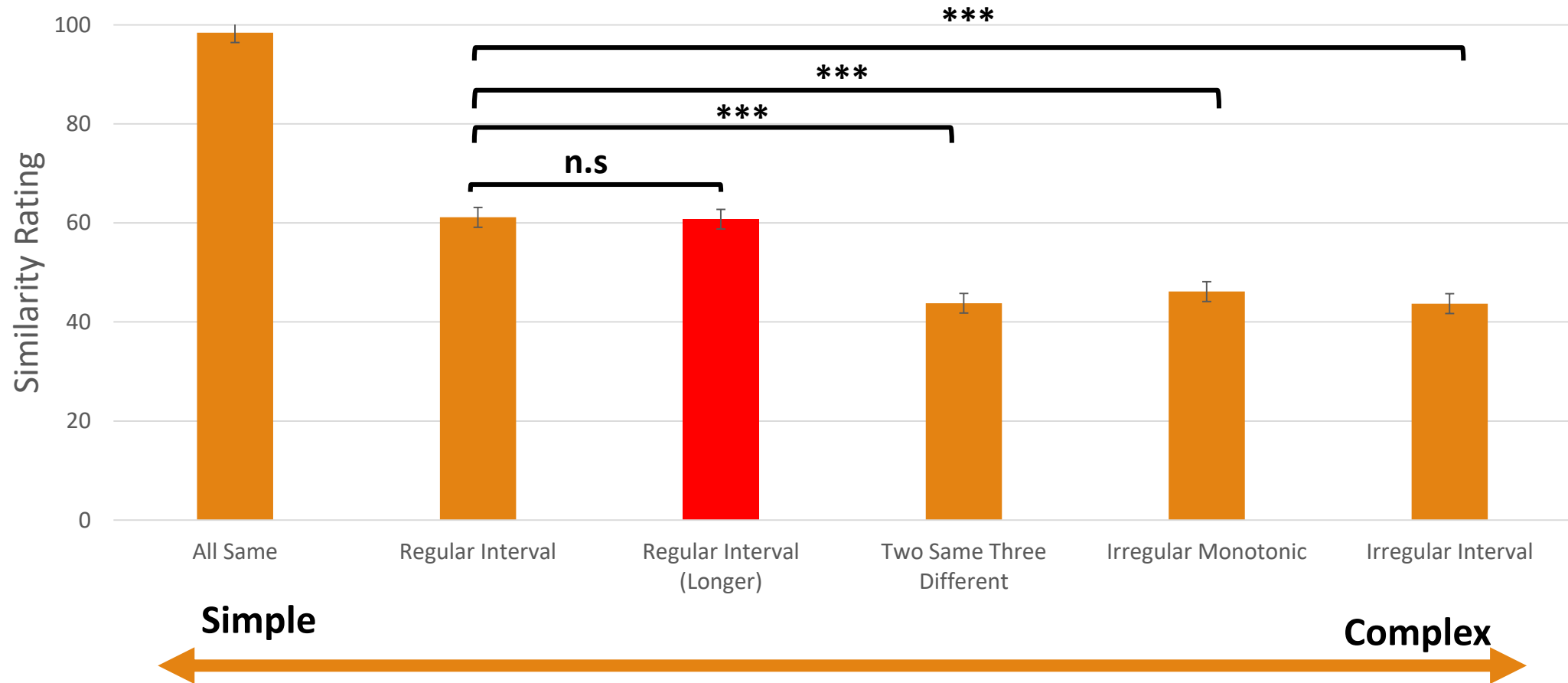
$\alpha, \beta \geq 0,$

- 1) Regularity is a shared feature for each individual item (numerator increases), in which case longer strings should receive a bigger boost from regularity as compared to shorter strings

- 2) Regularity could be a shared overarching factor (denominator increases), in which case longer strings should show a decreased boost as compared to shorter strings.

- 3) Regularity could be some sort of “gestalt” factor that is merely added to final similarity judgments in which case the length of the string should not affect similarity judgments**

Study 2: Results



* $p < .05$, ** $p < .01$, *** $p < .001$

Summary

- We find that regularity in differences among a group of objects increases judgments of similarity
- Set of results somewhat conflicts with the prevalent similarity models
 - [2, 2, 0, 6, 4] vs. **[7, 14, 21, 28, 35]**

Thank you!

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