

# Shape scission: causal segmentation of shape

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#### Crumpled!









#### Crumpled!



#### Crushed!













 Evidence that observers can infer the causal history of objects (Chen & Scholl, 2016; Leyton, 1989; Op de Beeck, Torfs, & Wagemans, 2008; Pinna, 2010; Pittenger & Todd, 1983; Spröte & Fleming, 2013, 2016; Spröte, Schmidt, & Fleming, 2016)



• We suggest that we do this by separating shape into causal layers (shape scission)













# To what extent can we separate shape into causal layers?





























































































"Name or describe the change or process that happened to the object."



• Responses (n = 16)







Responses (n = 16)





inflated cracked kneaded aged dried (out) grated imprinted irradiated



Responses (n = 16)





inflated kneaded aged dried (out) grated imprinted irradiated



• Responses (n = 16)

| indented | 100% |  |
|----------|------|--|
|          |      |  |
|          |      |  |
|          |      |  |
| dried    | 81%  |  |
|          |      |  |
|          |      |  |
|          |      |  |



# indented/pushed in



• Responses (n = 16)

| indented | 100% |            |
|----------|------|------------|
| melted   | 100% |            |
|          |      |            |
|          |      |            |
| dried    | 81%  |            |
|          |      |            |
|          |      | $\bigcirc$ |
|          |      |            |



# showered melted/runny



• Responses (n = 16)

| indented | 100% |            |
|----------|------|------------|
| melted   | 100% |            |
| cut      | 94%  |            |
|          |      |            |
| dried    | 81%  |            |
|          |      |            |
|          |      | $\bigcirc$ |
|          |      |            |







• Responses (n = 16)

| indented | 100% |            |
|----------|------|------------|
| melted   | 100% |            |
| cut      | 94%  |            |
| grown    | 88%  |            |
| dried    | 81%  |            |
|          |      |            |
|          |      | $\bigcirc$ |
|          |      |            |



# grown/bulged out pierced/pressed in with rod

grown over / colonized



• Responses (n = 16)

| indented | 100% |            |
|----------|------|------------|
| melted   | 100% |            |
| cut      | 94%  |            |
| grown    | 88%  |            |
| dried    | 81%  |            |
| inflated | 69%  |            |
|          |      | $\bigcirc$ |
|          |      |            |



pulled heated/melted inflated squashed



• Responses (n = 16)

| indented  | 100%        |            |
|-----------|-------------|------------|
| melted    | 100%        |            |
| cut       | 94%         |            |
| grown     | 88%         |            |
| dried     | 81%         |            |
| inflated  | <b>69</b> % |            |
| stretched | 57%         | $\bigcirc$ |
|           |             |            |



# filed off/carved pierced from inside **Stretched** spikes kneaded



Responses (n = 16)

| indented                         | 100% |            |
|----------------------------------|------|------------|
| melted                           | 100% |            |
| cut                              | 94%  |            |
| grown                            | 88%  |            |
| dried                            | 81%  |            |
| inflated                         | 69%  |            |
| stretched                        | 57%  | $\bigcirc$ |
| twisted/pulled/<br>strangulated/ | 44%  |            |



twisted off strangulated fissioned pulled apart gnawed



- Classification experiment (n = 15)
  - "Group objects according to what happened to them"




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- Classification experiment (n = 15)
  - "Group objects according to what happened to them"
  - "Group objects according to the shape they had before something happened to them"





























 Observers can identify "transformations" across objects and "objects" across transformations



- Observers can identify "transformations" across objects and "objects" across transformations
- Can they infer other characteristics of transformations on top of this causal separation?



























• Responses

painting (n = 14)





• Responses

painting (n = |4|)



#### 3D mesh base shape





• Responses

painting (n = |4|)



#### 3D mesh transformed shape





• Responses

painting (n = 14)



#### 3D mesh difference





• Responses

painting (n = |4|)



#### prediction from difference





• Responses

painting (n = |4|)



#### prediction from difference



# **Experiment 2: Location** VS. r = 0.62 (across all) 1.0 correlation between painting and prediction 0.8 0.6 0.4 0.2 0.0

transformation class






















































#### 5 levels of deformation (examples)





#### 5 levels of deformation (examples)







Not deformed

Strongly deformed



• Analysis



deformation



• Analysis



deformation



• Analysis



deformation



- Mesh measures
  - magnitude of deformation: average
    distance of each deformed vertex from non deformed vertex location (per stimulus)





- Mesh measures
  - magnitude of deformation: average
    distance of each deformed vertex from non deformed vertex location (per stimulus)



area of deformation: average area of deformed faces (per stimulus)





Mesh measures





- Multiple linear regression
  - $\circ$  Explained variance: R<sup>2</sup> = 0.50
  - $_{\circ}$  Regression weights
    - Magnitude of deformation: 0.47
    - Area of deformation: 0.45





• Analysis: magnitude of deformation per transformation







Analysis: area of deformation per transformation







- Multiple linear regression per transformation ullet
  - $\circ$  Average explained variance: R<sup>2</sup> = 0.66

0.6

0.4

0.2

0

magnitude

area







- Multiple linear regression per transformation
  - $_{\odot}$  Average explained variance: R<sup>2</sup> = 0.66





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  - ...how objects have been altered by forces in their past (meaning, location, and magnitude of transformations)



- We can identify "transformations" across objects and "objects" across transformations, by separating observed features by their causal origin (*shape scission*)
- This allows us to make inferences about...
  - ...what other members of the same class might look like (classification)
  - ...how objects have been altered by forces in their past (meaning, location, and magnitude of transformations)
- Shape features to build these inferences are chosen depending on transformation

### Conclusion



- Main challenges
  - $_{\odot}\,$  Identify perceptual information used to separate causal layers

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- Main challenges
  - $_{\odot}\,$  Identify perceptual information used to separate causal layers
  - $_{\odot}\,$  Identify neural representation of this information
  - $_{\odot}\,$  Define computations that extract properties of the layers



#### Thank you for your attention!

#### ...and thanks to my colleagues from Gießen







### MDS transformation classification







#### MDS shape classification





Mesh measures





0 0.2 0.4 0.6 0.8 1.0 area of deformation

### In the limits...



We should not think about recovering causal history as an all or nothing process, but as a process that can take place on different levels of abstraction and resolution.

Depending on the specific stimulus and task, the inferences presumably span both perceptual and cognitive abilities.

- At one extreme, detecting crumples, dents or cracks in objects is presumably primarily a perceptual process.
- At the other extreme, inferring the culprit of a crime from a bloody dagger and some crumbs of mud on the floor is clearly a taxing cognitive task.
- Many other causal history inferences fall somewhere between these two extremes, presumably enabling different levels of detail in the inference about the sequence of events that led to the observed state of the object.