

How do we sort these goddamn robots: Label availability changes object classification method

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How do we sort objects into categories in the world? Typically, in the mind, objects are inextricably tied to their word labels. Seeing a tree evokes the word ‘tree’, seeing your friend evokes their name, etc.

Here we build a model of image categorization in which the categorization of objects is dependent on the labeled examples available during learning. If there is one labeled object presented, generalization of the label category will occur across objects that are most similar along the dimension of lowest visual variance. If there is more than one object labeled, then generalization will occur across the dimension of highest visual variance.

The task is for subjects to decide if a robot came from the FEP machine or not (is it a FEP or not?). The model reflects this by generating robots whose identities can be determined by sorting according to the dimension of highest variance, but not by sorting according to the dimension of lowest variance. The model is as follows: generate a bottom according to a normal distribution with high variance, then pair that bottom with a top generated from one of two other normal distributions each with variances that are lower than the bottom dimension variance but are not the same.

Generative model parameters:

- Robot bottom length $\sim N(\mu_1, \sigma_1)$ (Dimension that can diagnose robot’s machine origin)
- Robot top length $\sim N(\mu_2, \sigma_2)$ or $\sim N(\mu_3, \sigma_3)$ (Drawn from one of two distributions to fool the subject)
- **Head length - Body Length = Diff $\sim N(\mu_1 - top\mu, \sigma_1 + top\sigma)$ (Diff)
- Uniform distance between Head and Body
- $\sigma_1 \ll \sigma_2$

Schema:

- (1.1) Robots generated according to model defined above
- (1.2) Subjects presented with a variety of robots with one robot labeled ”FEP”

(1.3) Subjects presented with another labeled robot that was previously unlabeled and are asked to judge if it's a "FEP"

(2.1) Subjects presented with a variety of robots with three robots labeled "FEP"

(2.2) Subjects presented with another labeled robot that was previously unlabeled and are asked to judge if it's a "FEP"

**Confounding factor: Subjects may judge "fepness" by the difference between the lengths of the head and the body. BUT this might actually be the dimension of lowest variance, visually.

$\text{Diff} \sim \text{FoldedNormal}(\mu, \sigma)$ where $\mu_Y = \sigma \sqrt{\frac{2}{\pi}} e^{(-\mu^2/2\sigma^2)} + \mu(1 - 2\Phi(\frac{-\mu}{\sigma}))$ and $\sigma_Y = \mu^2 + \sigma^2 - \mu_Y^2$

What is the "variance" of color?

What is the "variance" of pattern?

Would the effect of categorizing by difference between Head and Body length be eliminated if you shortened the presentation time?

TODO:

- Modify the generative model such that there are actually two machines that generate robots whose identity can't be diagnosed by the dimension of least variance. (DONE)
- Convert generative robot model to javascript (DONE)
- Learn JsPsych to create experiment (DONE)
- Generate the robots on the fly within the JsPsych code, maybe create a jsPsych plugin?

Important Links

- JsPsych: <http://docs.jspsych.org/tutorials/rt-task/#part-9-displaying-the-data>
- ProbMods Playspace: <https://probmods.org/v2/exercises/02-generative-models.html>

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Generative Model Development Process:

1D clustering \rightarrow 1D labeling \rightarrow 2D clustering \rightarrow 2D labeling

1D clustering model informed by Gibson 2013.

Some notation:

Let Y be the probability that an object of a given length falls into one of two categories.

$Y \sim P(\mathbf{A}, x_i)$ where x_i are the i unlabeled observations and \mathbf{A} is the vector of categories.

$Y \sim \text{Cat}([\text{cat1}, \text{cat2}], \mathbf{p})$

$\mathbf{A} \sim \text{Dir}(\alpha)$ where $\alpha = [100, 100]$ for unlabeled examples, but changed as labeled examples introduced.

$\alpha \sim \text{Pois}(\mu \text{ of the given category label})$