

Cognitive

Topics in ~~User~~[^] Modelling

in Interactive Information Retrieval

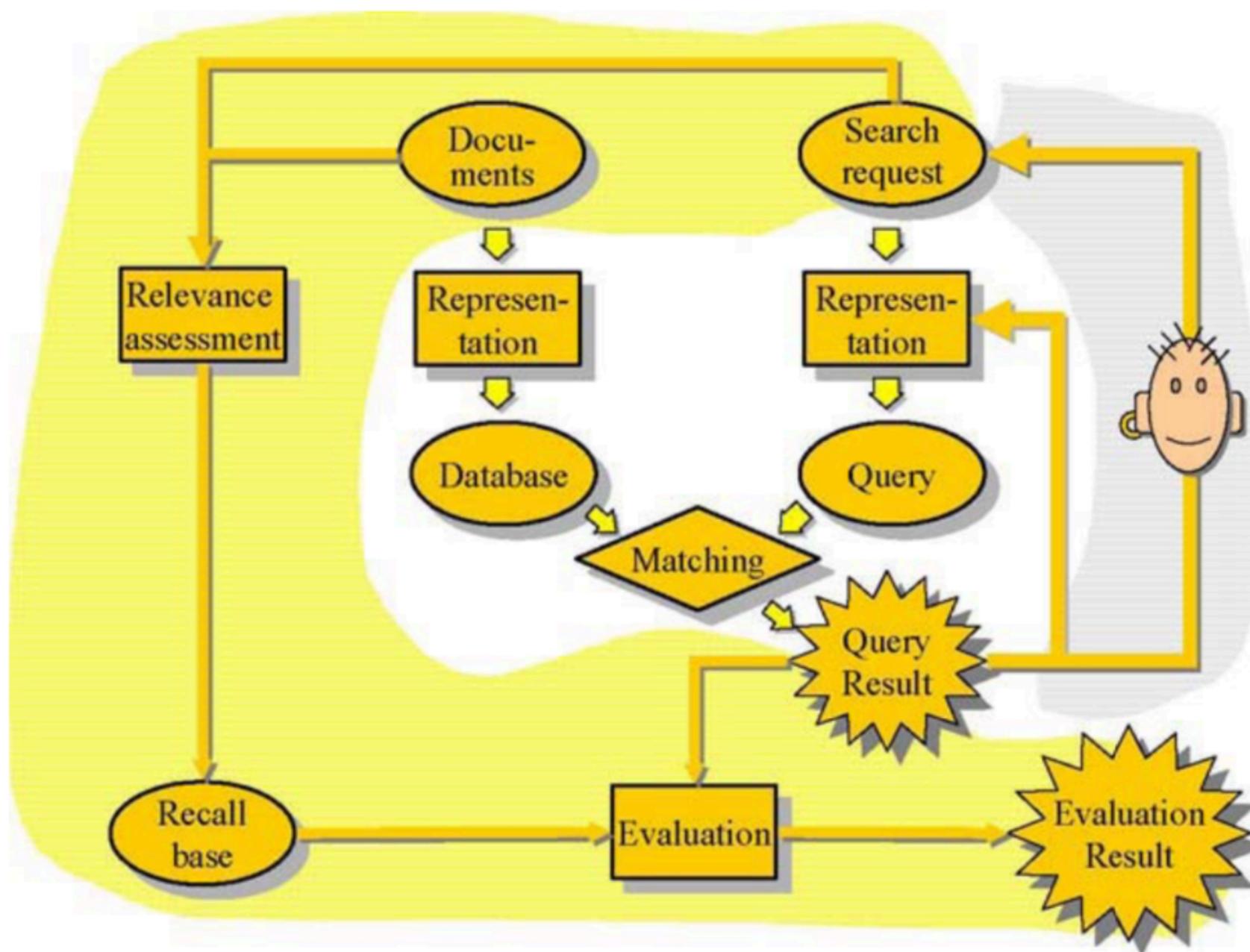
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Schedule

- **Changes will appear on the course webpage**
- 04.09.19 Lecture 1: Introduction to IR and IIR
- **11.09.19 Lecture 2: Cognitive modelling**
- 25.09.19 Deadline for topic selection (title + 3 papers min.)
- 09.10.19 Presentation of chosen topic (5 mins, 5 slides)
- 30.10.19 Feedback session
- 20.11.19 Final presentations (20 mins, 20 slides) - if necessary
- 27.11.19 Final presentations (20 mins, 20 slides)
- 11.12.19 Deadline for final paper submission

Recap 1: IR evaluation



Recap 2: IIR evaluation

Research Questions, Hypotheses, and Theory. Explicit research questions were found in 19.3% of the studies ($n = 29$), explicit hypothesis were found in 10.7% ($n = 16$) of the studies, and both a research question and a hypothesis were found in 4.7% of the studies ($n = 7$). In 65.3% ($n = 98$) of the studies, there was neither an explicitly stated research question nor hypothesis.

[objectives are to] “compare two search systems,” which suggests an implicit research question focused on basic evaluation.

D. Kelly and C. Sugimoto (2013). “A systematic review of interactive information retrieval evaluation studies, 1967–2006”. JASIST 64.4, pp. 745–770.

Recap 2: IIR evaluation

statistics, while 9% ($n = 11$) did not provide any indication of which type of analysis was used, despite claiming statistically significant results or presenting probability values.

Almost all the analyses were performed variable-by-variable and were conducted to compare the systems. Only a small percentage of articles described statistical analyses that attempted to model performance using multiple input variables ($n = 6, 5\%$).

Why cognitive models?

- Cognitive models relate models of **psychological processes** to behavioral data
- A cognitive model should be viewed as a **hypothesis** for an experiment
- We use **data** to identify which model has the best empirical support (model selection)
- We can perform evaluation on the basis of **model parameters** instead of raw data

Essay structure

- Essay will have 3 sections:
 - **An IIR component (or search task)** (e.g. ranking, relevance feedback, implicit relevance feedback) - what does it do? how is it implemented? how is its effectiveness validated?
 - **A cognitive process** (e.g. categorisation, decision making, implicit learning) - what does it study? describe the model, what type of experiment is used to gather data?
 - **Cognitive modelling in IIR** (e.g. modelling relevance feedback as a categorisation process) - sketch an experimental design, what old results can be replicated? what new results will we get?

Essay structure

- Two approaches:
 - **Correlation:** Gain an understanding of IIR component/search task + cognitive model ⇒ propose hypothesis relating the two + experiment to test hypothesis
 - **Integrative:** Gain a deep understanding of IIR component/search task + cognitive model ⇒ propose an experiment to fit the model itself

Relevance feedback

The screenshot shows a web browser window with a dark theme. The address bar displays the URL `127.0.0.1:5000`. The search bar contains the text "cognitive model". Below the search bar, there are two article cards:

Modeling Life as Cognitive Info-Computation cognitive model

Authors: Gordana Dodig-Crnkovic **Venue:** arXiv Computer Science **Date:** 28/01/2014

This article presents a naturalist approach to cognition understood as a network of info-computational, autopoietic processes in living systems. It provides a conceptual framework for the unified view of cognition as evolved from the simplest to the most complex organisms, based on new empirical and theoretical results. It addresses three fundamental questions: what cognition is, how cognition works and what cognition does at different levels of complexity of living organisms. By explicating the info-computational character of cognition, its evolution, agent-dependency and generative mechanisms we can better understand its life-sustaining and life-propagating role. The info-computational approach contributes to rethinking cognition as a process of natural computation in living beings that can be applied for cognitive computation in artificial systems.

Reusing processes and documenting processes: toward an integrated framework cognitive model

Authors: Françoise Détienne, Jean-François Rouet, Jean-Marie Burkhardt, Catherine Deleuze-Dordron **Venue:** arXiv Computer Science **Date:** 04/12/2006

This paper presents a cognitive typology of reuse processes, and a cognitive typology of documenting processes. Empirical studies on design with reuse and on software documenting provide evidence for a generalized cognitive model. First, these studies emphasize the cyclical

Positive relevance feedback (blue speech bubble pointing to the first article)

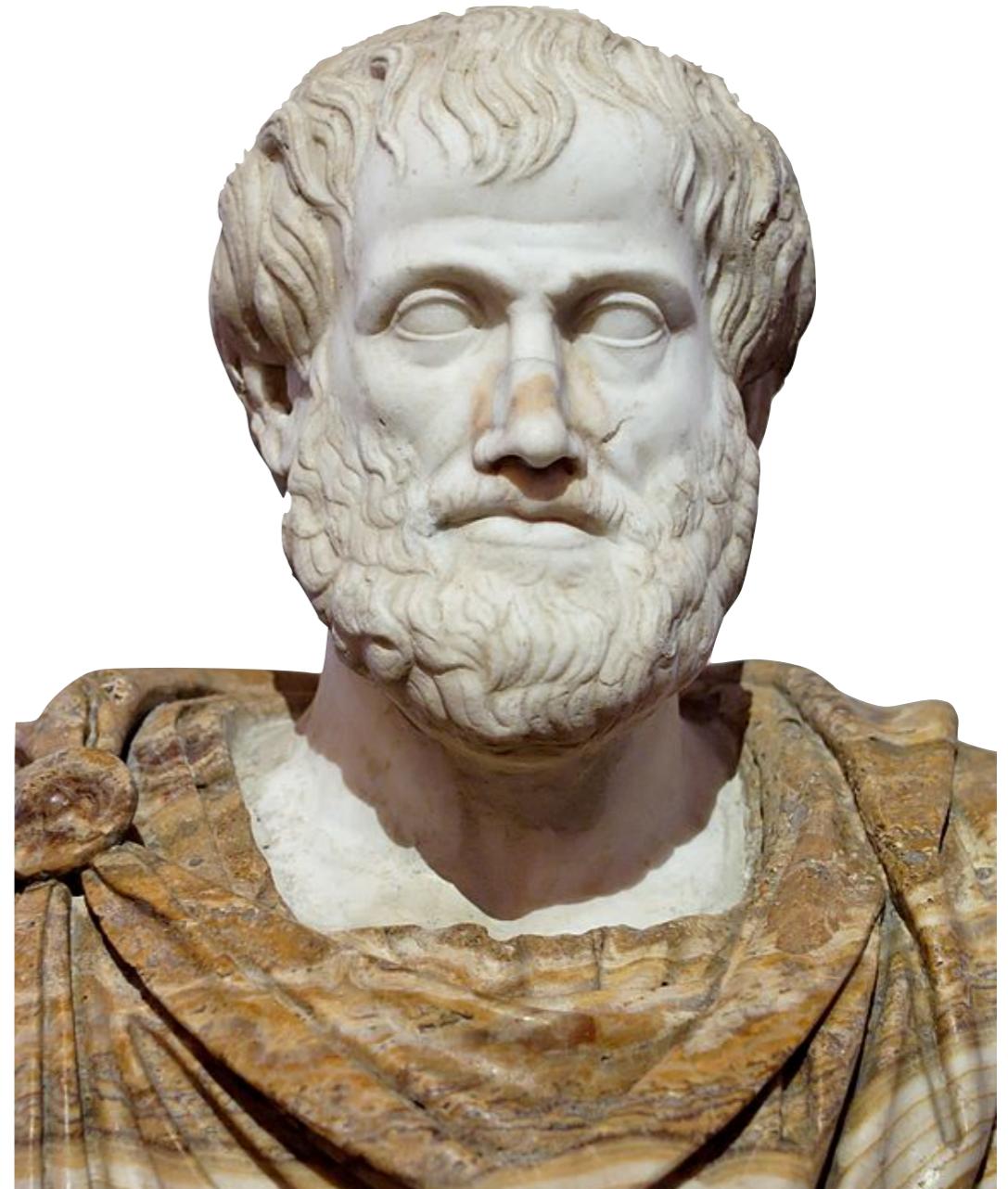
Negative relevance feedback (orange speech bubble pointing to the second article)

Theory of Categorization

- Can we determine if something is the member of a category?
 - Aristotelian model
 - Prototype models
 - Exemplar models

Aristotelian categorization model

- Classical view of categories: Plato + **Aristotle** (*The Categories*)
- Categories defined by list of features shared by all category members
- Properties are necessary conditions of category membership: entity must have **all features** to be a member of category
- Categories are strictly defined, mutually exclusive and collectively exhaustive
- Members are equal; no entity is more of that category than another



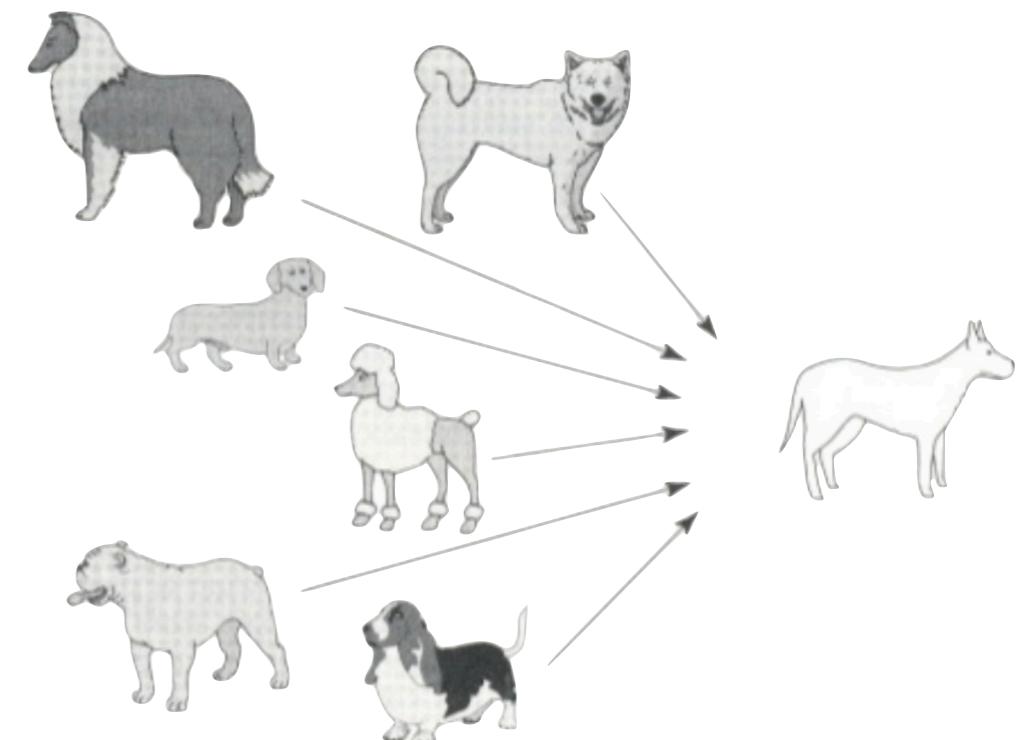
Aristotelian categorization

- Criticisms of Aristotelian categorization:
 - Entities can belong to a category, without sharing all features
 - Some entities are better exemplars of categories than others (degrees of membership)
 - Membership not strictly binary



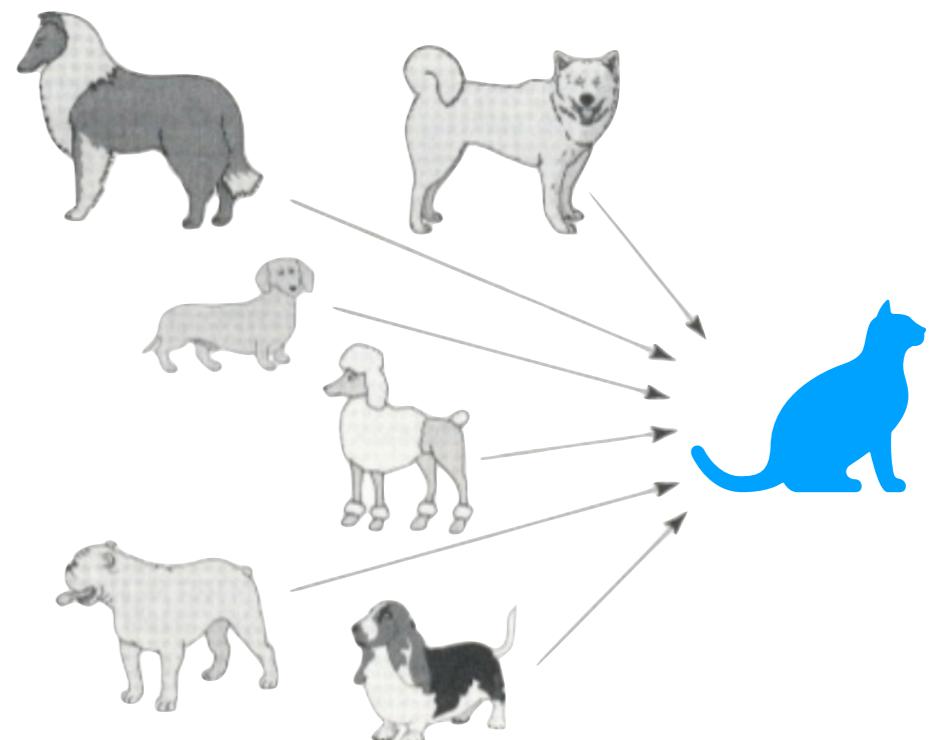
Prototype models

- Category judgments are made by comparing an object to a **prototype** (summary, average) in terms of **similarity**
- Prototype need not be real, purpose is to define membership by similarity - could emphasise features that distinguish between categories
- Categorises are not "out there" in the world, but rooted in human experience



Prototype models

- Criticisms of Prototype models:
 - Information can be lost, i.e. it is difficult to model:
 - differences in category size
 - differences in category variability
 - correlations between features
 - multimodal distributions
 - outliers!



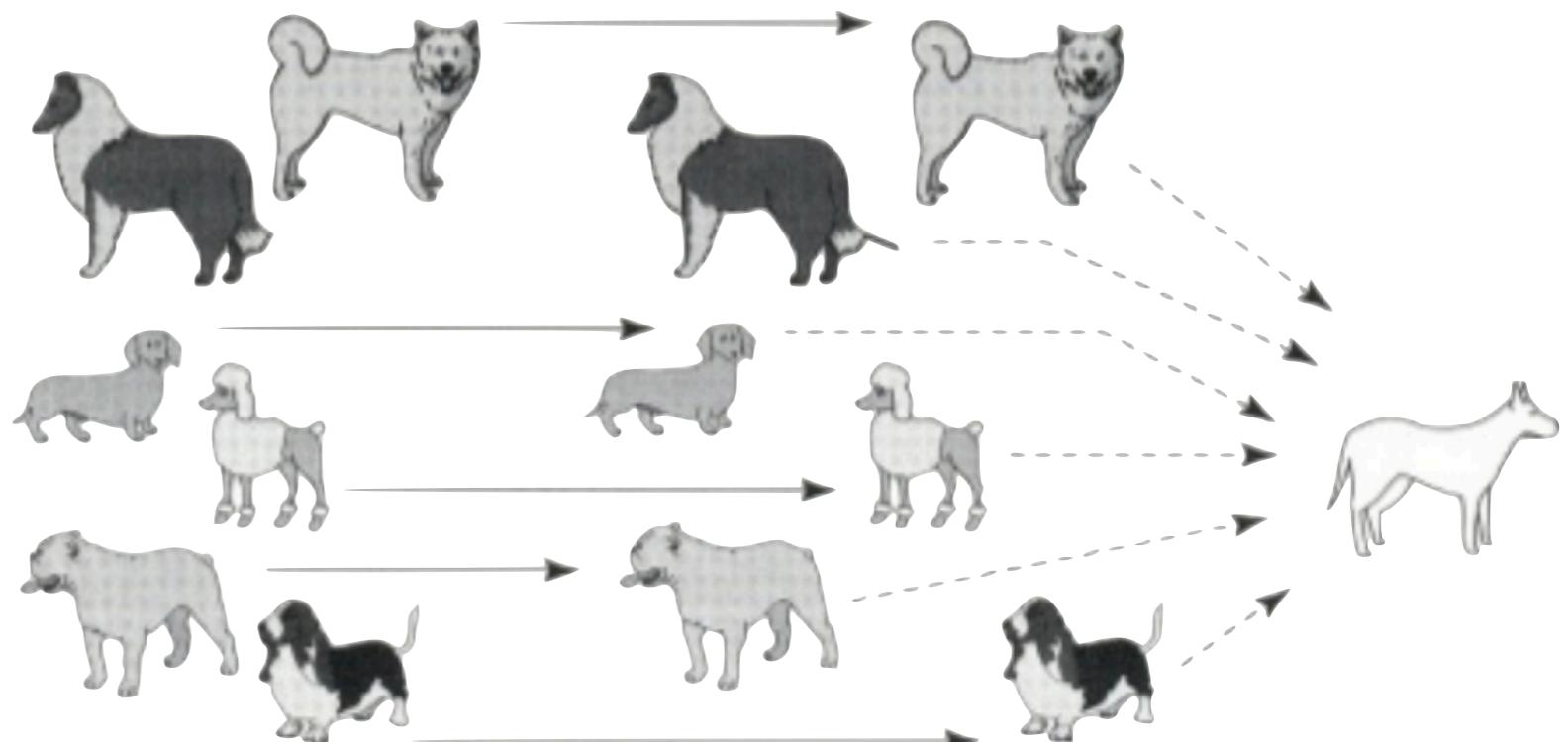
Exemplar models

- Category judgments are made by comparing an object to **all members of category** (so-called **exemplars**) in terms of similarity

- Similarities are aggregated to make a categorization decision

- **Note similarities:**

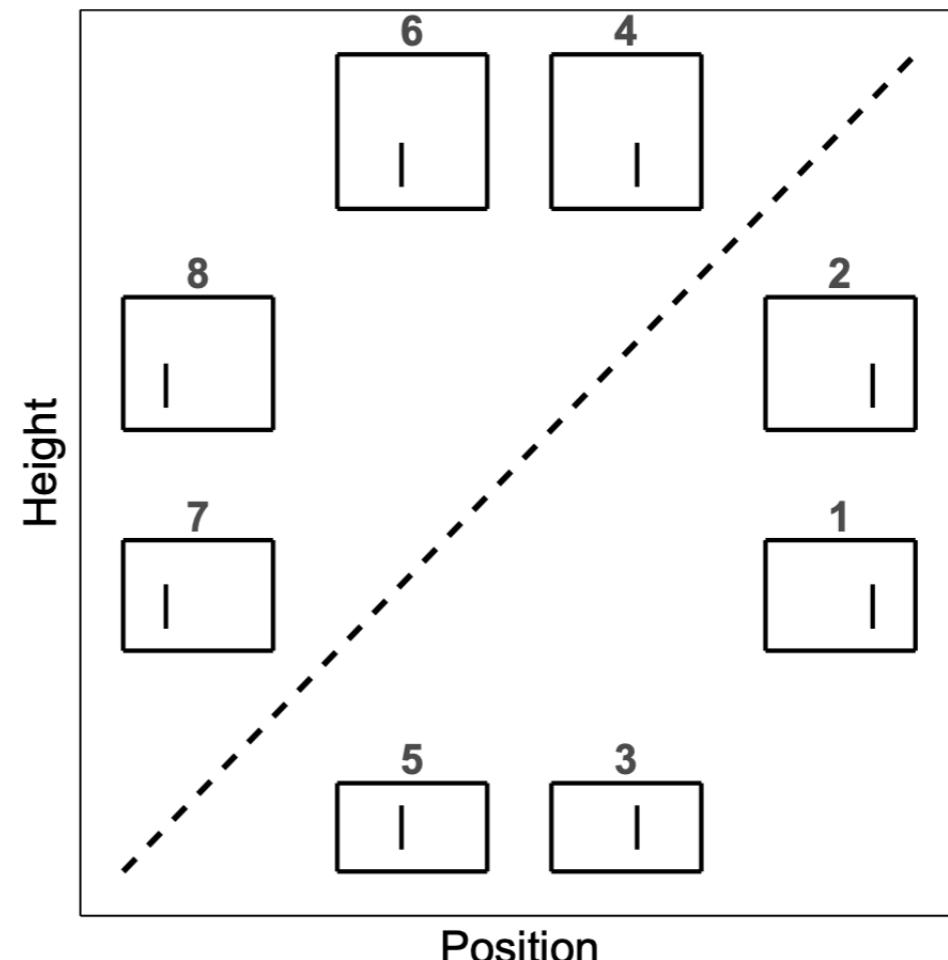
- 1 exemplar/category = prototype



- low variability of exemplars = little information loss with prototype

Categorization Example

- Categories A and B (above and below $x=y$)
- Two features:
 - position of line
 - height of box
- **What is the relative importance of features for categorization?**



Generalized Context Model (GCM)

- Exemplar model of categorization
- Category representation is just a list of category members (exemplars)
- Assume simple case (2 features per exemplar, 2 categories):
 - We need a distance function (w is the attentional weight):
$$d_{ij} = w|p_{i1} - p_{j1}| + (1-w)|p_{i2} - p_{j2}|$$
 - ...and a similarity function (c scales the drop-off in similarity with increasing distance):
$$s_{ij} = \exp(-c \cdot d_{ij})$$

R.Nosofsky (1986). “**Attention, similarity, and the identification–categorization relationship.**”
In: Journal of experimental psychology: General 115.1, p. 39.

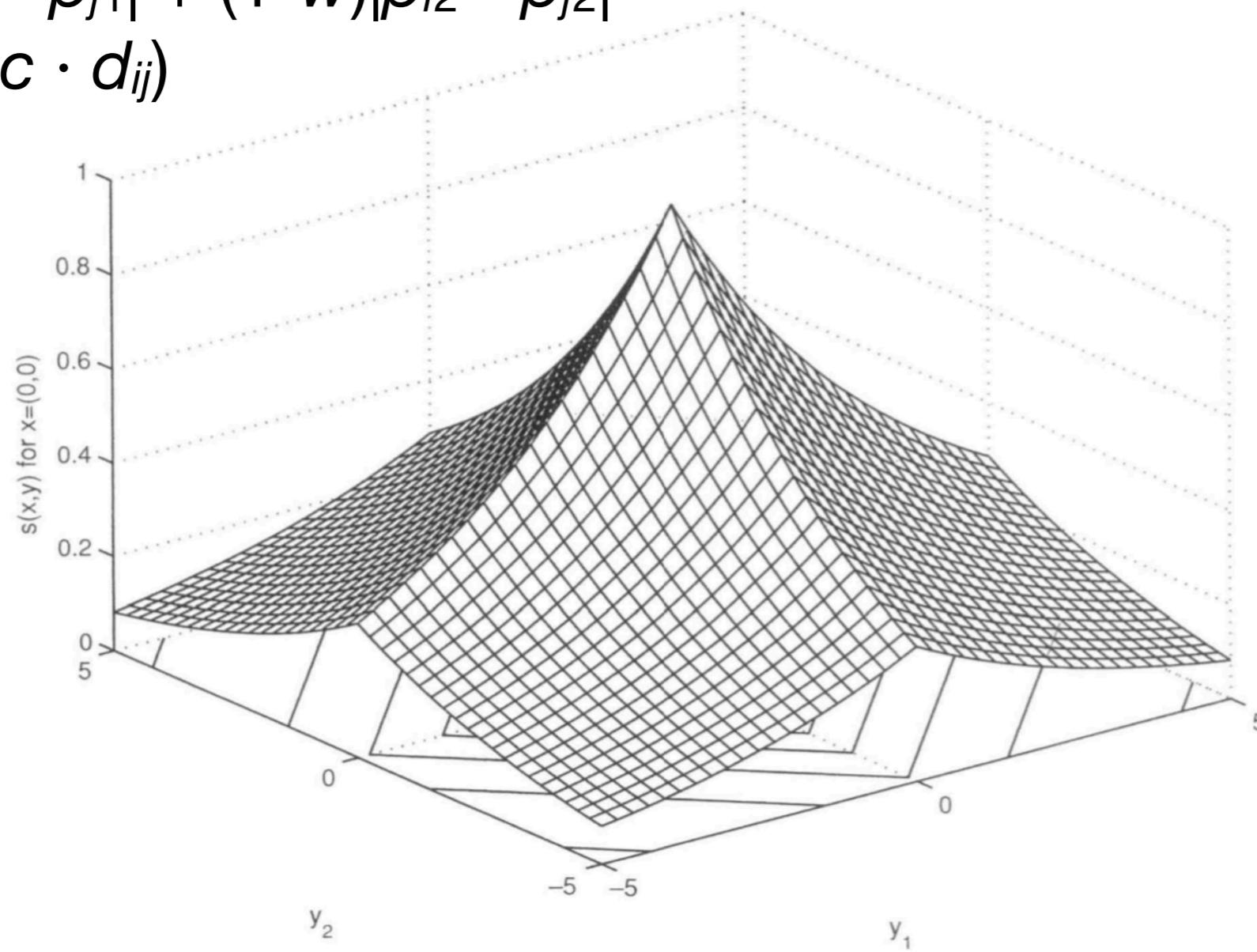
Similarity function

$$d_{ij} = w|p_{i1} - p_{j1}| + (1-w)|p_{i2} - p_{j2}|$$

$$s_{ij} = \exp(-c \cdot d_{ij})$$

$$c = 0.5$$

$$w = 0.5$$



J.Kruschke (2008). “**Models of categorization**”. In:
The Cambridge handbook of computational psychology, pp. 267–301.

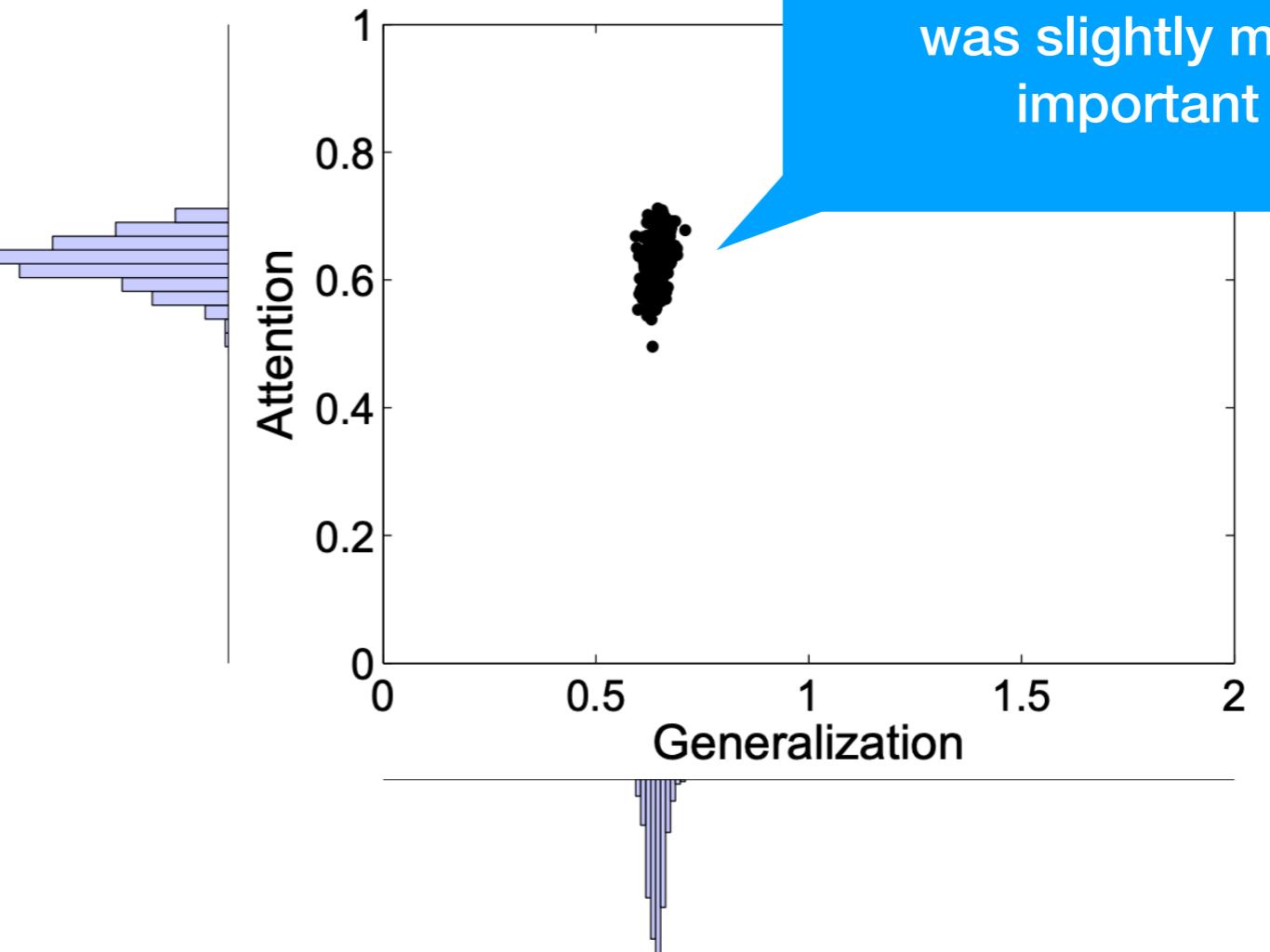
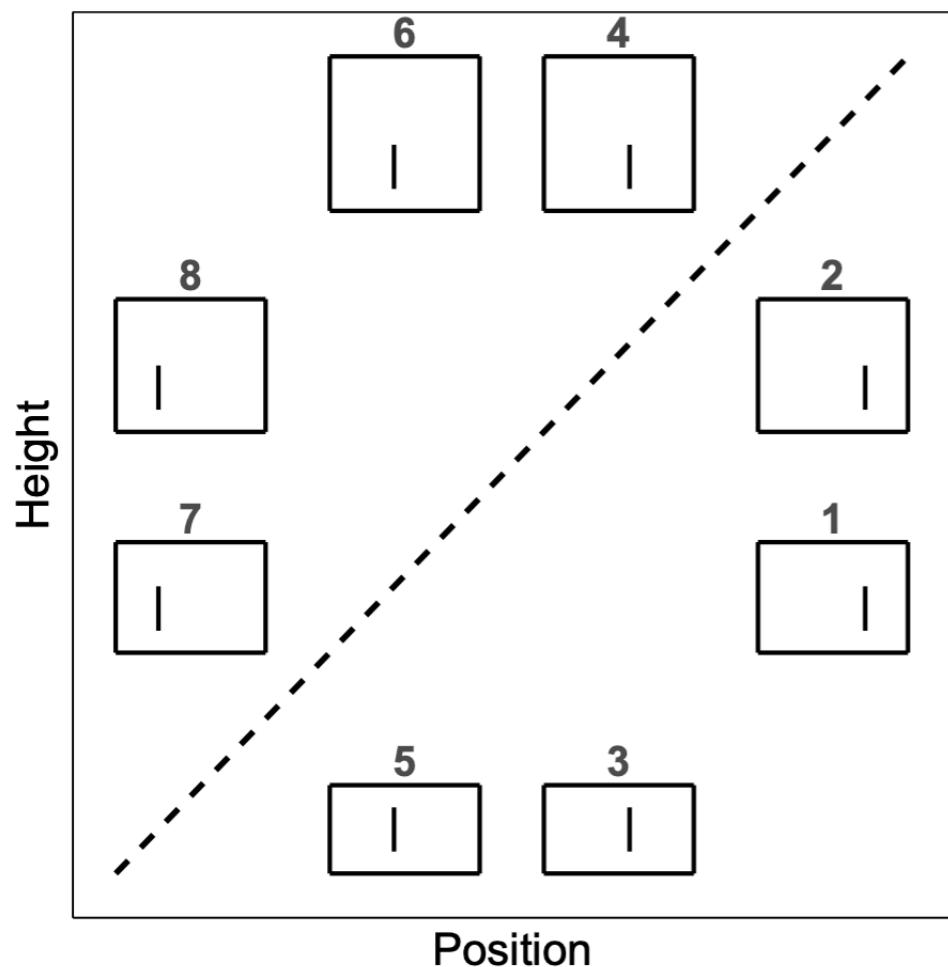
Generalized Context Model (GCM)

- Probability of classifying exemplar i into category A (as opposed to category B) is:

$$P(R_i = A|i) = \frac{\sum_{j \in A} s_{ij}}{\sum_{j \in A} s_{ij} + \sum_{j \in B} s_{ij}}$$

R.Nosofsky (1986). “**Attention, similarity, and the identification–categorization relationship.**”
In: Journal of experimental psychology: General 115.1, p. 39.

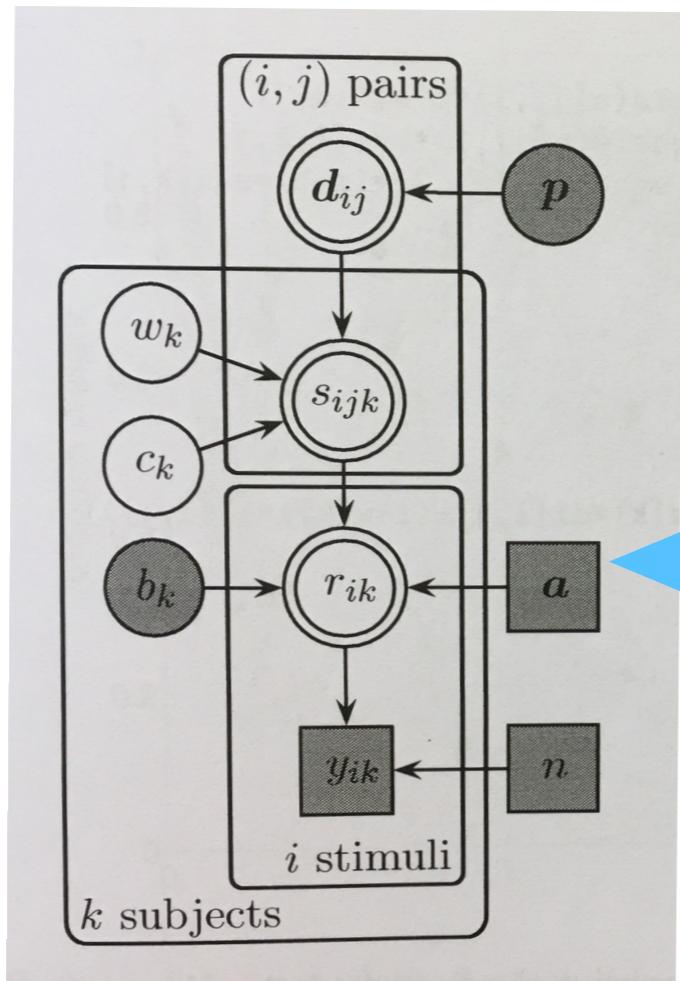
Categorization Example



M.Lee, and E.Wagenmakers (2014). "Bayesian cognitive modeling: A practical course".
Cambridge university press.

Generalized Context Model (GCM)

- Bayesian GCM with repeated measures:

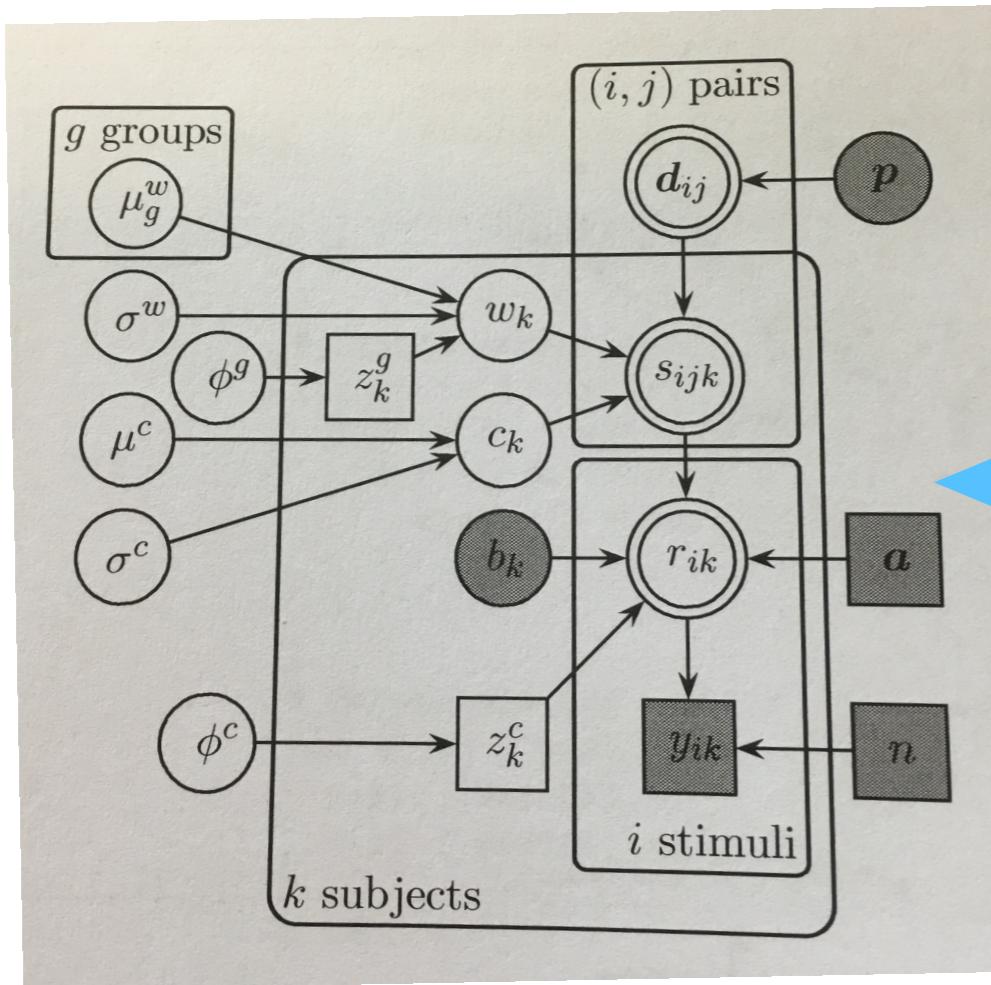


Essay idea 1: for relevance feedback, how does the attentional weight distribution (w_k) vary between:

- systems?
- user groups?
- search tasks?

Generalized Context Model (GCM)

- Bayesian GCM with repeated measures and a latent-mixture including contamination:



Essay idea 2: for relevance feedback, how does the proportion of contamination users change between:
- systems?
- user groups?
- search tasks?

Generalized Context Model (GCM)

- Probability of classifying exemplar i into category A is (alternate version):

$$P(R_i = A|i) = \frac{\left(\sum_{j \in A} s_{ij}\right)^\gamma}{\left(\sum_{j \in A} s_{ij}\right)^\gamma + \left(\sum_{j \in B} s_{ij}\right)^\gamma}$$

- $\gamma = 1$: same response as original GCM
- $\gamma < 1$: responses are increasingly random
- $\gamma > 1$: responses are increasingly deterministic

Essay idea 3: for relevance feedback, how does γ differ between lookup and exploratory search?

S.McKinley and R.Nosofsky (1995). “Investigations of exemplar and decision bound models in large, ill-defined category structures.” In: Journal of Experimental Psychology: Human Perception and Performance, 21.1, p. 128.

A.Medlar and D.Glowacka (2018). “How Consistent is Relevance Feedback in Exploratory Search?”
In: Proc. of the 27th ACM CIKM. pp. 1615–1618.

Essay suggestions

- **IIR interface components:**

- ranking
- relevance feedback
- implicit relevance feedback

- **Search tasks:**

- Lookup/exploratory search

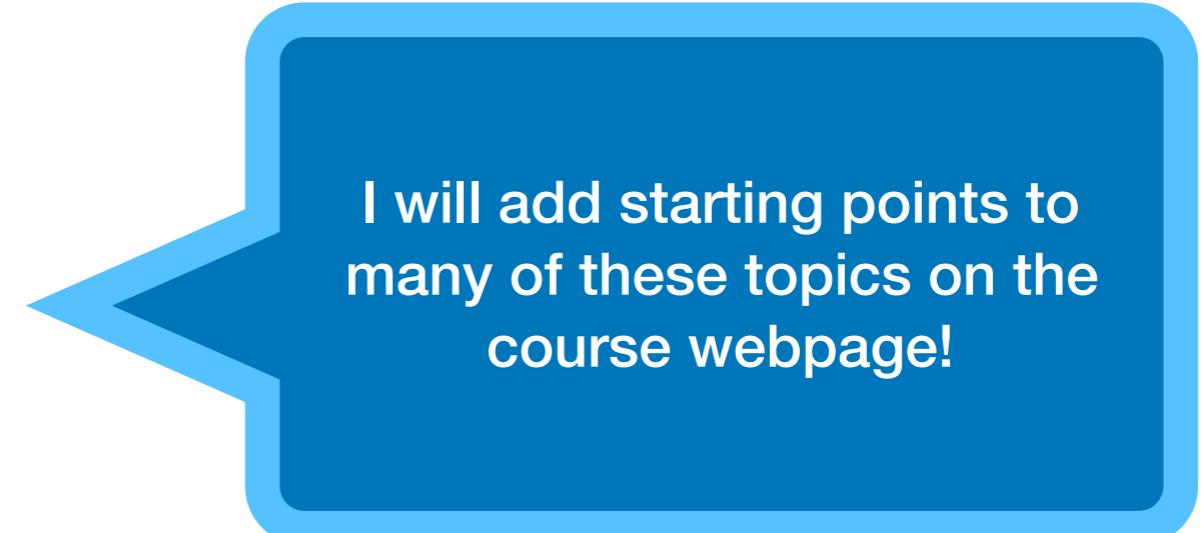
- **Cognitive models:**

- **Decision making**

- Categorization processes (prototype vs exemplar)
 - Signal detection theory (discriminability vs bias)
 - Diffusion decision processes (berry-picking)

- **Other**

- Working memory, perceptual speed, risk taking, implicit learning, etc.



I will add starting points to many of these topics on the course webpage!

Next deadline...

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