

Inference: Conscious and Unconscious

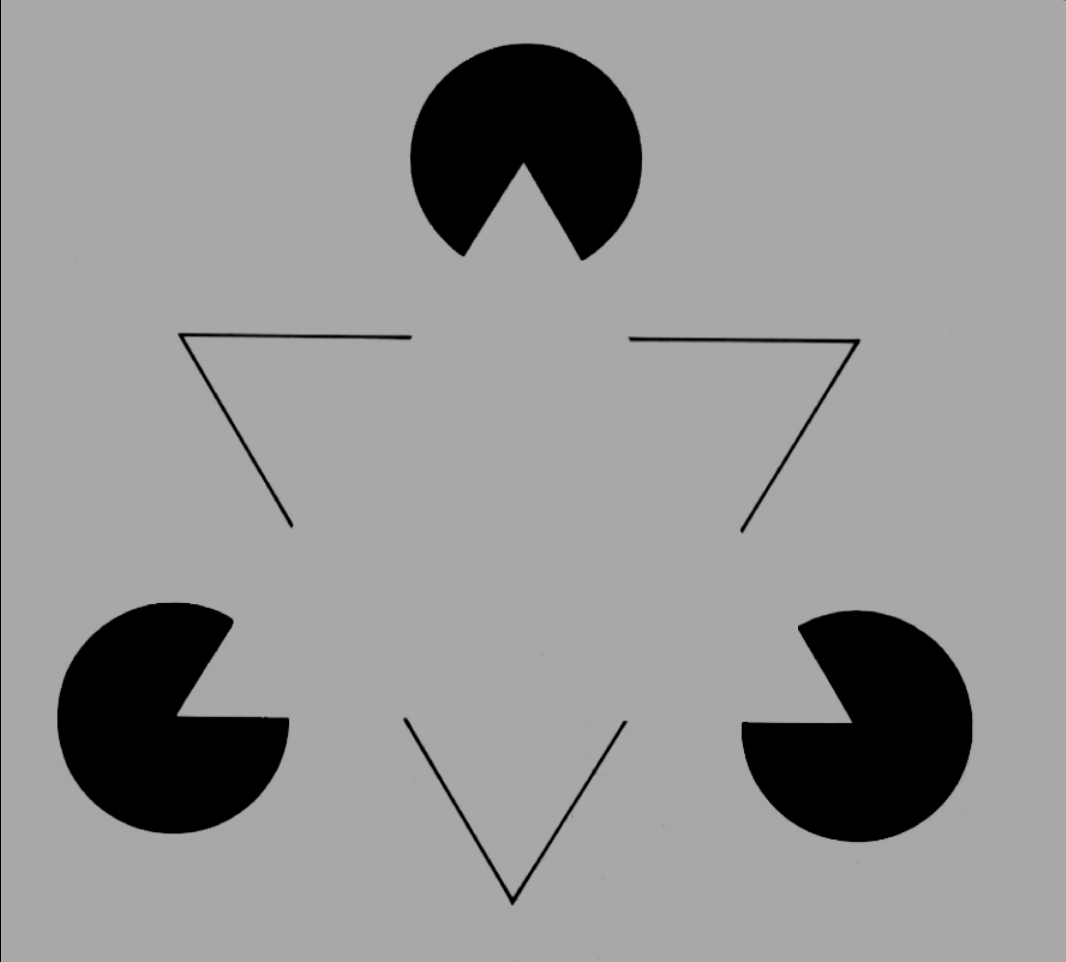
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SymSys 100
April 20, 2010

A Conscious Inference

- I wake up in morning and ask myself *has it rained overnight?*
- I look out the window and notice that the ground is wet.
- I consider whether I should conclude that it rained last night.
- I consider the fact that it is April, and rain is relatively rare.
- I consider that I have a sprinkler system, and ask myself if it is set to operate on Tuesdays.
- I conclude that any conclusion I can reach by looking at my own back yard would be tentative; so I go outside and notice that the street and sidewalks along my block are all wet.
- I conclude that it did indeed rain last night.

Some Examples of “Unconscious Inferences”



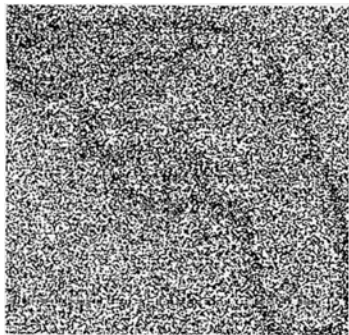
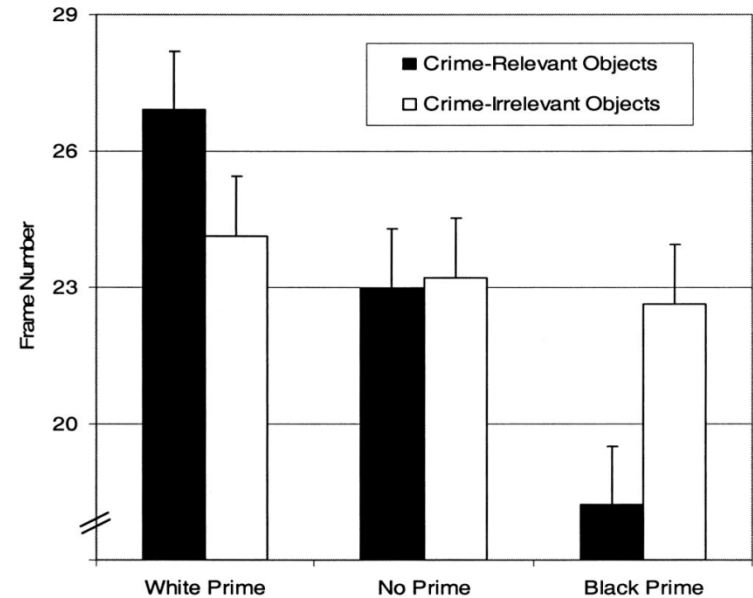






Eberhardt: *Seeing Black*

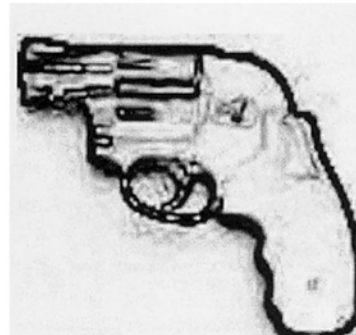
- Participants were shown very brief presentations of black faces, white faces, or nothing (20 stimuli).
- They then identified pictures that gradually became clearer frame by frame (500 msec/frame).
- Exposure to black faces led to faster recognition of crime objects, even though participants were not aware that they had been shown any faces.



Frame 1



Frame 20



Frame 41

Perception as 'Unconscious Inference'

- Helmholtz coined the term in the 19th century, drawing on ideas going back to the ancients.
- The term suggests a 'little man in the head', entertaining hypotheses, weighing evidence, and making decisions.
- But we can think of it differently: perhaps your brain is wired to
 - Use context together with direct information
 - Combine multiple sources of information
 - Allow 'hypotheses' to exert mutual influences on each other



Rest of This Lecture

- Examine a “normative theory” of inference under uncertainty
 - How can we reason from given information to conclusions when everything we know is uncertain?
 - How can we combine different sources of evidence?
- Consider two experiments that produce data consistent (in a sense) with the normative theory.

Next two lectures

- A neuro-mechanistic examination of unconscious inference
 - How neurons can produce output matching the normative theory
 - How networks of neurons can work together to produce collective behavior consistent with the normative theory.

An inference question

- An apartment building has a surveillance camera with automatic face recognition software. If the camera sees a known terrorist, it will trigger an alarm with 99% probability. If the camera sees a non-terrorist, it will trigger the alarm 1% of the time.
- Suppose somebody triggers the alarm. What is the chance he/she is a known terrorist?

The Normative Theory

Bayesian Inference: Definitions

- Hypothesis (It's a terrorist)
- Evidence (Alarm rang)
- Probability:
 - Expected Ratio of Occurrences to Opportunities
 - NB: Sometimes these are *estimated* or *subjective* probabilities
- $P(E|H)$: Likelihood
 - = $P(E \& H) / P(H)$: In this case: .99
- $P(E|\sim H)$
 - = $P(E \& \sim H) / P(\sim H)$: In this case: .01
- $P(H|E)$
 - = $P(E \& H) / P(E)$
- Can we compute $P(H|E)$?

Deriving Bayes' Rule

- The definition $P(H|E) = P(E\&H)/P(E)$ implies $P(H|E)P(E) = P(E\&H)$
- Similarly, $P(E|H) = P(E\&H)/P(H)$ implies $P(E|H)P(H) = P(E\&H)$
- It follows that $P(H|E)P(E) = P(E|H)P(H)$
- Divide both sides by $P(E)$ we obtain $P(H|E) = P(E|H)P(H)/P(E)$
- Final Step: Compute $P(E)$

In our case, the probability that the alarm goes off when a person walks by the camera

- Can we derive $P(E)$ from quantities we've already discussed?

$$P(E) = P(E|H)P(H) + P(E|\sim H)P(\sim H)$$

- Substituting, we get Bayes' Rule:

$$P(H | E) = \frac{P(E | H)P(H)}{P(E | H)P(H) + P(E | \sim H)P(\sim H)}$$

Importance of 'priors' or 'base rates'

- If terrorists and non-terrorists are equally likely, $p(H|E) = .99$
- If only one person in 100 is a terrorist, $p(H|E) = .5$
- If only one person in 10,000 is a terrorist, $p(H|E)$ is $\sim .01!$

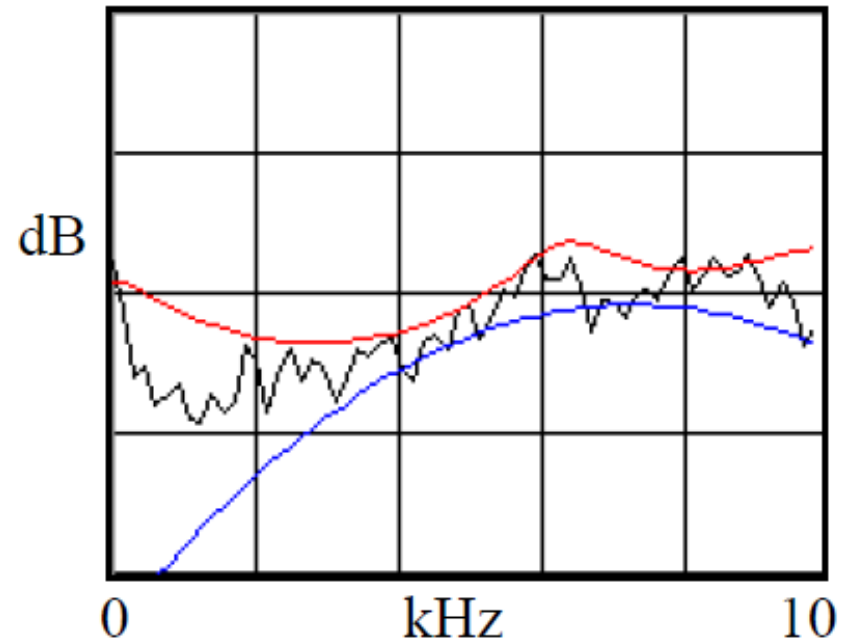
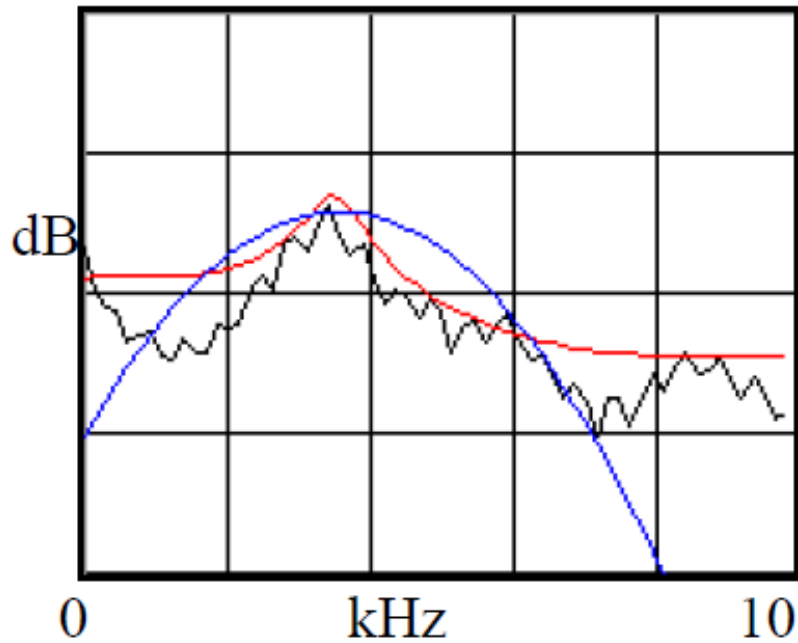
Lessons From this Example

- Priors are crucial if we are to derive a correct inference from the data!
- For items with low 'base rates' (priors), people often overestimate the posterior, $p(H|E)$.
- In explicit inferences, people often neglect base rates (or underweight them)

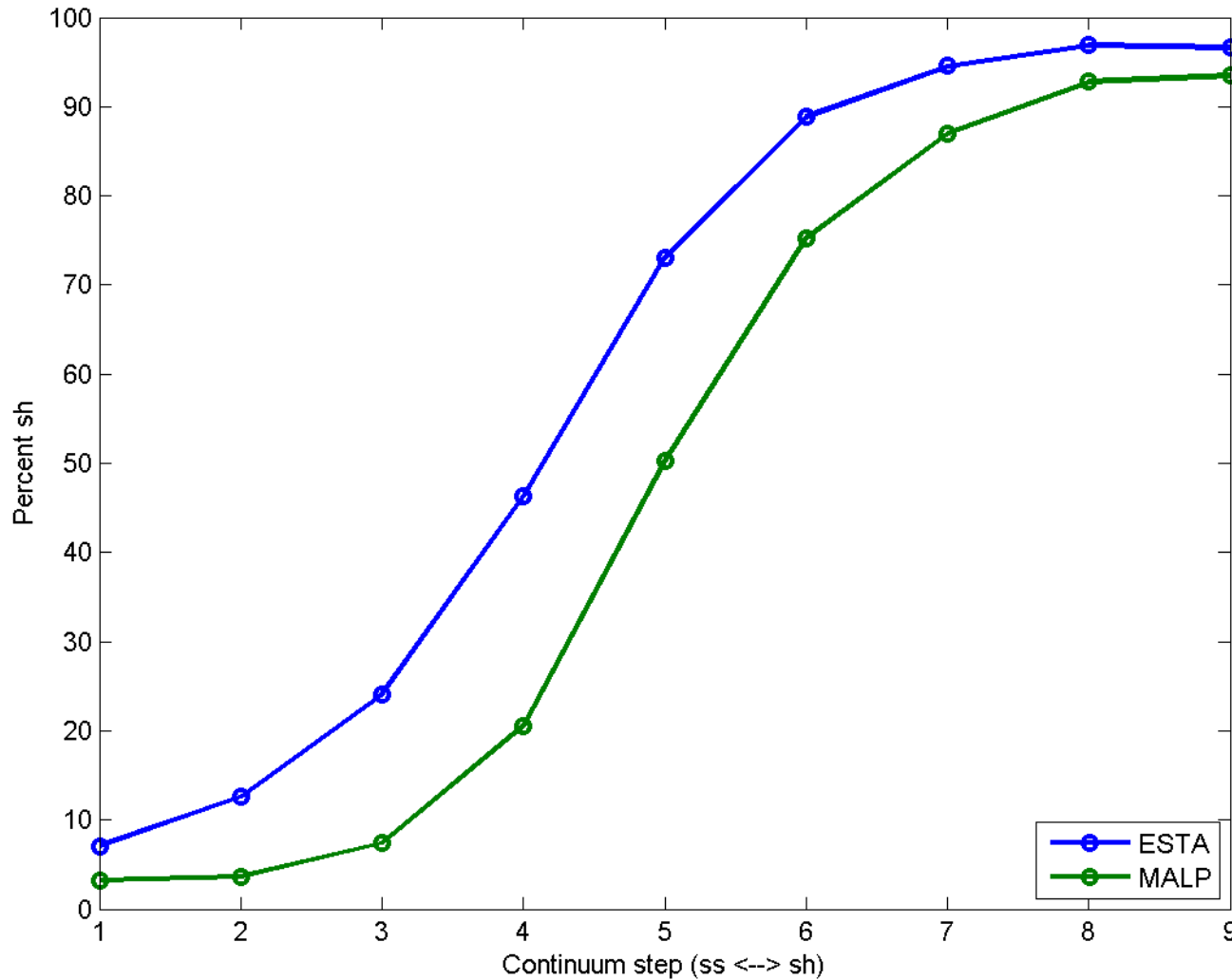
Ganong Experiment

- Spoken words such as ‘establish’ or ‘malpractice’ are presented but the final fricative sound is replaced with an ambiguous sound, somewhere on a continuum from ‘ss’ to ‘sh’.
 - Stimuli are blends of natural ‘ss’ and ‘sh’ sounds.
- The participant must decide: is the last sound ‘ss’ or ‘sh’.

Natural 'sh' and 'ss'



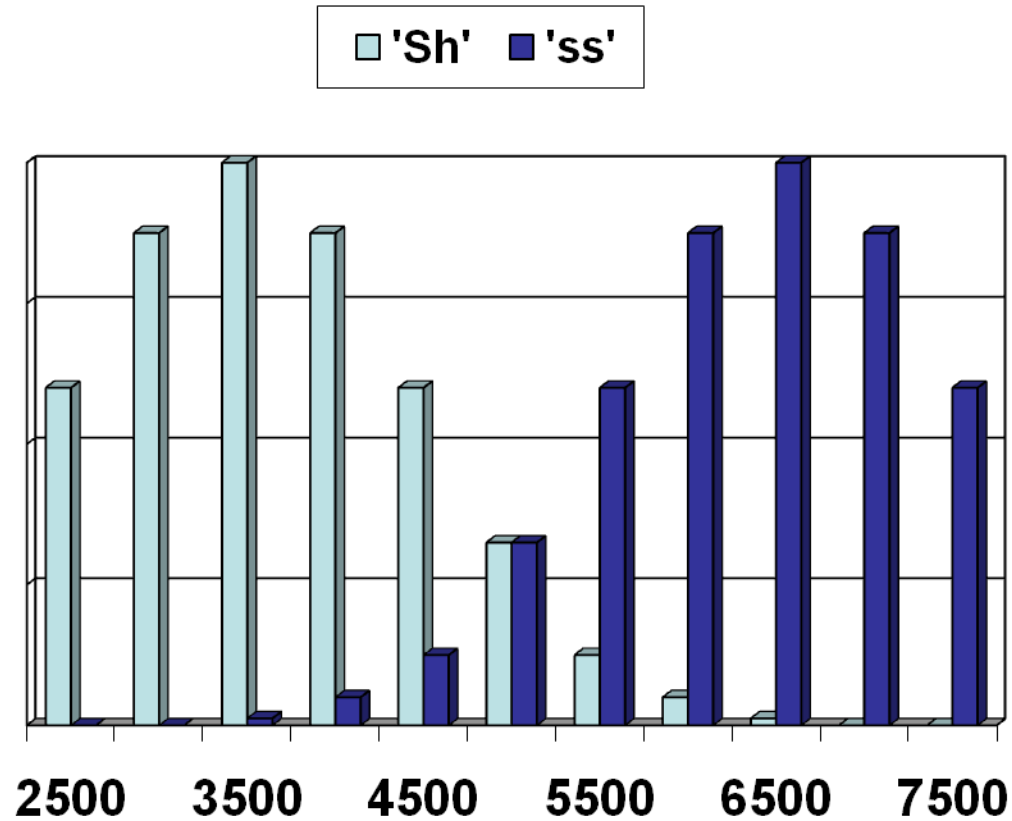
Data from experiment performed last year:
P('sh') for 'establi..' and 'malpracti...' contexts



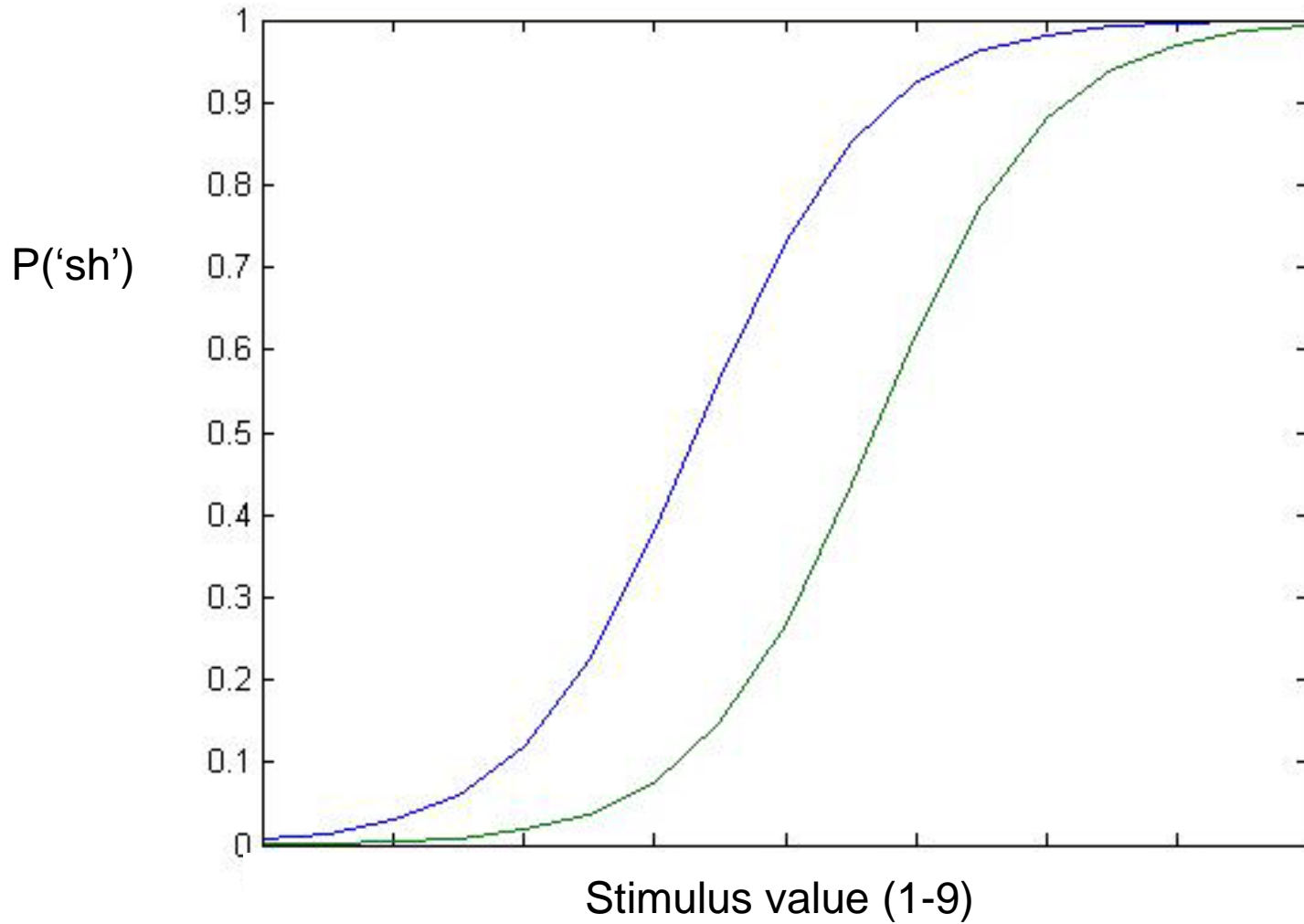
Can we understand these data from a Bayesian Perspective?

- We can think of the different stimuli used as differing in their probability, given that the stimulus was 'sh' or 'ss'.
- We can think of 'establi...' increases the prior probability of 'sh' relative to 'ss'
- Similarly, we can think of 'malpracti...' as increasing the prior probability of 'ss' relative to 'sh'
- The effect is substantial, but not overwhelming:
$$\frac{p(\text{sh}|\text{establi}...)}{p(\text{sh}|\text{establi}...) + p(\text{s}|\text{establi}...)} = .63$$
$$\frac{p(\text{sh}|\text{establi}...)}{p(\text{s}|\text{establi}...)} = .63/.37 \approx 1.7$$
- Is the effect too big, or is it too small?

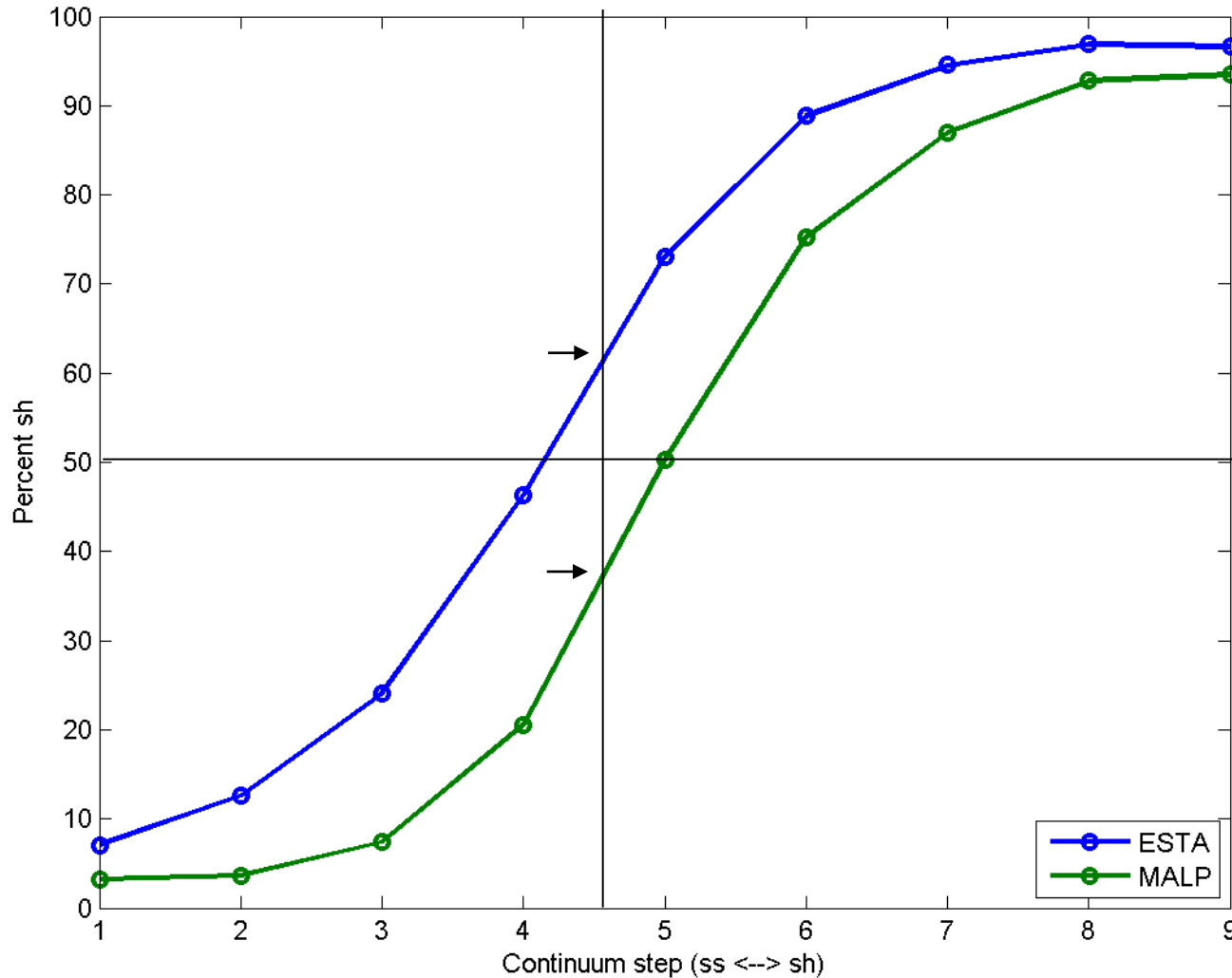
$P(\text{stim}|\text{sh})$ & $P(\text{stim}|\text{ss})$ [Hypothetical!]



Expected form of the data from Ganong experiment



Data from experiment performed last year: P('sh') for 'establi..' and 'malpracti...' contexts



How might we combine information from different sources?

- For example, visible and heard speech (The *McGurk* effect)
- Auditory cue: starting frequency of syllable
- Visible cue: degree of mouth opening
- Treat A and V as *conditionally independent*
 - $P(Ea \& Ev | H) = P(Ea | H)P(Ev | H)$ for $H = 'ba'$ or $'da'$
- Then we get this normative model:

$$P('da' | Ea \& Ev) = \frac{P(Ea | 'da')P(Ev | 'da')P('da')}{P(Ea | 'da')P(Ev | 'da')P('da') + P(Ea | 'ba')P(Ev | 'ba')P('ba')}$$

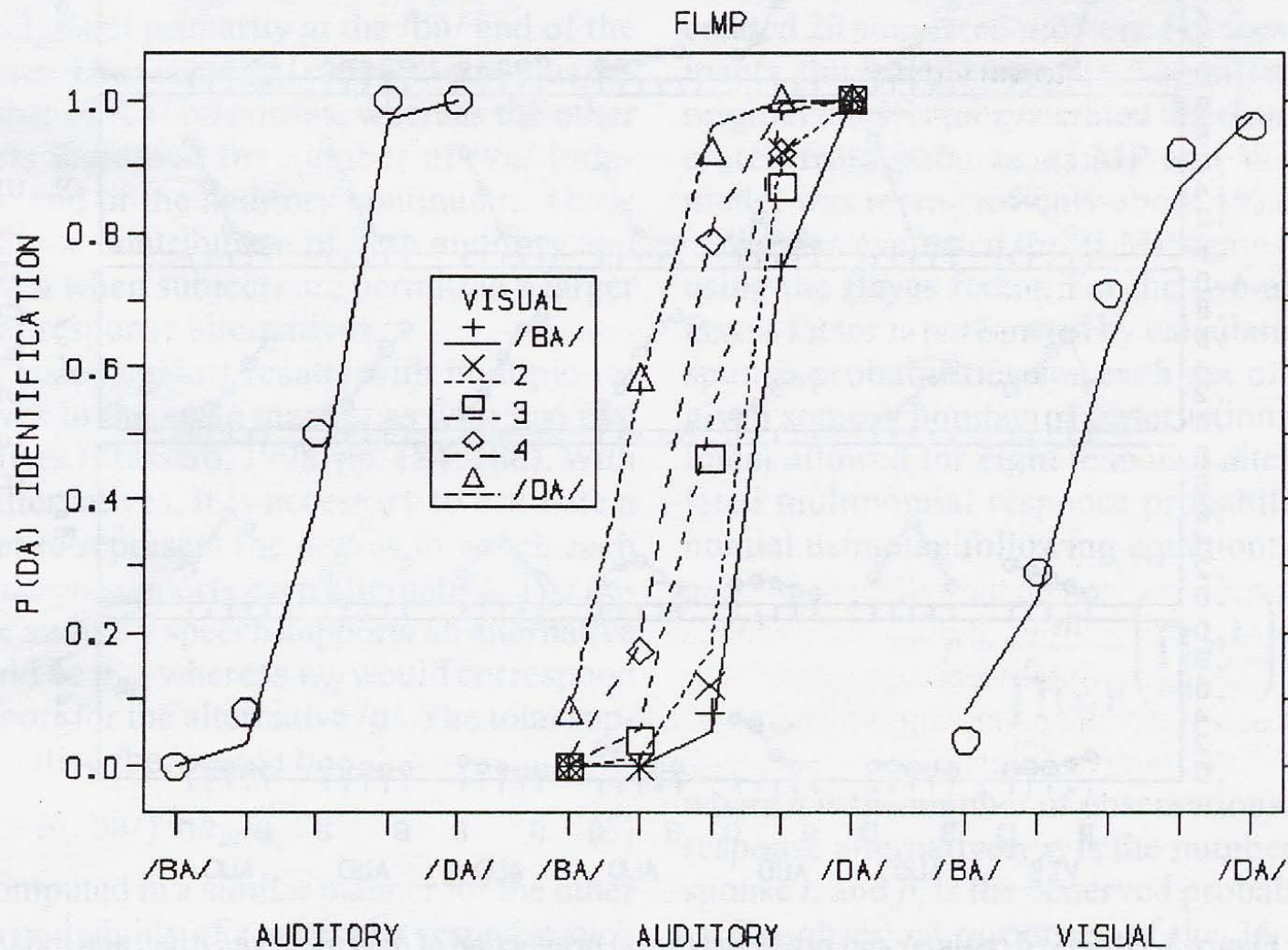


Figure 2. The points give the observed proportion of /da/ identifications for a typical observer in the auditory-alone (left panel), the factorial auditory–visual (center panel), and the visual-alone (right panel) conditions as a function of the five levels of the synthetic auditory and visual speech varying between /ba/ and /da/. The lines give the predictions of the FLMP.

Summary

- People make unconscious inferences all the time.
- The responses they make often approximately match an estimate of the posterior, assuming that context adjusts their priors towards alternatives that are consistent with the given context.
- People also often combine different sources of information in a way that is consistent with Bayes rule, and the additional assumption that the different sources of evidence are treated as conditionally independent.
- Next time we will consider how neurons in the brain could carry out such computations.