

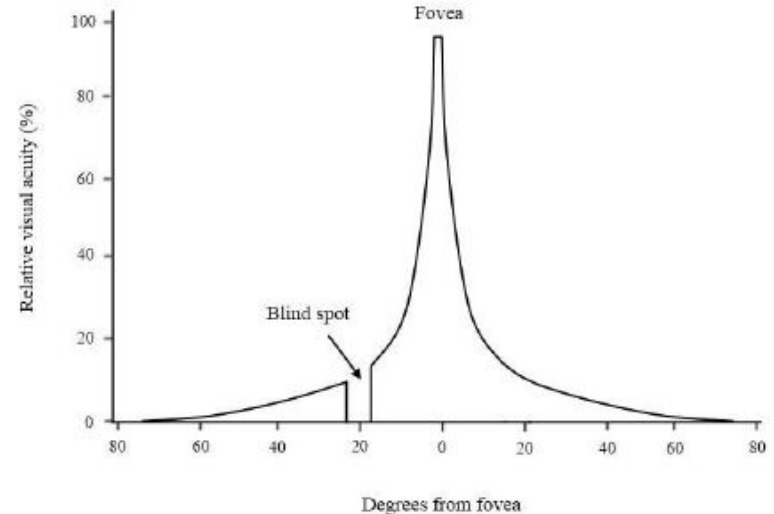
# Eye movements & psychophysics

INDP Human Cognitive Neuroscience 2024

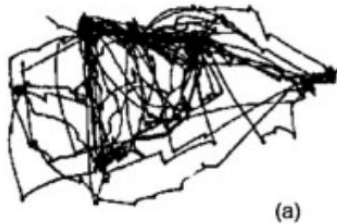
# Why do humans need to make eye movements?

Despite a  $\sim 180^\circ$  visual field, good visual acuity in humans is restricted to just  $1\text{-}5^\circ$  of visual angle spanned by the fovea.

That's approximately size of your thumbnail held at arm's length.



# Eye movements in cognitive neuroscience



(a)



(b)



(d)



(f)



(c)



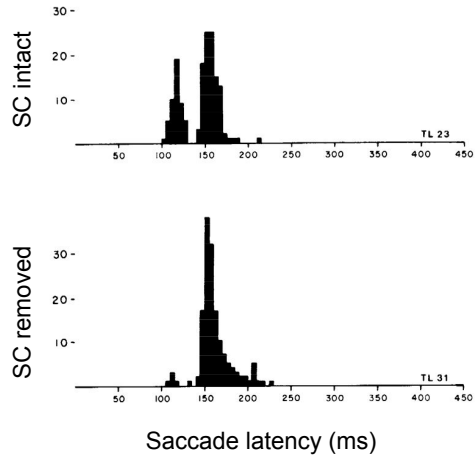
(e)



(g)

# “Bottom-up” contributions to eye movements

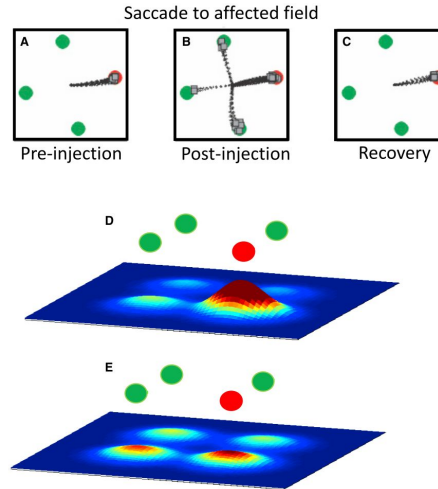
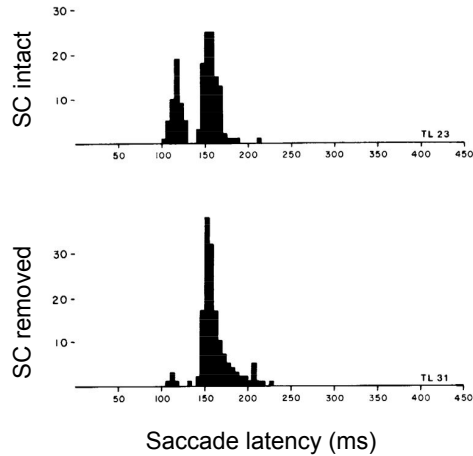
Superior colliculus ablation abolishes rapid saccades to a target.



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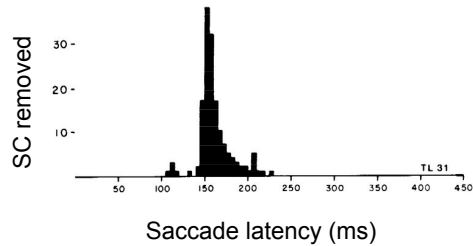
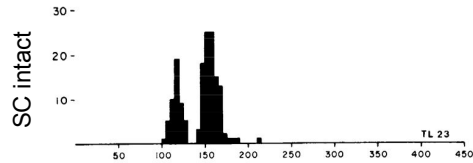
Superior colliculus ablation abolishes rapid saccades to a target.

SC ablation hinders the ability to saccade to a target, and topographic SC activity predicts target detection.

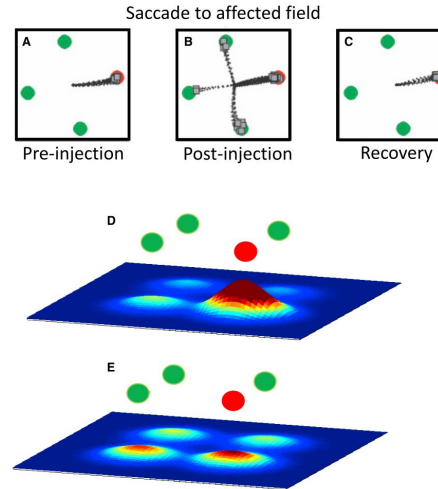


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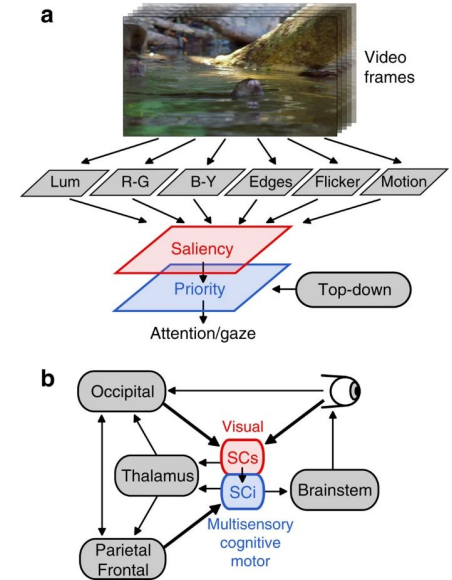
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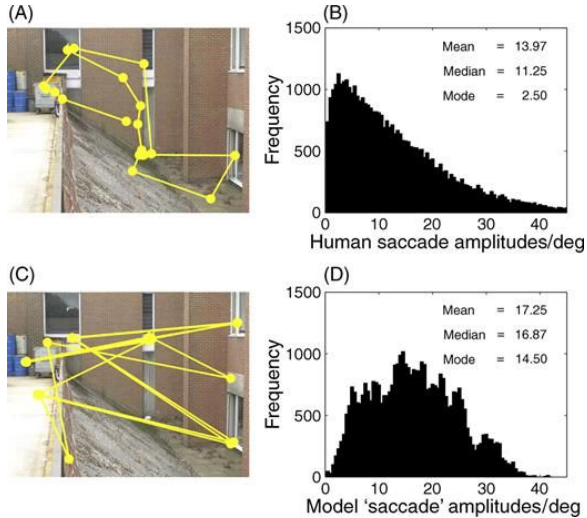


“Saliency map” models predict fixations when free-viewing a scene and are encoded in SC activity



# “Top-down” contributions to eye movements

Salience models do not account for the basic statistics of eye movements during free-viewing.

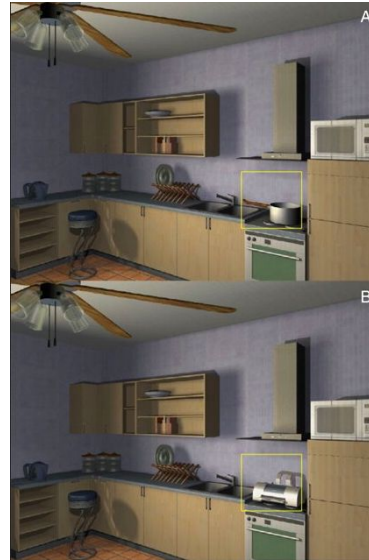
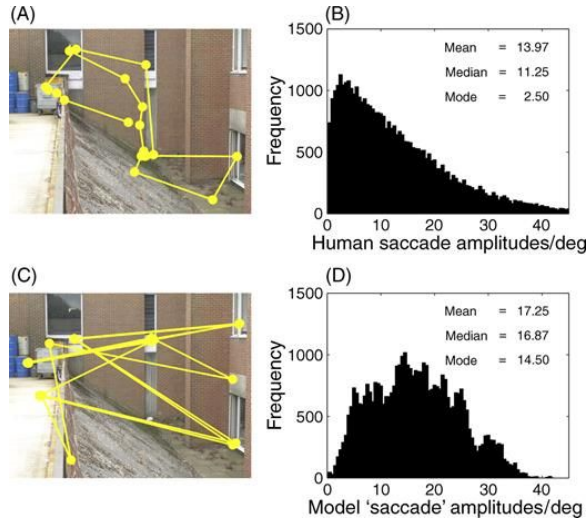


Shinoda et al. (2001), Vo & Henderson (2009), Tatler et al. (2011), Borji & Itti (2014)

# “Top-down” contributions to eye movements

Salience models do not account for the basic statistics of eye movements during free-viewing.

While performing a task, our expectations determine where we look in the scene.

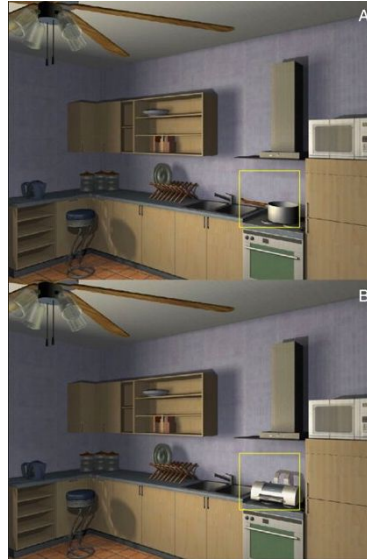
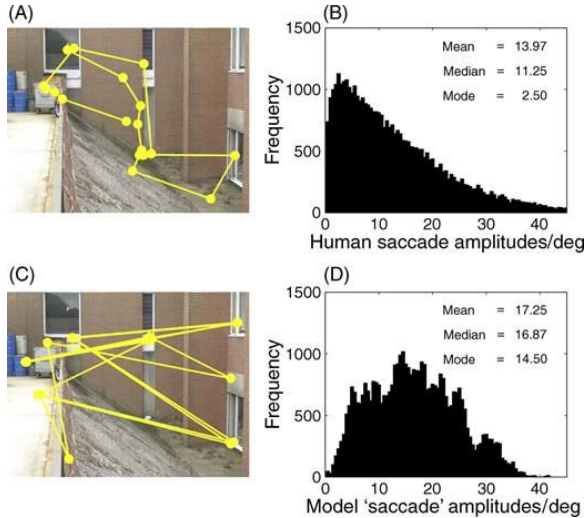
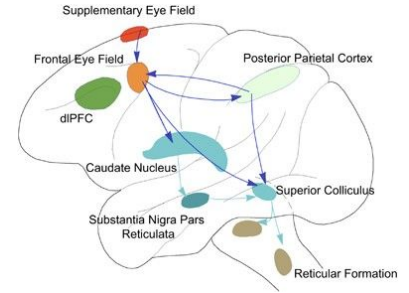




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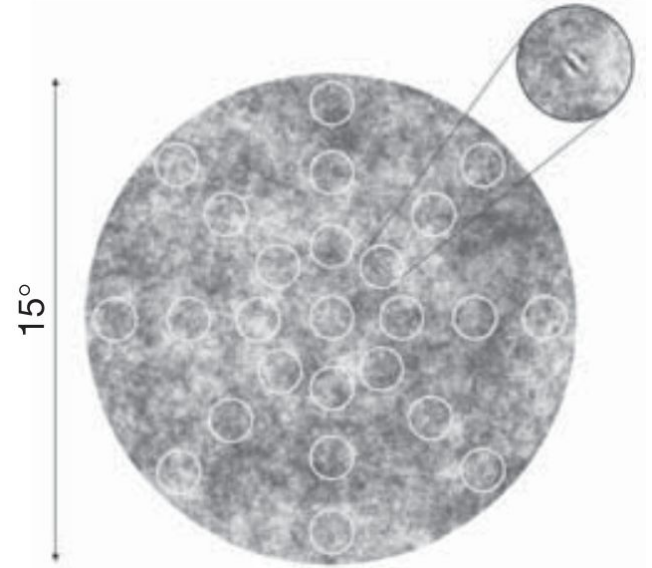


# Modeling human visual search

## Optimal eye movement strategies in visual search

**Jiri Najemnik & Wilson S. Geisler**

Najemnik & Geisler (Nature, 2005)



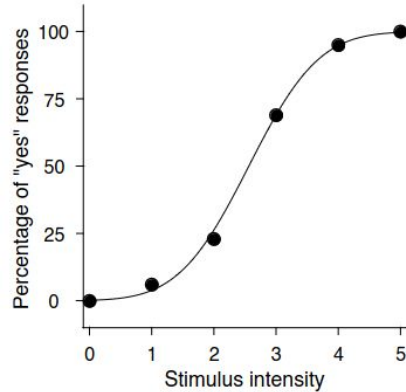
# Modeling human visual search

- 1) Extract a visibility map
- 2) Simulate a Bayesian searcher
- 3) Compare simulated searchers to human data

# 1) Quantifying visibility: the psychophysics toolkit

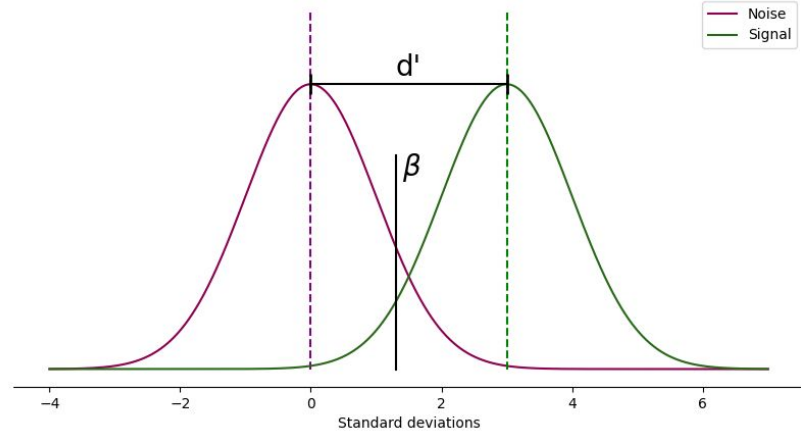
# 1) Quantifying visibility: the psychophysics toolkit

## Psychometric functions



$$\psi = \gamma + (1 - \gamma - \lambda) \cdot F(x; \alpha, \beta)$$

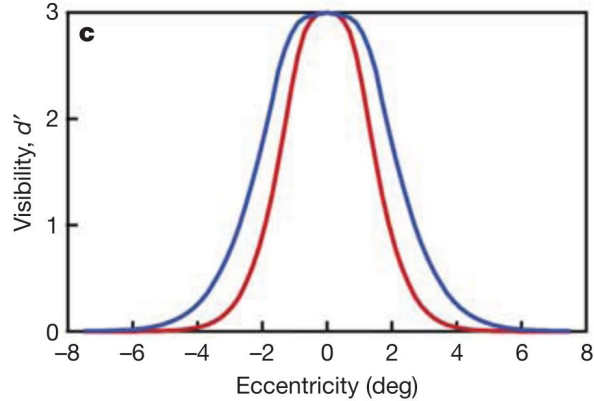
## Signal detection theory



# Modeling human visual search

1) Extract a visibility map

2) Simulate a Bayesian searcher



3) Compare simulated searchers to human data

## 2) Where to look next: Bayesian search

- 1) **Compute sensory evidence** for each point in space by multiplying a hypothetical visual response with the visibility function

## 2) Where to look next: Bayesian search

$$P(H \mid E) \propto P(E) \cdot P(E \mid H)$$

$P(H \mid E)$

“posterior” → probability that the target is at a particular location, given the sensory evidence

$P(E \mid H)$

“likelihood” → probability of the sensory evidence, given the hypothesis about the target location

$P(E)$

“prior” → probability of the target being at the particular location

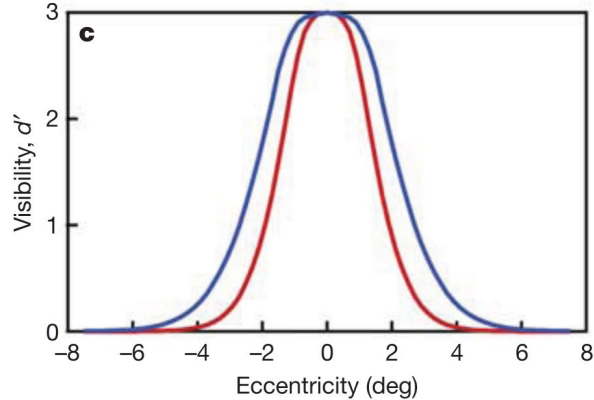


## 2) Where to look next: Bayesian search

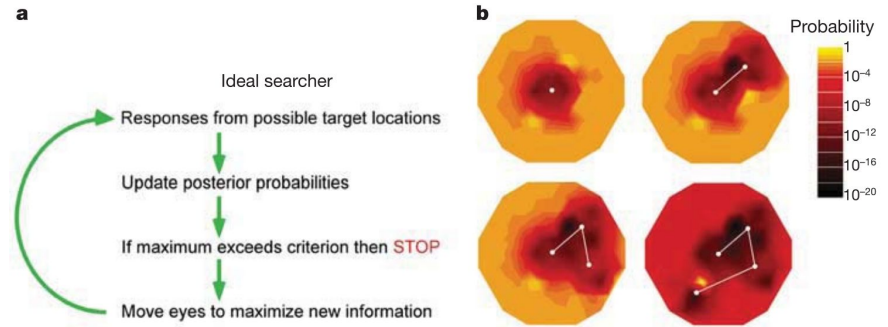
- a) **Compute sensory evidence** for each point in space by multiplying a hypothetical visual response with the visibility function
- b) **Combine the sensory evidence with prior knowledge** using Bayes' law to get the posterior probability of the target for all locations in space
- c) **Check if the criterion for identifying the target is met** by comparing the posterior probabilities to a threshold
- d) **Make a saccade to a new location** by...?

# Modeling human visual search

1) Extract a visibility map









2) Simulate a Bayesian searcher



3) Compare simulated searchers to human data

# Modeling human visual search

Characteristics of the Bayesian searcher:

- () Parallel information processing for all locations in space
- () Perfect integration of information across fixations
- () Uses saccades to actively explore the visual space
  - () 'Inhibition of return' / moderate length saccades
  - () 'Center-of-gravity saccades'
  - () 'Exclusion saccades'