## Models of visual search

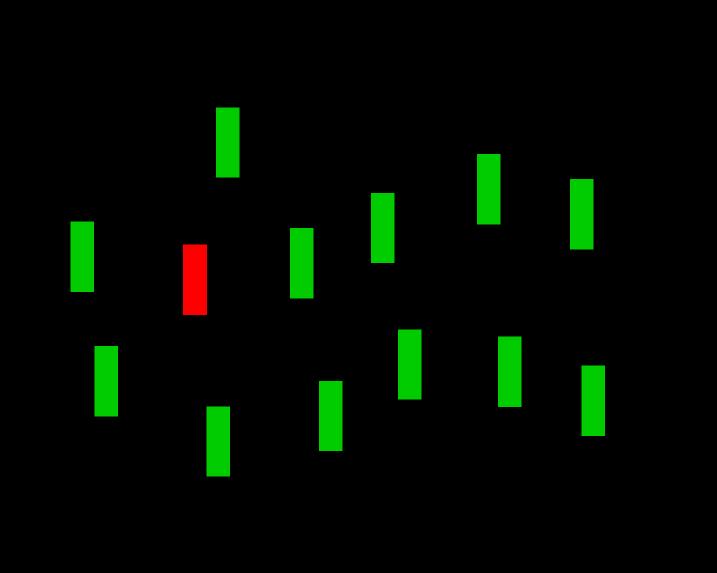
Krista A. Ehinger

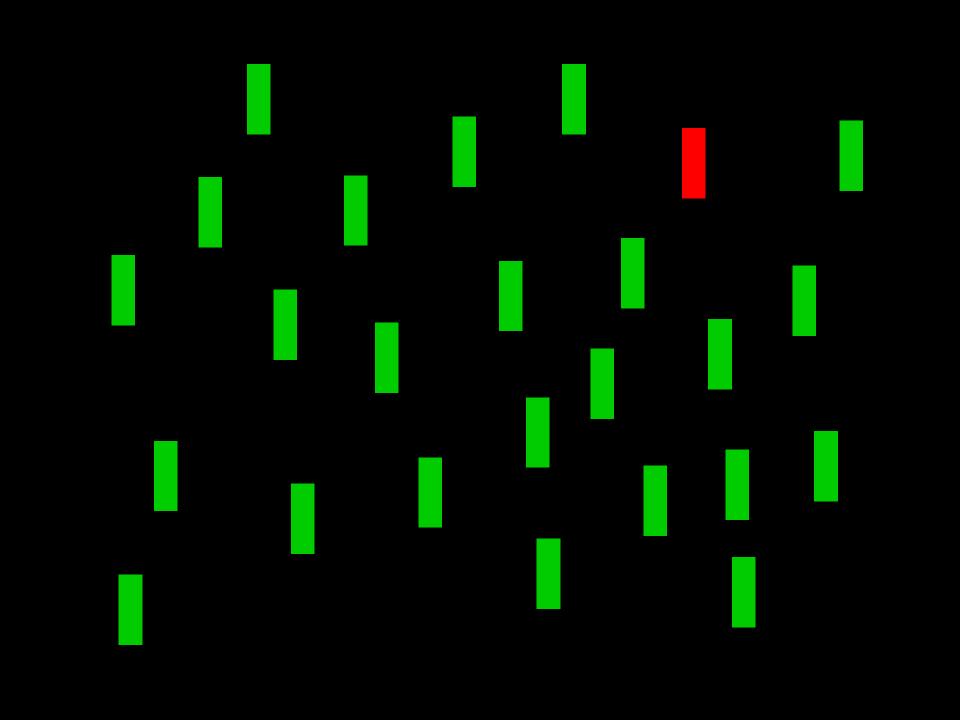
PSYC 6229 Statistical Modelling of Perception and Cognition March 12, 2019

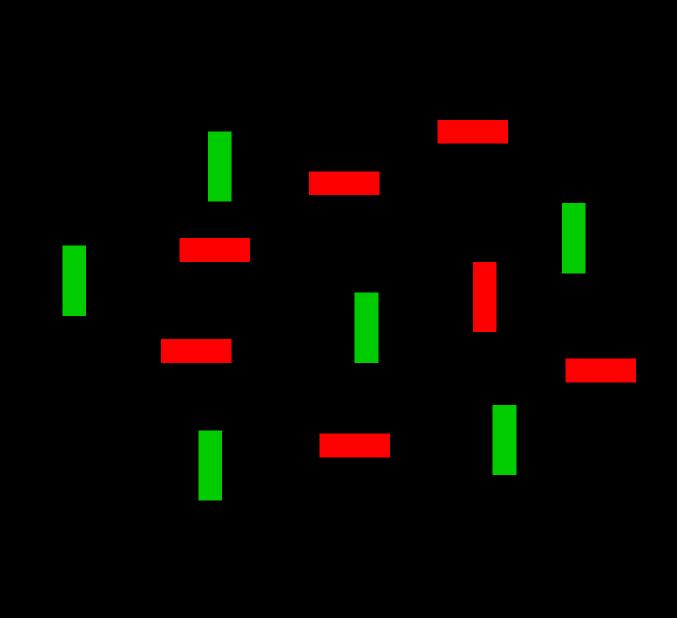
#### Visual search

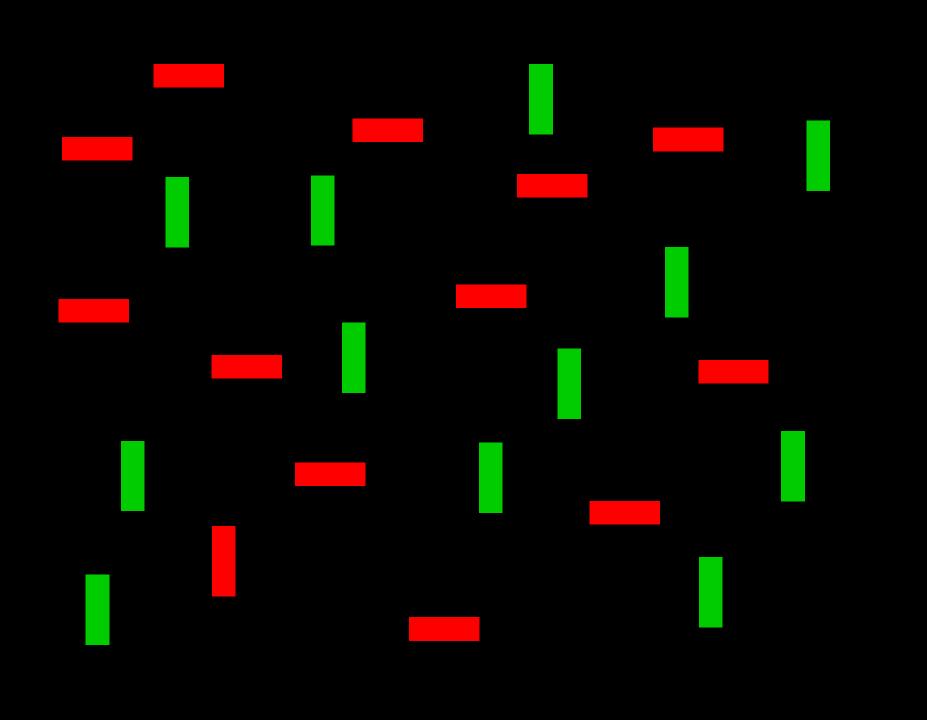
- Task: find a target object in a visual display
- Measure reaction times and/or errors
- Why is it interesting?
  - How easy/hard the search is tells us something about visual encoding – some targets just "pop out" of the display but others don't
  - Important real-world applications: security, medical screening, search and rescue

Clap when you find this:









Clap when you find this:

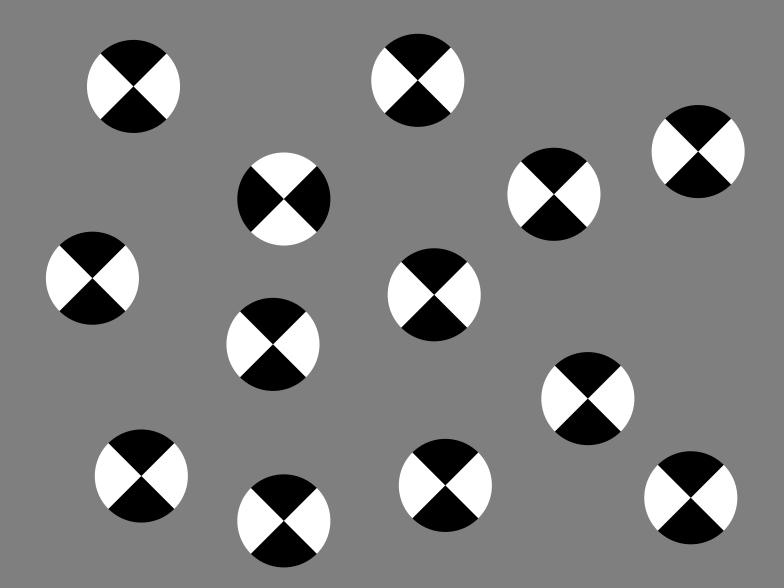
an elephant

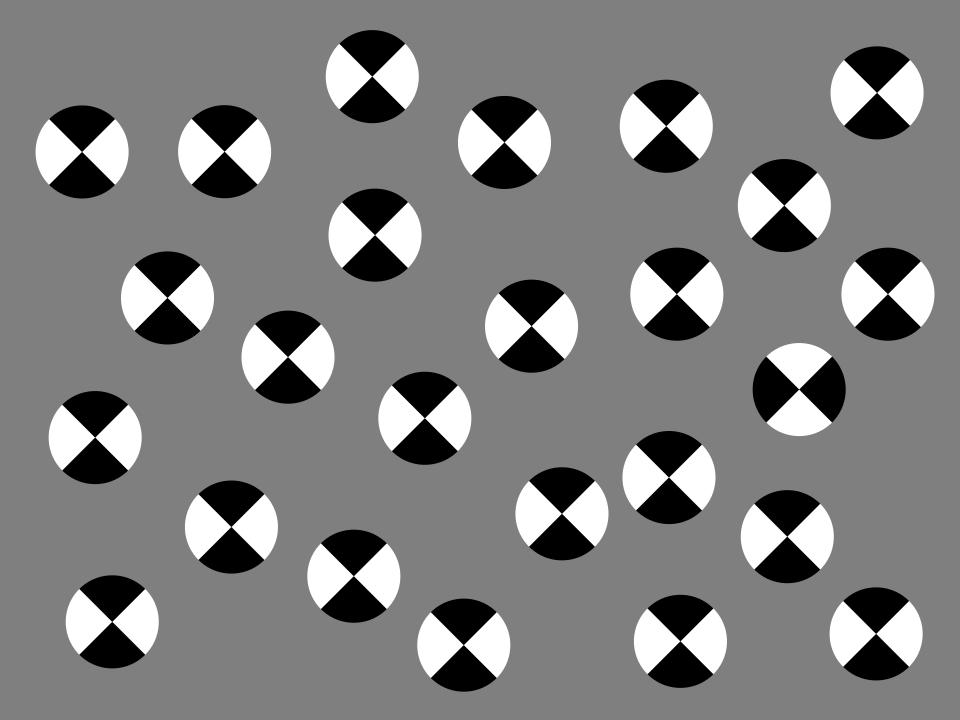




Clap when you find this:



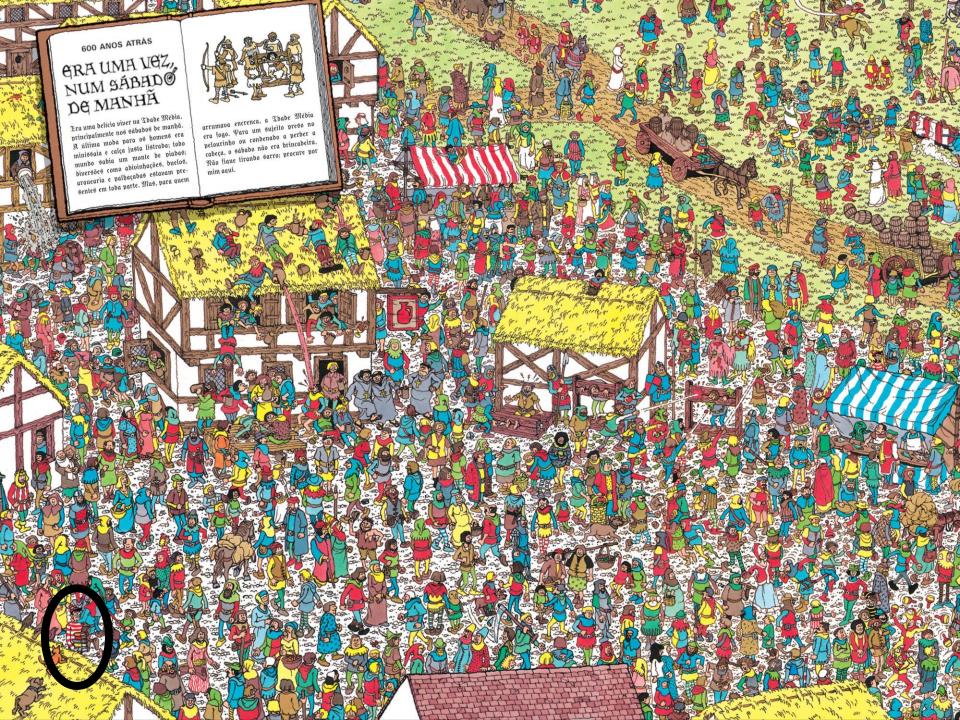




#### Search is harder if...

- The distractors are more similar to the target
- The distractors are more heterogeneous



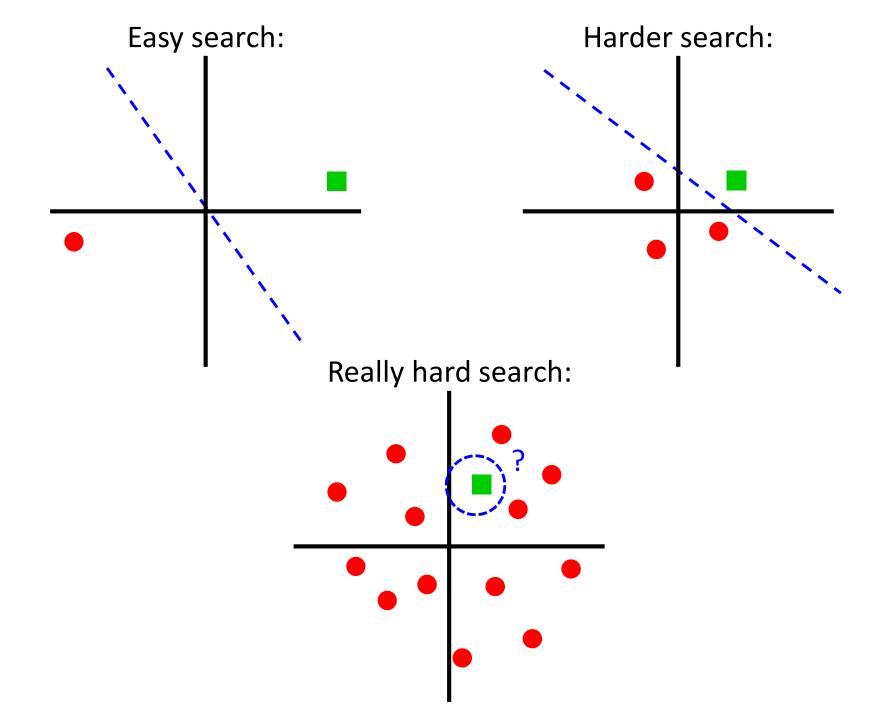




Slide from: R. Rosenholtz

#### Search is harder if...

- The distractors are more similar to the target
- The distractors are more heterogeneous
- The distractor distribution (in some feature space) is less separable from the target distribution (in the same feature space)



## Search efficiency

- Search efficiency = how much slower search becomes as you add more distractors to the display
- Measured as search slope: msec/item
- Simplistically, you can group search tasks into two categories:
  - Efficient = flat slope = search time does not depend on number of distractors
  - Inefficient = steep slope = search time increases with number of distractors

#### 2AFC task

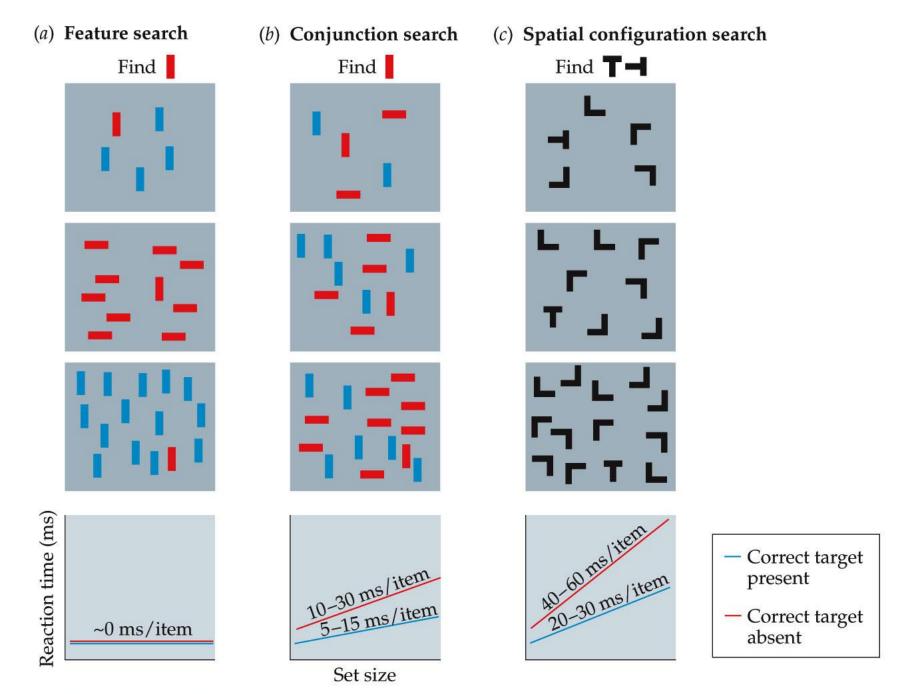
- Task: Is the target present? (yes/no)
- 50% of displays include one target, 50% of displays have no target
- Number of items in display (set size) varies
- Measure response time and/or accuracy

#### Efficient search

Number of distractors doesn't really affect search time "Parallel" – can process the whole visual field more or 10 r less simultaneously 8 7 6 3 Target present Target absent 30 35 45 40 50 Items

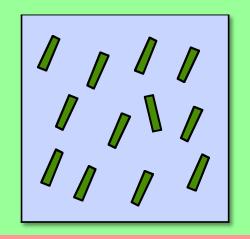
#### Inefficient search

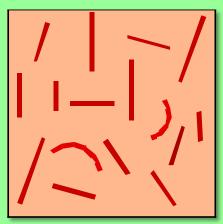
More distractors -> takes longer to find the target "Serial" – it looks like you cannot process all items at 10 ( once Target present Target absent Items



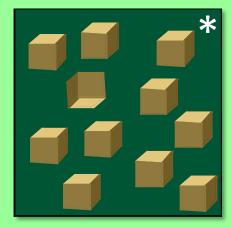
## Search efficiency

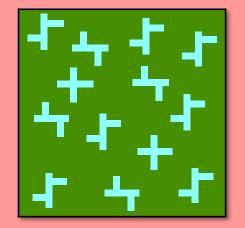
#### Efficient search

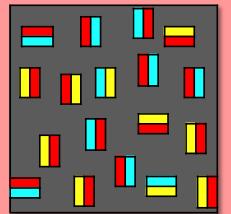


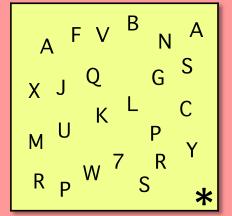


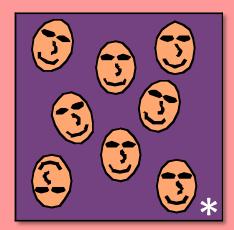












Inefficient search

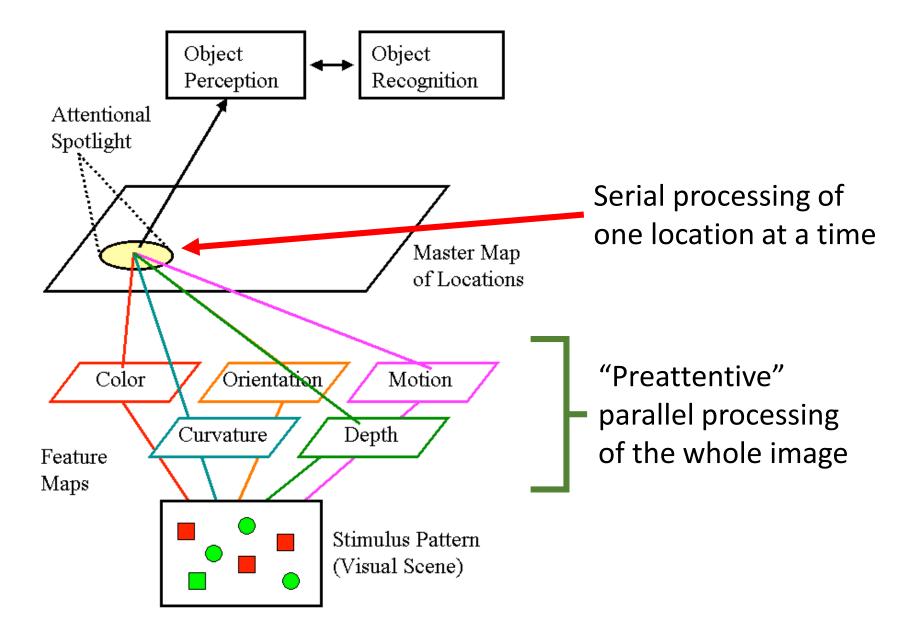
\*with caveats

# Models of visual search

## Model 1: Two-stage processing

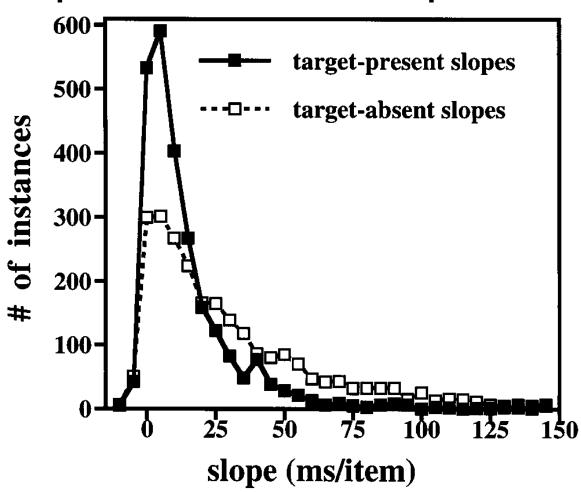
- Examples: Feature Integration Theory (Treisman);
  Guided Search (Wolfe)
- Visual information is processed in two stages
- The first stage processes the whole image in parallel; sometimes this step alone is sufficient to find the target (= efficient search)
- Harder search tasks require a second processing stage which is serial: you move your eyes/attention through the display and process each item (or small groups of items) one at a time (= inefficient search)

#### Feature Integration Theory

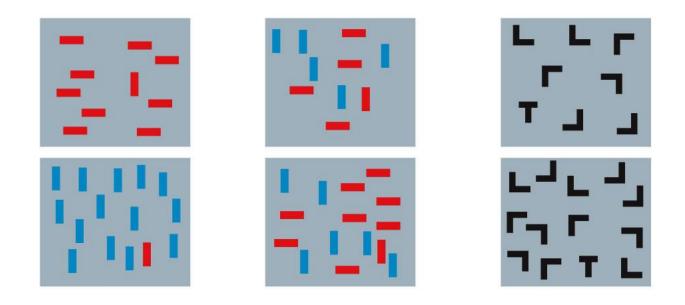


## Serial vs. parallel

#### Slopes from 2500 search experiments



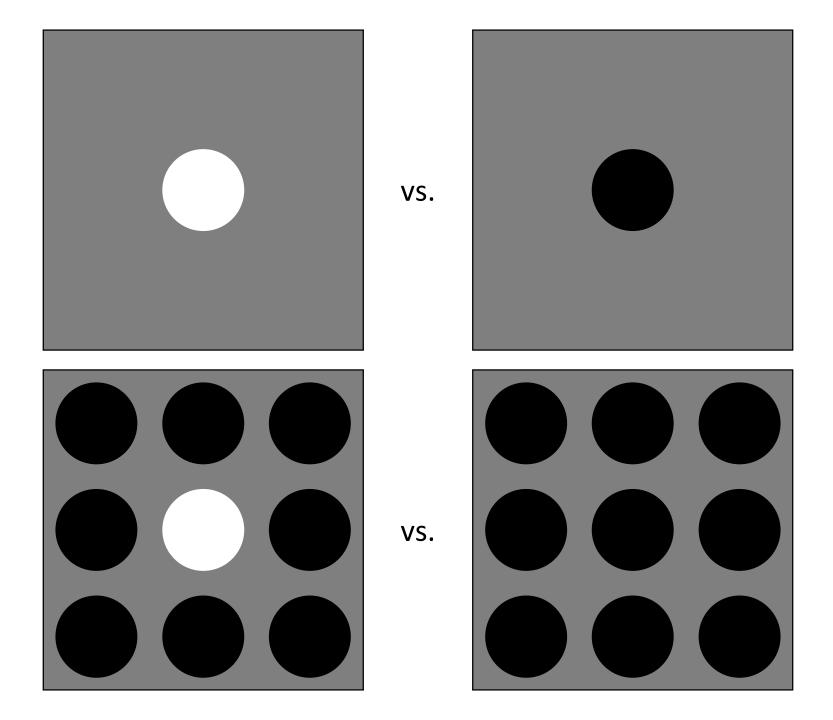
## What affects search efficiency?

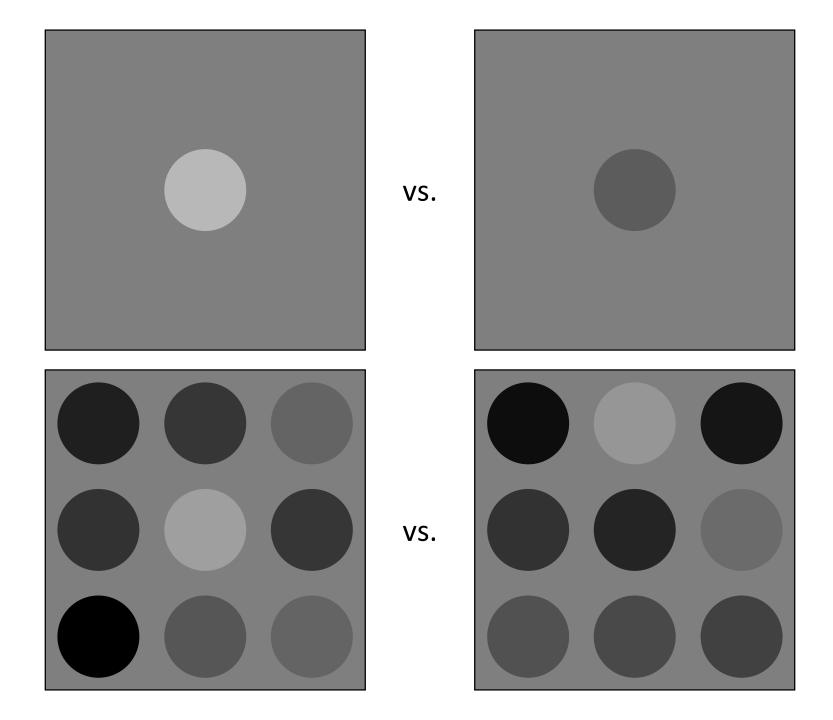


- Presence of a "basic" feature
- Conjunctions are easier than configurations
- Early vs. later stages of visual processing

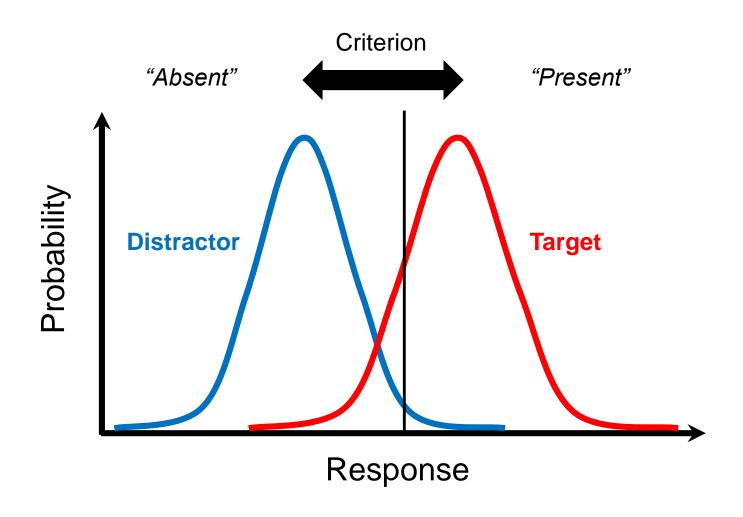
## Model 2: Parallel processing (SDT)

- Examples: Palmer et al. (1993); Verghese (2001)
- Most search tasks involve parallel processing, and increased errors (or slower responses) with set size are a natural consequence of distinguishing between larger sets of items in a signal-detection framework
- If the discrimination is trivially easy, difficulty will not increase much with set size (= efficient search)
- If the discrimination is difficult, set size will have more impact (= inefficient search)



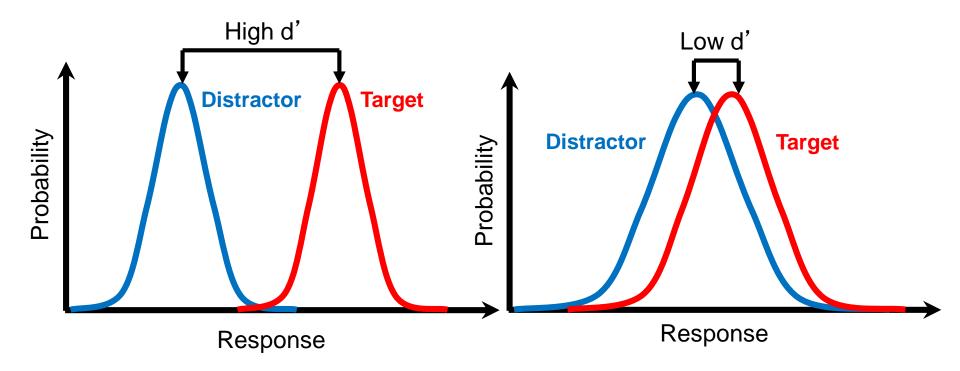


## Signal-detection theory

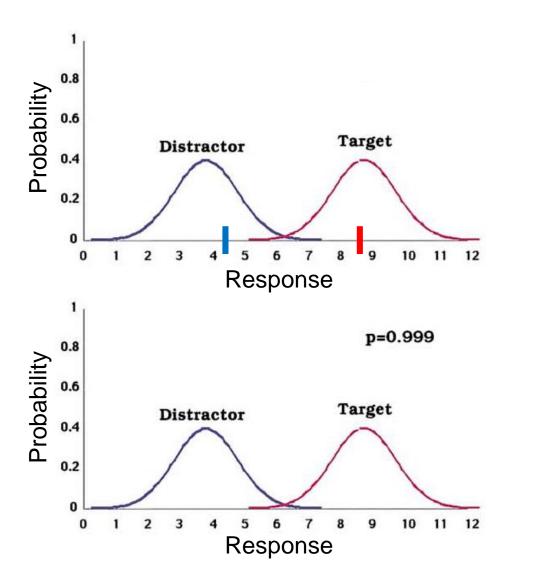


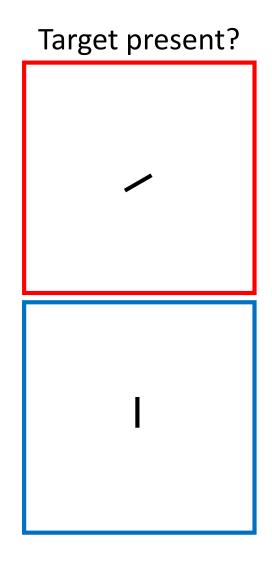
## Discriminability

• D prime (d') = 
$$\frac{\mu_T - \mu_D}{\sqrt{\frac{1}{2}(\sigma_T + \sigma_D)}} = Z(hit) - Z(FA)$$

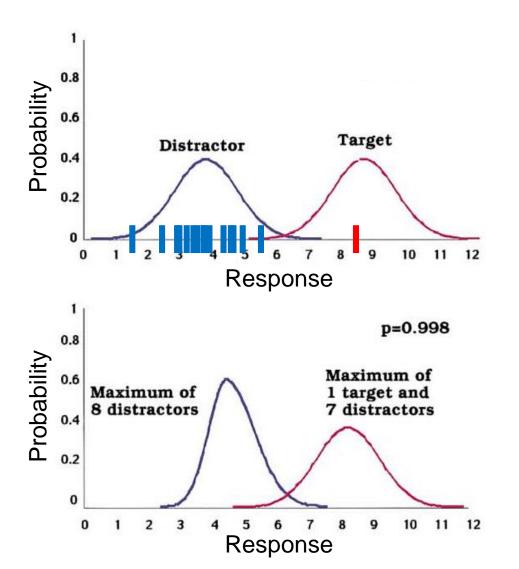


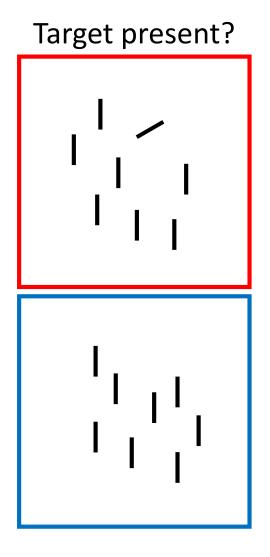
### SDT and search: Maximum Rule

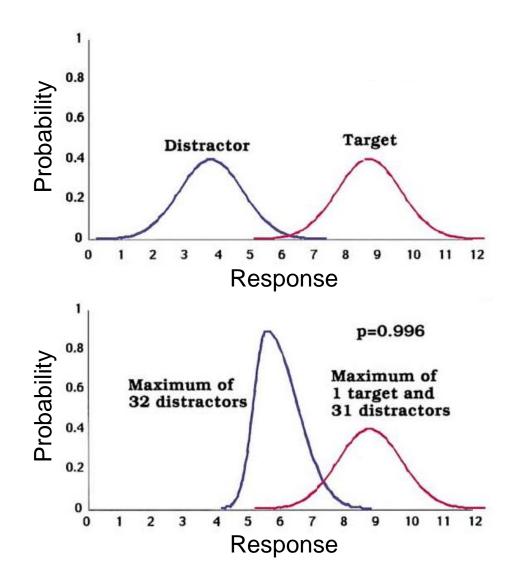


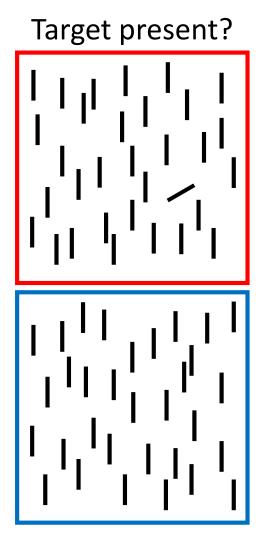


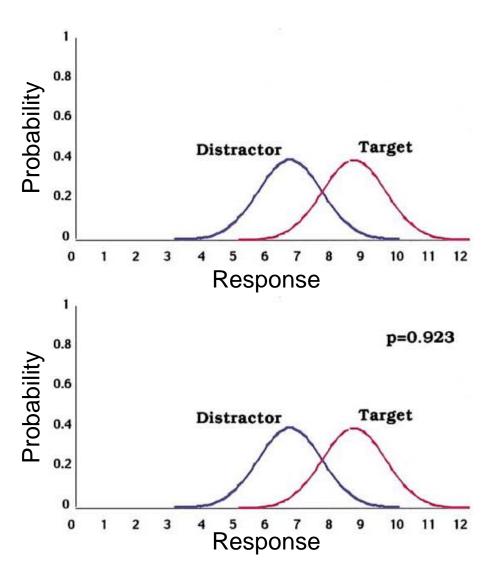
### SDT and search: Maximum Rule

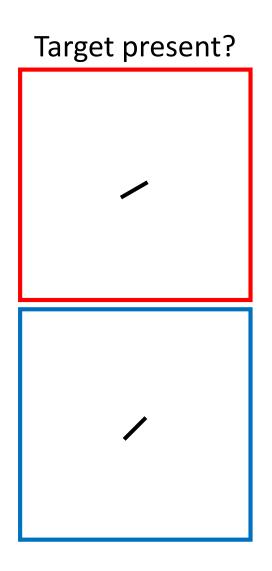


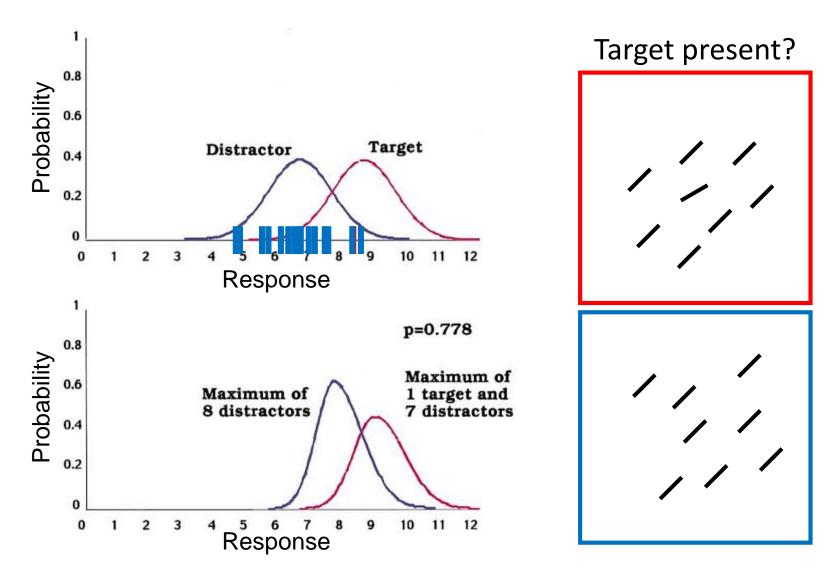


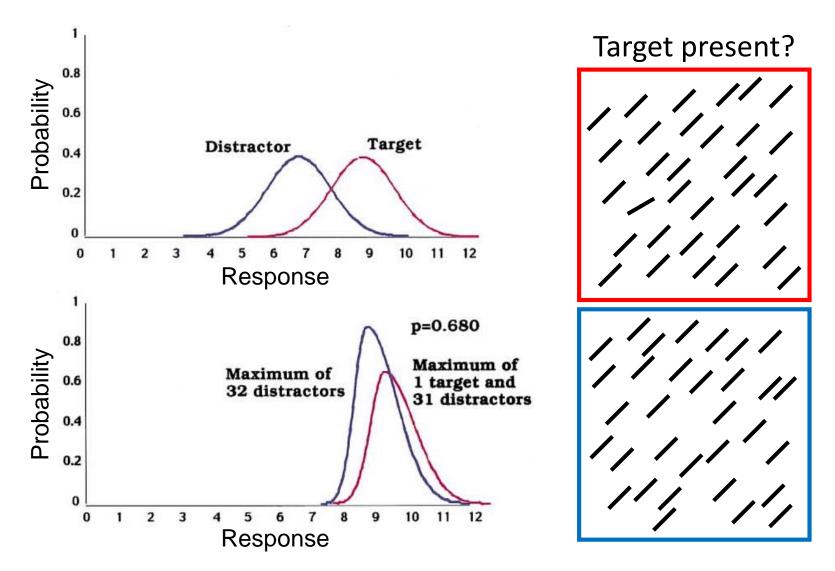




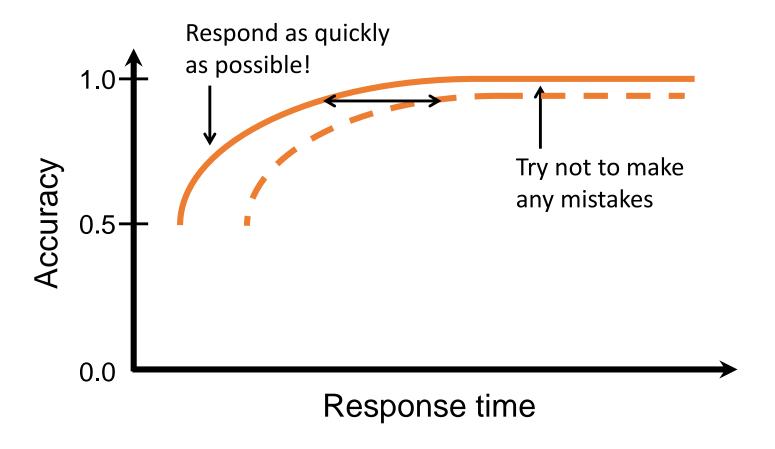




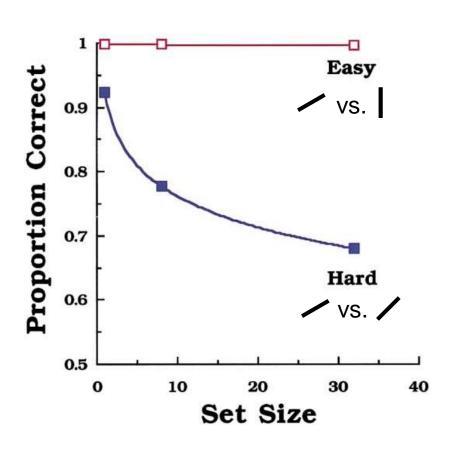


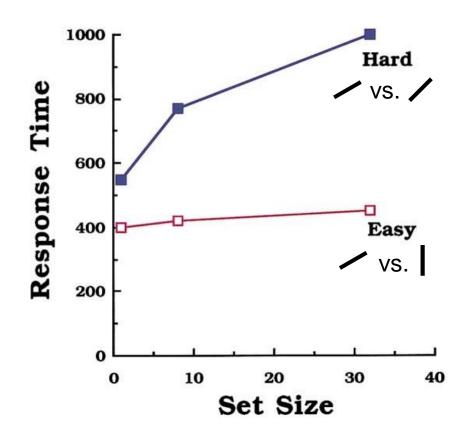


# Speed-accuracy trade-off

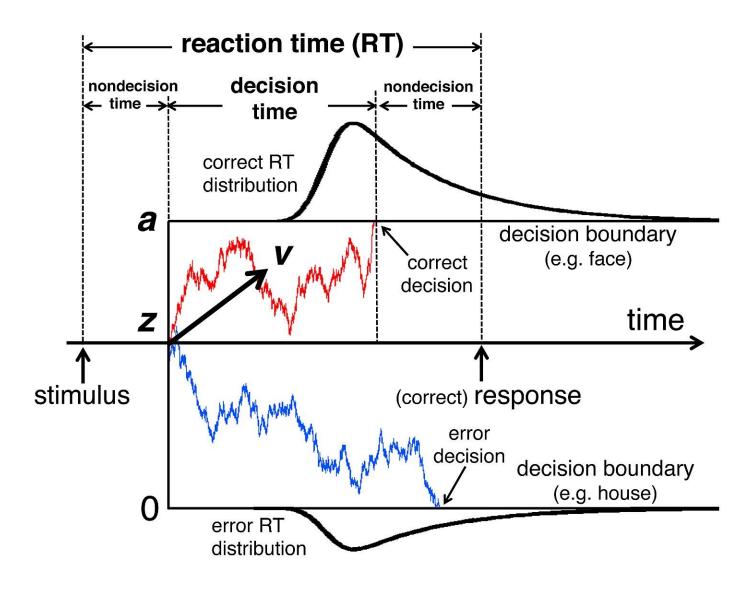


# SDT and response times

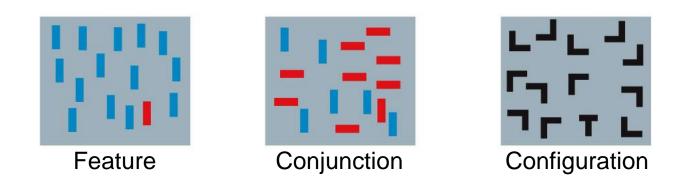




### Drift diffusion model

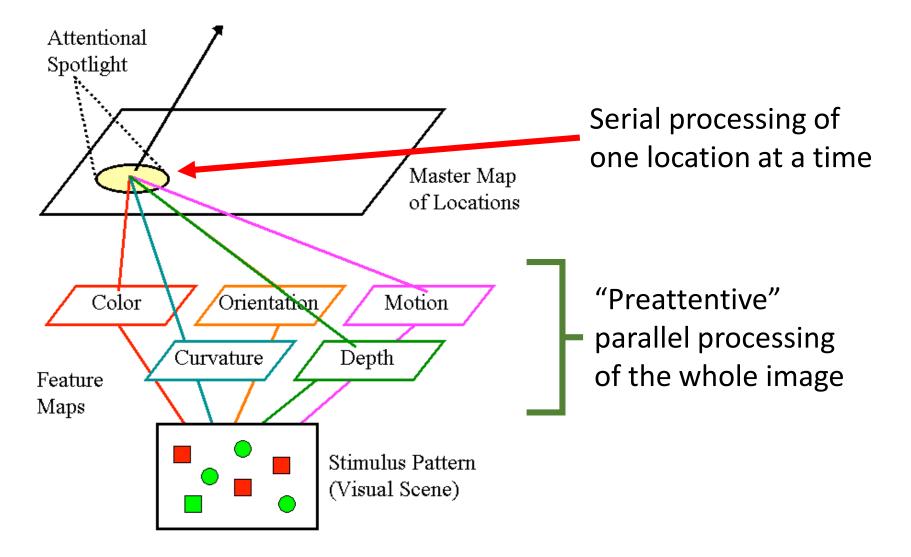


# What affects search efficiency?

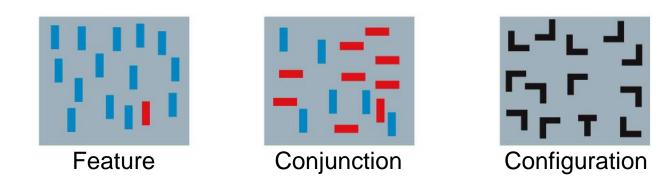


- According to the two stage-model?
- According to the SDT model?

# Two-stage model

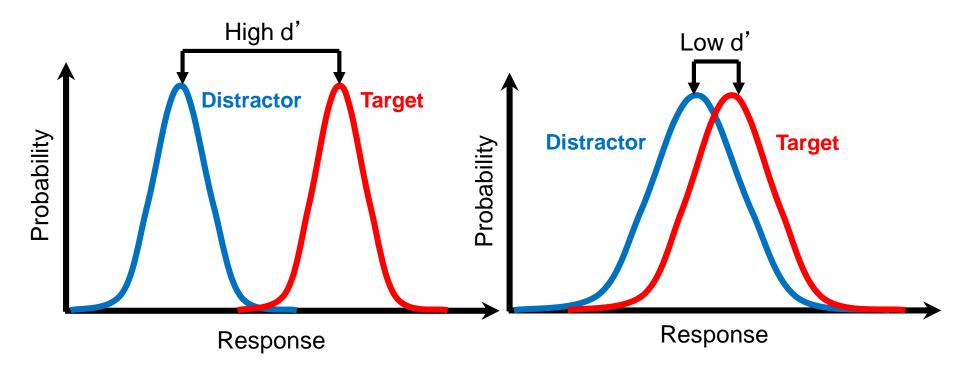


# What affects search efficiency?

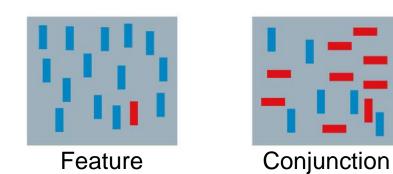


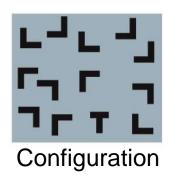
- According to the two stage-model?
  - Efficiency depends on how many preattentive maps must be combined to build the target feature
- According to the SDT model?

### SDT model



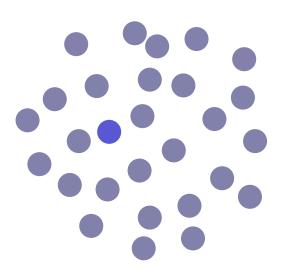
# What affects search efficiency?

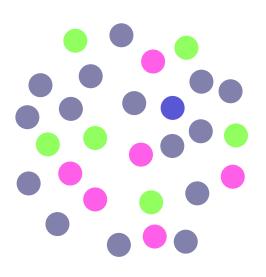


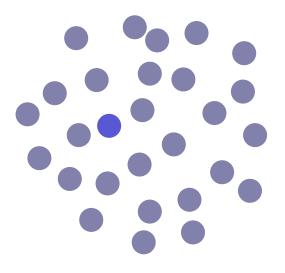


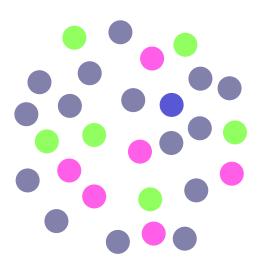
- According to the two stage-model?
  - Efficiency depends on how many preattentive maps must be combined to build the target feature
- According to the SDT model?
  - Efficiency depends on the d' of target-distractor discrimination

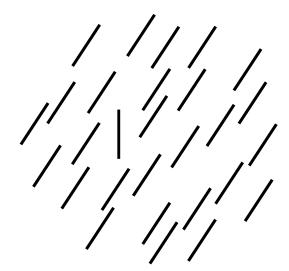
# Example: Homogenous vs. heterogeneous distractors

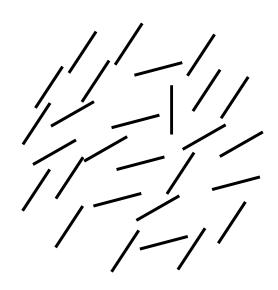


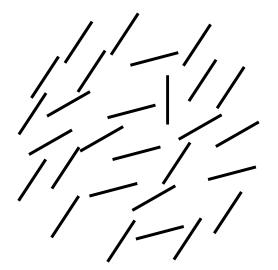


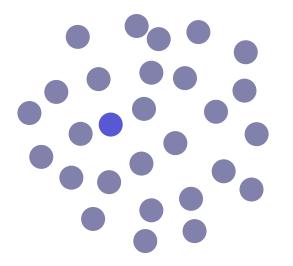


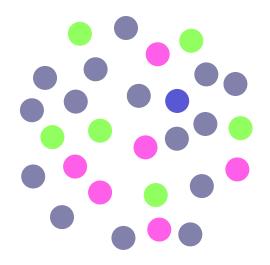




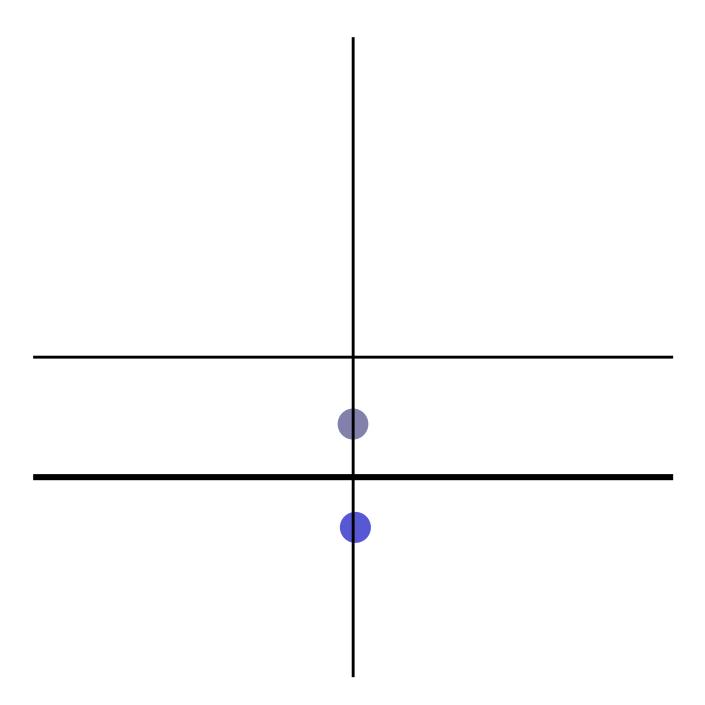






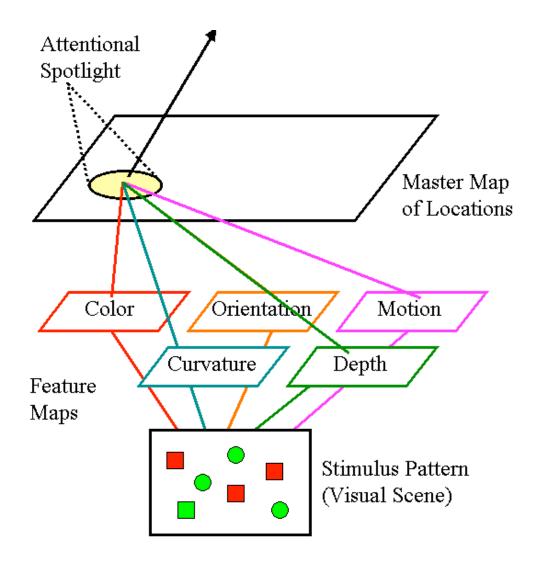


Search is easier when the distractors are homogeneous



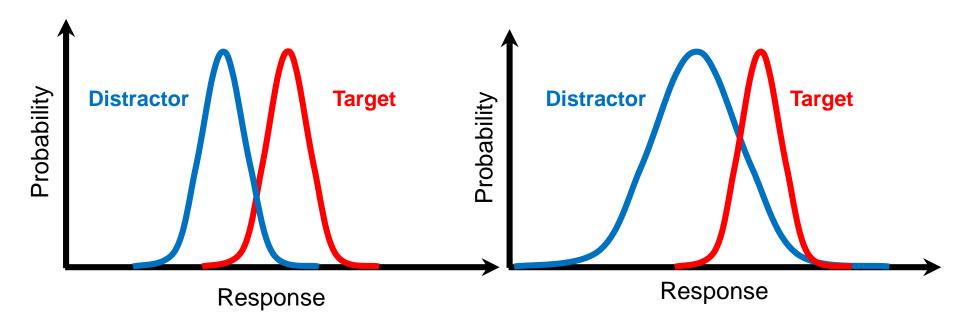


# Two-stage: Distractors



Why would heterogeneous distractors be more difficult?

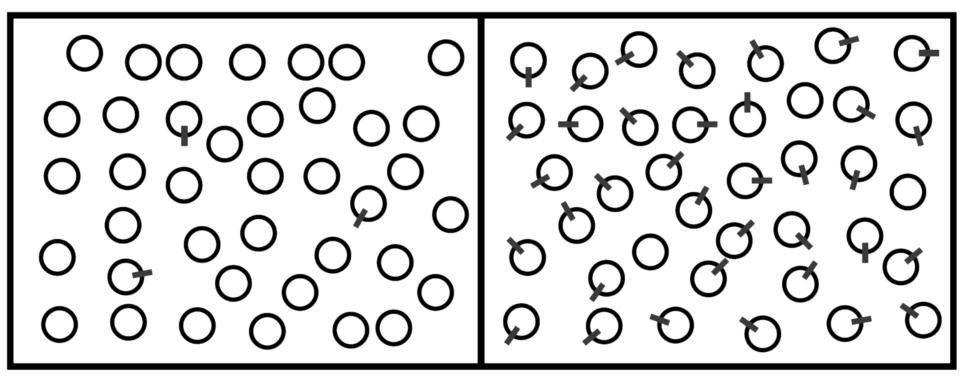
#### SDT: Distractors



What happens as set size increases?

# Example: Search asymmetries

# Search asymmetries

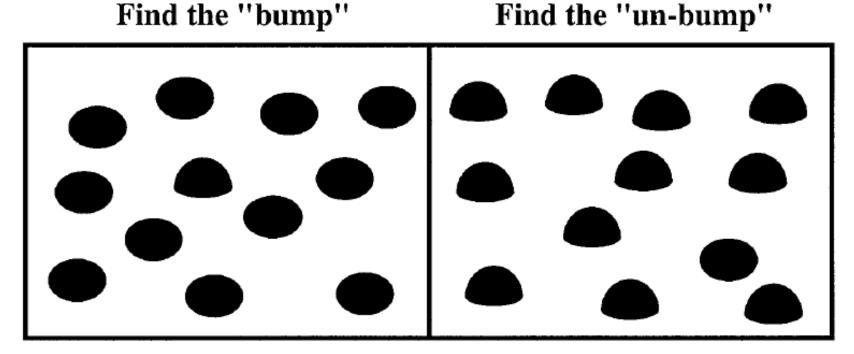


FIND **\Q** Three of them

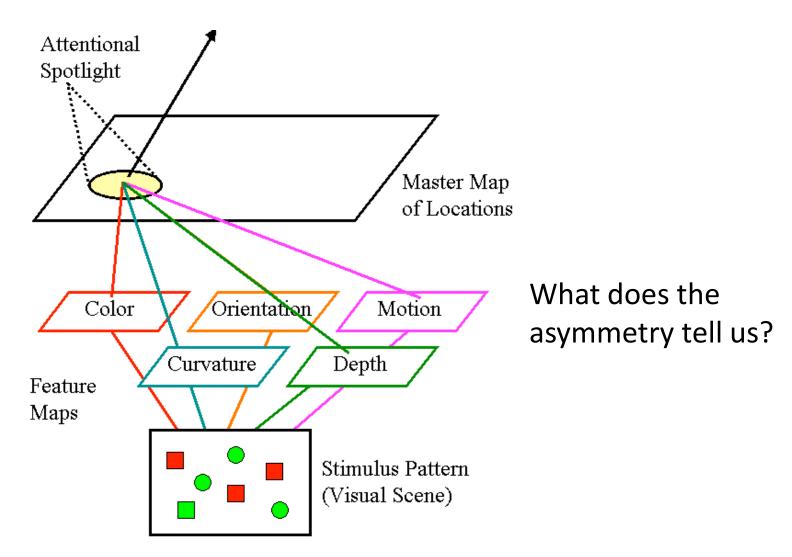
FIND **O** Three of them

# Search asymmetries

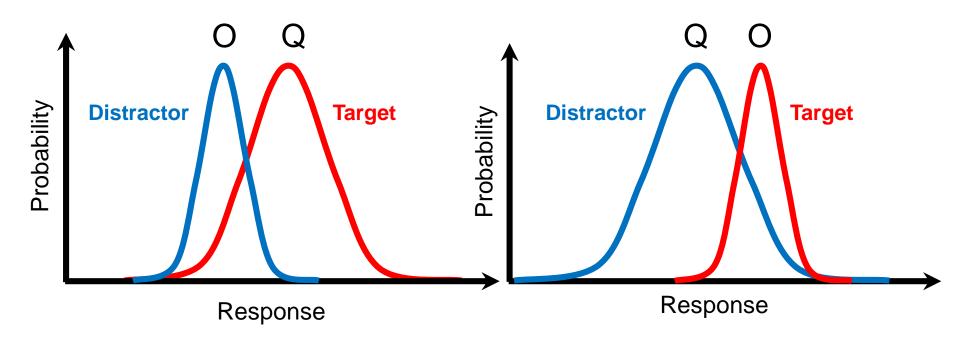
Find the "un-bump"



# Two-stage: Search asymmetries

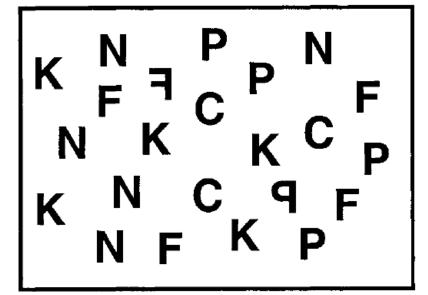


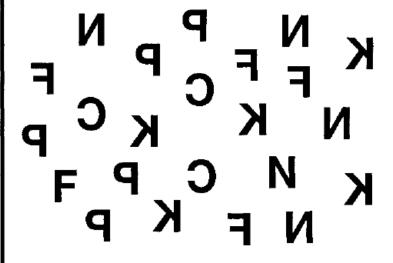
# SDT: Search asymmetries



What happens as set size increases?

# Another search asymmetry





Find the mirror-reversed letters

Find the normal letters

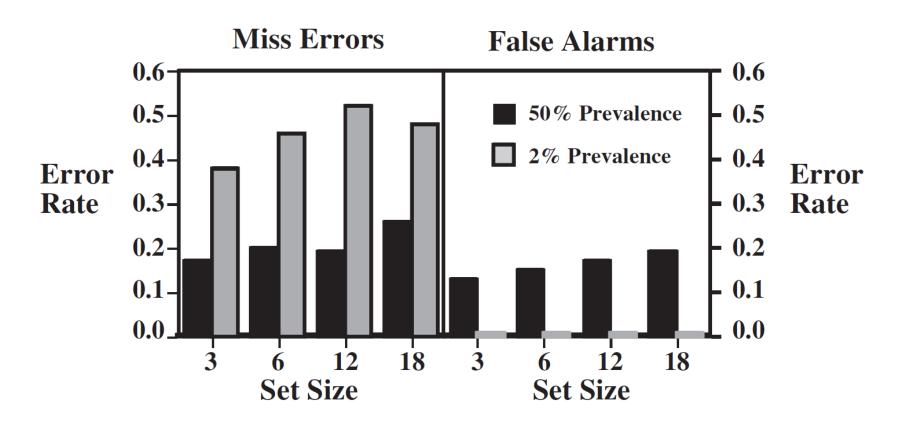
Search for an *unfamiliar* target among familiar distractors is generally faster than the reverse – why?

# Example: Target prevalence

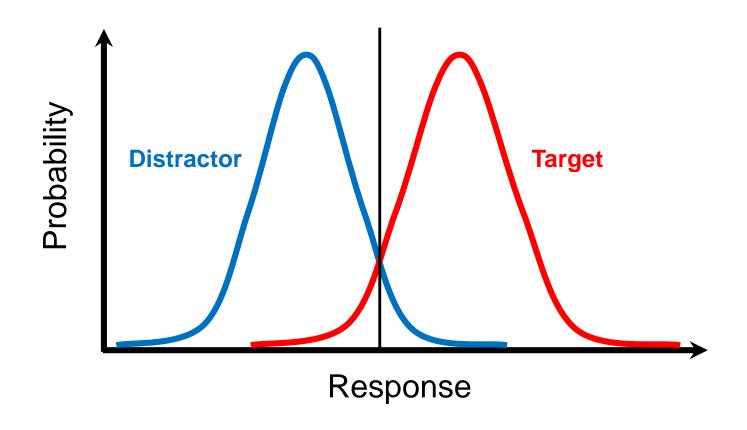
#### Prevalence effects

- In experiments, target is present on 50% of trials
- In real-life search, target is often much rarer:
  - Medical screening
  - Baggage screening
  - Search and rescue
- Does lower target prevalence affect search?

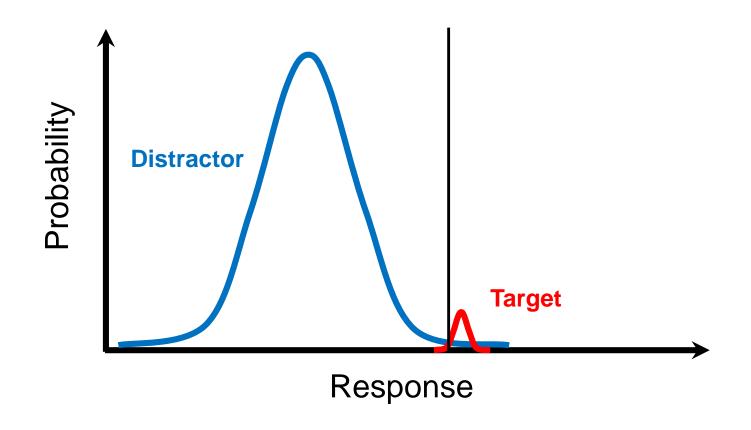
#### Prevalence effects



# SDT: Target prevalence



# SDT: Target prevalence



# Conclusions

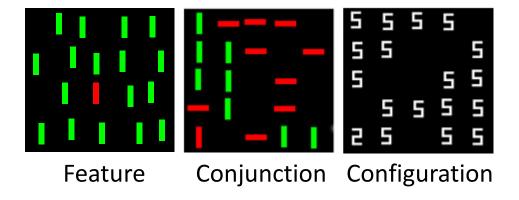
#### Visual search models

- Usually investigated with 2AFC tasks in which one target appears among a number of distractors
- More than one model to explain visual search:
  - Two-stage model
  - Parallel SDT model
- Many findings can be explained by both models
- Comparisons can be difficult because each model focuses on a different measure (RT vs. errors)
- Aspects of both models may be needed to fully explain search

# R Demonstration

# Demo 1: Search slopes

- Experiment data from: <a href="http://search.bwh.harvard.edu/new/data\_set\_files.html">http://search.bwh.harvard.edu/new/data\_set\_files.html</a>
- Three visual search tasks:



# Demo 2: SDT simulation

#### Discussion

- What are some advantages / disadvantages to each model?
- How could you interpret a negative search slope according to each model?
- If two search tasks have the same search slopes but different intercepts, what does that mean?
- What experiments could distinguish between the two models?

#### References

- Eckstein, M. P., Thomas, J. P., Palmer, J., & Shimozaki, S. S. (2000). A signal detection model predicts the effects of set size on visual search accuracy for feature, conjunction, triple conjunction, and disjunction displays. *Perception & Psychophysics*, 62(3), 425-451.
- Palmer, J. (1994). Set-size effects in visual search: The effect of attention is independent of the stimulus for simple tasks. *Vision Research*, 34(13), 1703-1721.
- Palmer, J. (1998). Attentional effects in visual search: Relating search accuracy and search time. In *Visual Attention*, R. Wright, ed. (New York: Oxford University Press), pp. 295–306.
- Palmer, J., Ames, C. T., & Lindsey, D. T. (1993). Measuring the effect of attention on simple visual search. *Journal of Experimental Psychology: Human Perception and Performance*, 19(1), 108-130.
- Verghese, P. (2001). Visual search and attention: A signal detection theory approach. *Neuron*, *31*, 523-535.