

# Feature Transitions with Saccadic Search: Size, Color, and Orientation Are Not Alike

*Stella X. Yu*

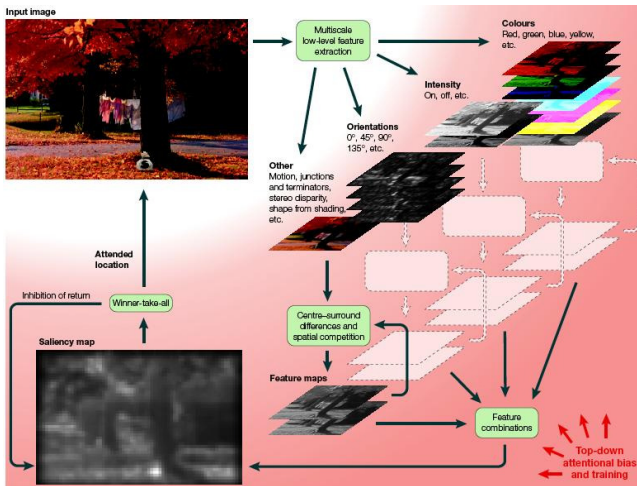


Computer Science

Boston College

NIPS 2010

# Size, Color, and Orientation: Elementary Features



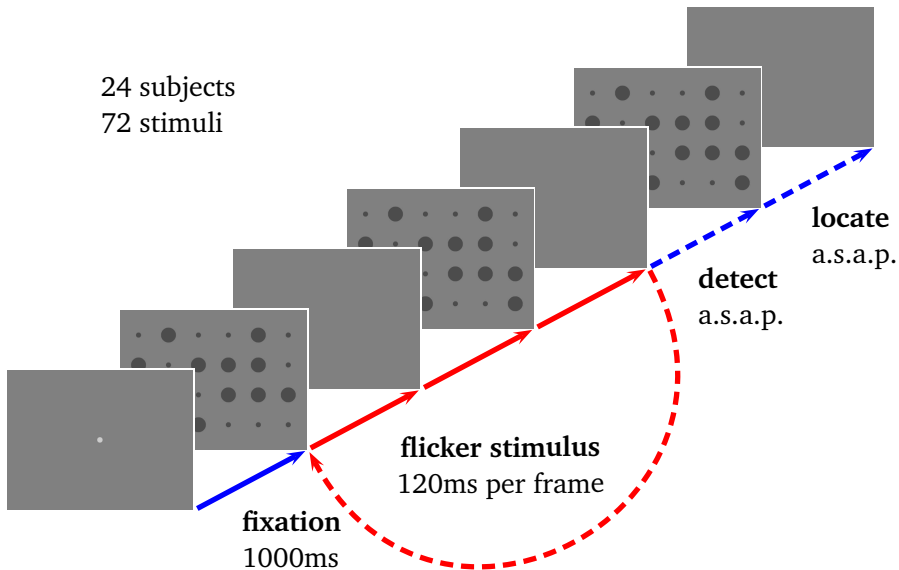
Do these elementary features with alike parallel local detections in space have alike serial deployment of attention over time?

## Saccadic Search Depends on Feature/Filter Type?




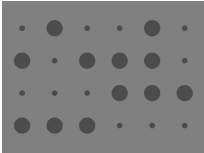
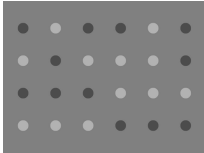
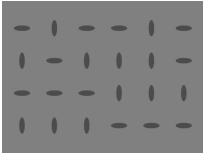
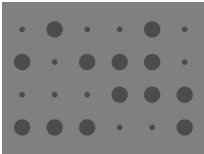
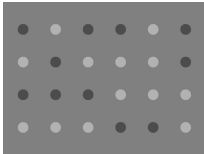
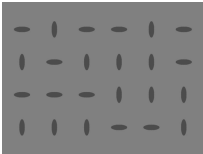


If attention depends only on the spatial map of responses regardless of which filters produce them, there should be little differences in the saccadic search for a flickering target among identically laid out disks rendered in comparable attributes of different features.

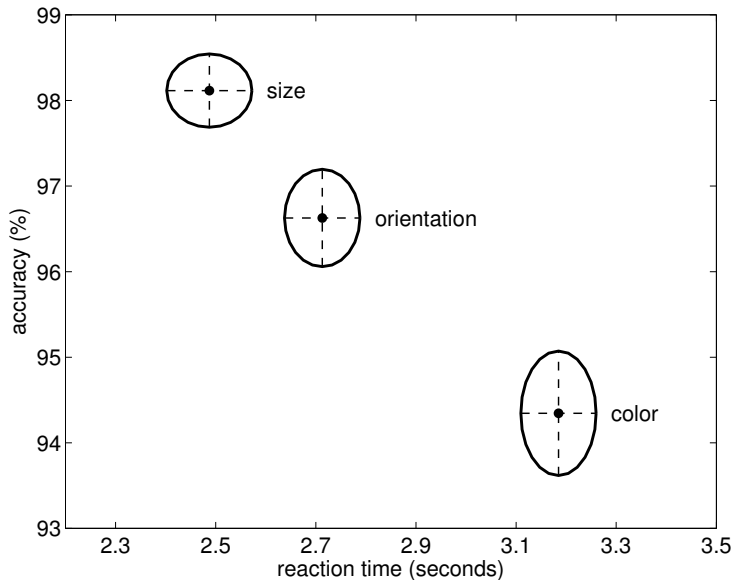
# Change Detection and Localization Experiment



# Stimuli: Different Features in Identical Layout

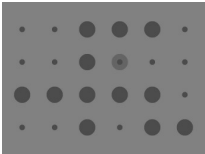
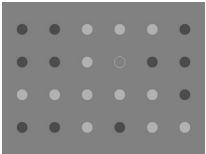
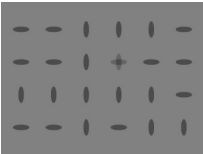
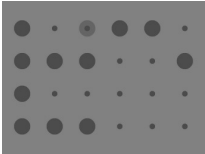
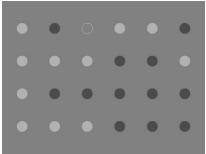
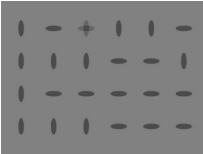
flicker stimuli	size	color	orientation
identical layout	 1    2	 1    2	 1    2
1st image			
2nd image			

## Overall Saccadic Search Performance



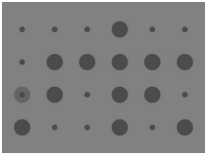
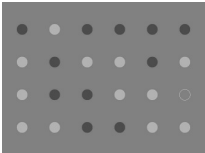
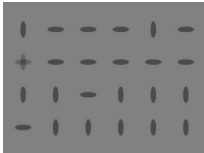
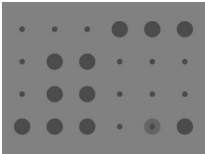
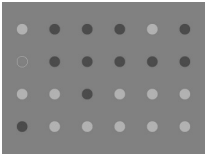
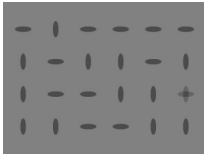
# Impact of Spatial Context on Detection: Common

Change is easier to detect in a homogeneous neighbourhood.

commonly best layout	 ( 100%, 1.36 s )	 ( 100%, 1.52 s )	 ( 100%, 1.20 s )
commonly worst layout	 ( 83%, 11.98 s )	 ( 78%, 4.18 s )	 ( 78%, 5.88 s )

# Impact of Spatial Context on Detection: Specific

... only if the items are large for size, or collinear for orientation

dimension specific best layout	 ( 100%,2.34 s ) ( 83%,5.44 s ) ( 83%,6.47 s )	 ( 100%,3.08 s ) ( 100%,2.09 s ) ( 80%,2.35 s )	 ( 94%,3.37 s ) ( 78%,4.08 s ) ( 94%,2.69 s )
dimension specific worst layout	 ( 90%,2.10 s ) ( 100%,2.65 s ) ( 100%,2.59 s )	 ( 94%,3.37 s ) ( 78%,4.08 s ) ( 94%,2.69 s )	 ( 100%,3.08 s ) ( 100%,2.09 s ) ( 80%,2.35 s )



# Neighbourhood Uniformity: $f$ -numbers

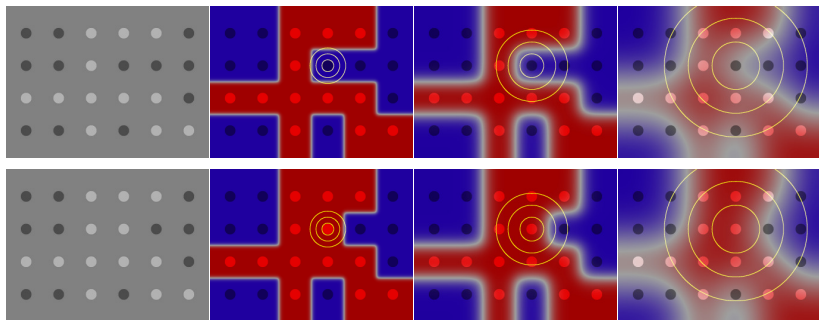
$$f_0(i) = \begin{cases} 0, & \text{no disk at } \text{loc}(i) \\ -1, & \text{disk type 1 at } \text{loc}(i) \\ 1, & \text{disk type 2 at } \text{loc}(i) \end{cases}, \quad f_\sigma(i) = \frac{\sum_j f_0(j) G(\text{dist}(i,j); \sigma)}{\sum_j G(\text{dist}(i,j); \sigma)}$$

$f_0$

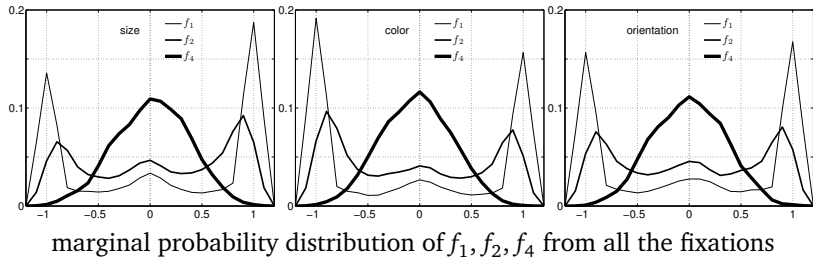
$f_1$

$f_2$

$f_4$



# Fixational Bias in the Spatial and Featural Domains





















## 1. common spatial bias

- ▶ fixating disks instead of empty spaces
- ▶ fixating disks separating large uniform groups

## 2. dimension-specific attribute bias

- ▶ large for size, black for color, vertical for orientation

## Saccadic Bias in the Featural Domain

size				color				orientation						
$\pi_a$	$P_{b a}$				$\pi_a$	$P_{b a}$				$\pi_a$	$P_{b a}$			
.43		.40	.10	.50	.52		.53	.09	.38	.46		.44	.08	.48
.04		.36	.15	.49	.04		.33	.16	.51	.04		.44	.12	.44
.53		.37	.12	.51	.44		.48	.09	.43	.50		.43	.10	.47
attractive bias				repulsive bias				reversible bias						

- ▶  $\pi_a$  is the overall probability of visiting feature  $a$
- ▶  $P_{b|a}$  is the probability of saccading to feature  $b$  at feature  $a$

## Summary

- ▶ That saccadic search is most effective in size and worst in color cannot be due to their alike local detectors in space, but is due to their selective biases in the spatial and featural domains over time.
- ▶ Focusing on the large group essentially cuts down the search space by half, whereas excursion into the white from the primary black group only hurts the spatial efficiency of search.

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