



Detecting Shapes in Raven's Matrices

By: Jacqueline Mok



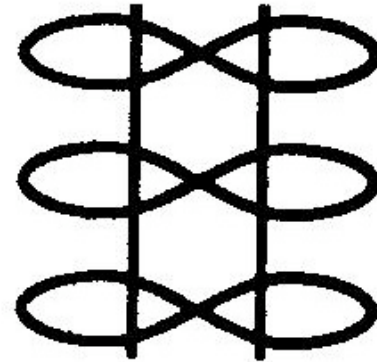
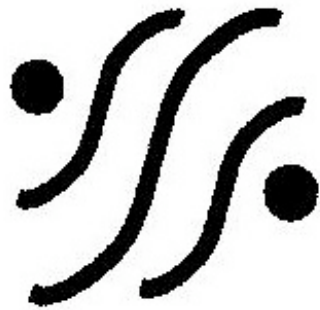
Outline

- Motivation
- Problem
- Techniques used
 - SIFT
 - Hough transform
 - Affine fitting
 - Clustering
- Adaptation to Rasmussen Model
- Conclusions
 - Future problems



Motivation

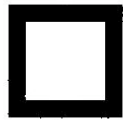
- Learn to recognize objects



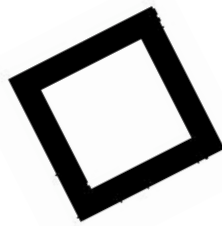


Motivation

- Computers are not THAT intelligent



= square



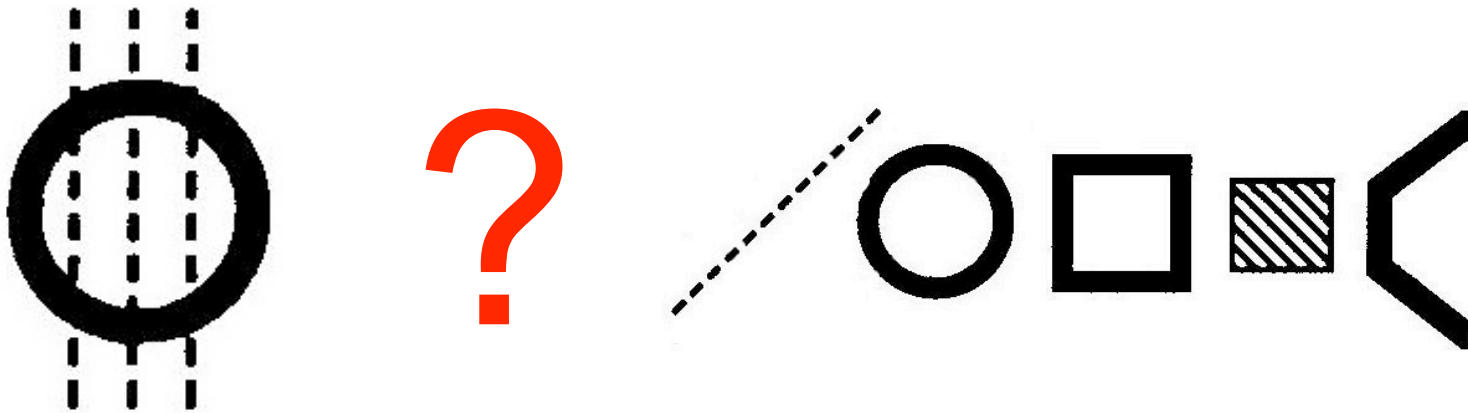
!= square

= object2



Problem

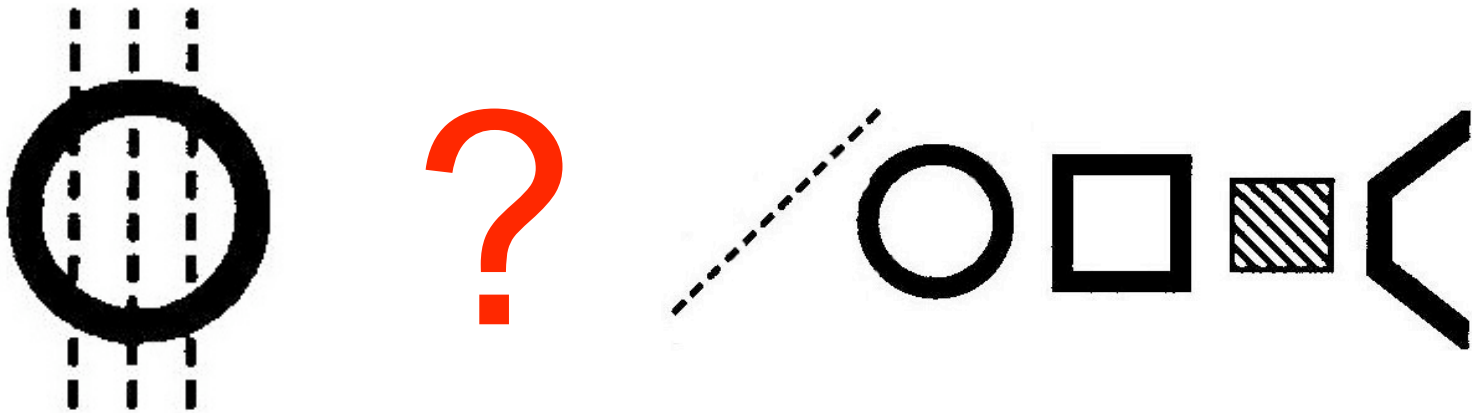
- Given a test image, is it possible to determine whether or not a model exists in the test image? How many times?





Problem

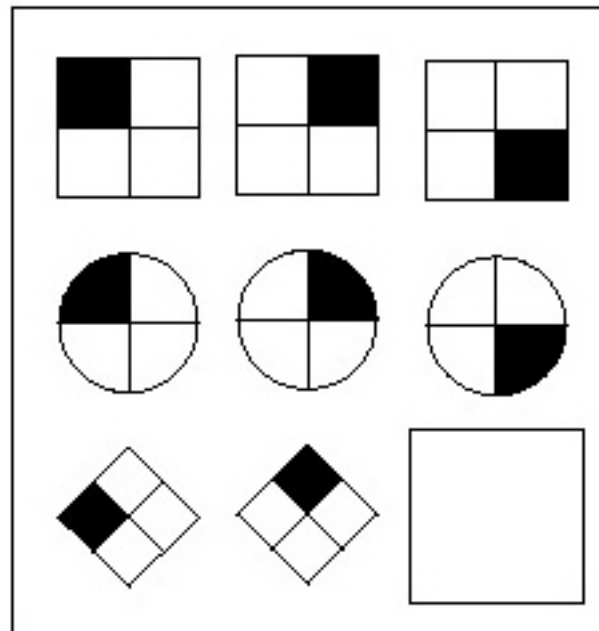
- What is the best (x,y) location, scale and orientation from a model to a given test image?





Problem

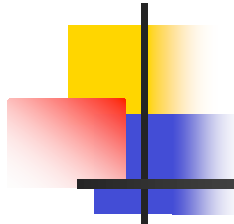
- Used Raven's Matrices





Techniques used

- SIFT (Lowe, 2004)
- Hough Transform
- Affine fitting
- Clustering

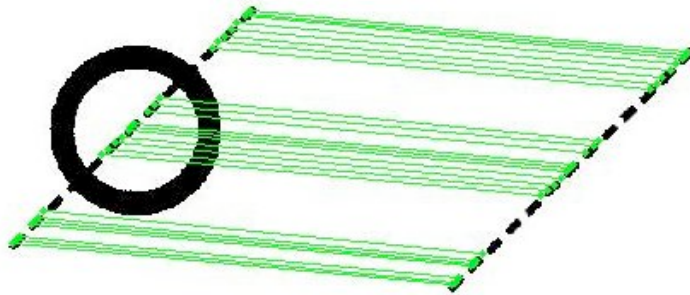
The logo for SIFT (Scale Invariant Feature Transform) features a stylized graphic on the left consisting of overlapping colored squares (yellow, red, blue) and a black crosshair. To the right of this graphic, the word "SIFT" is written in a large, blue, sans-serif font.

SIFT

- Scale Invariant Feature Transform
- David Lowe, 2004
- Match keypoints between test and model
- Provides location, scale, orientation of keypoints



SIFT

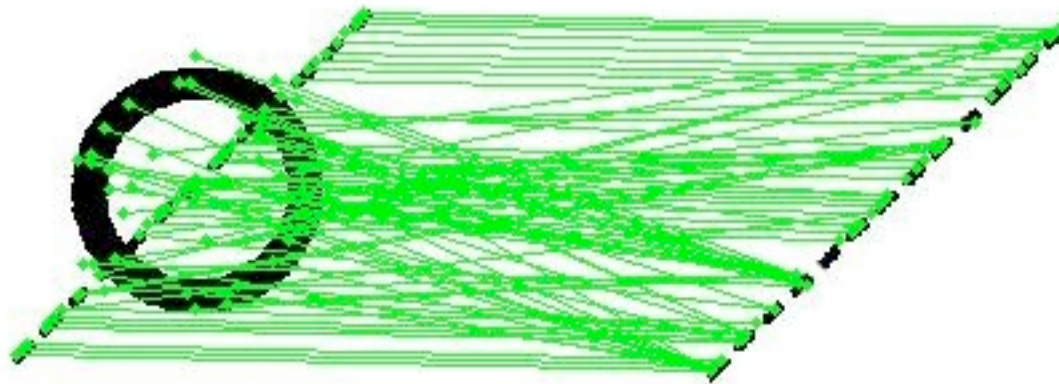


`siftmatch()`,
`plotsiftmatches()`

- Andrea Vedaldi, SIFT for Matlab
- `sift()`
 - Frames: $4 \times M$ matrix
 - Descriptors: 128-D vector

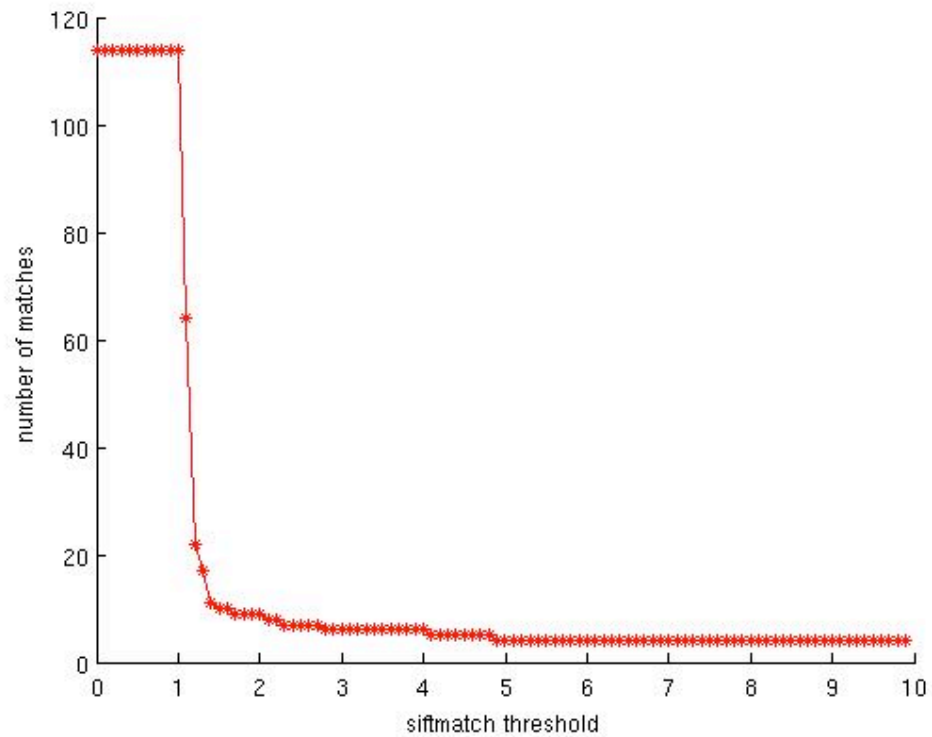
SIFT

```
[frames_test, test_desc] = sift(test, 'Threshold', THRESHOLD, ...  
    'EdgeThreshold', EDGE_THRESHOLD, 'BoundaryPoint', 0);  
matches = siftmatch(test_desc, model_desc, SIFTMATCH_THRES);  
plotmatches(test, model, frames_test, frames_model, matches);
```





SIFT



Siftmatch()



Hough Transform

- Feature extraction technique
- Image analysis, computer vision, digital image processing
- Voting procedure



Hough Transform

- Scene A: knife, blood, person on floor
 - Knife: kitchen, crime, camping
 - Blood: hospital, nose bleed, crime
 - Person on floor: sleeping, star gazing, crime
- Winner: Crime



Hough Transform

- Input: keypoint matches
- Create bins in Hough space
 - (x,y) location, scale, orientation
- Output:
 - 'winning' bin (most votes)
 - No winner (model DNE)
- Multiple winners



Hough Transform

$$\Delta scale = \frac{test_scale}{model_scale}$$

$$\Delta\theta = test_orientation - model_orientation$$

$$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} \Delta scale \times \cos(\Delta\theta) & -\sin(\Delta\theta) & x_t \\ \sin(\Delta\theta) & \Delta scale \times \cos(\Delta\theta) & y_t \\ 0 & 0 & 1 \end{bmatrix} \times \begin{bmatrix} -x_m + \frac{model_width}{2} \\ -y_m + \frac{model_height}{2} \\ 1 \end{bmatrix}$$



Hough Transform

- Vote 2 closest bins in each dimension
- Total 16 bins
- Hash table, serialize indices
- Track winning bin



Affine Fitting

$$\begin{bmatrix} x & y & 0 & 0 & 1 & 0 \\ 0 & 0 & x & y & 0 & 1 \\ & & \dots & & & \\ & & \dots & & & \end{bmatrix} \times \begin{bmatrix} m_1 \\ m_2 \\ m_3 \\ m_4 \\ l_x \\ l_y \end{bmatrix} = \begin{bmatrix} u \\ v \\ \vdots \end{bmatrix}, m = \begin{bmatrix} m_1 \\ m_2 \\ m_3 \\ m_4 \\ l_x \\ l_y \end{bmatrix}$$

(Lowe, 2004)

- At least 3 distinct model points mapped to test image

Affine Fitting



Test image



Model

dashed_line2.jpg located at approximately (156.8097,156.8097)

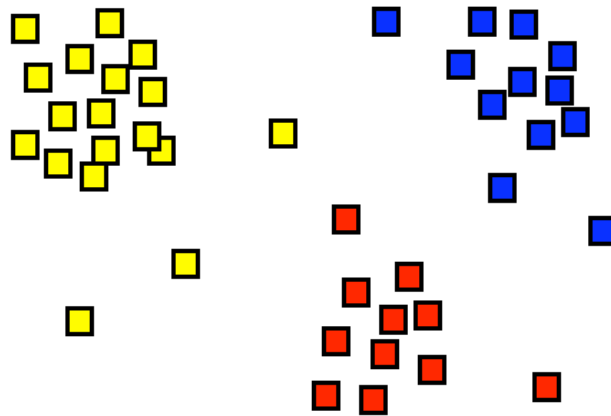
scale: 1.0062 times the model

orientation: 3.1173 radians



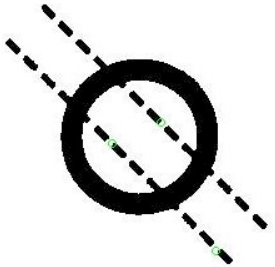
Clustering

- Unsupervised classification of patterns (observations, data, feature vectors) into groups (clusters) (Jain et al., 1999)



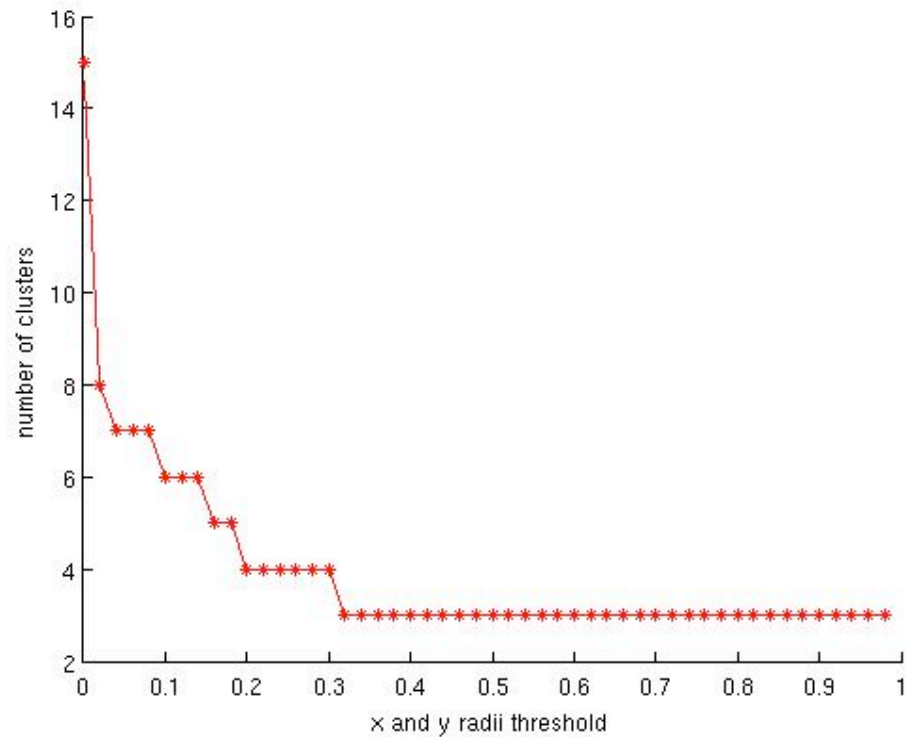


Clustering

- Track 30 bins with most votes 
- If first bin has match between test and model, proceed with clustering
- Matlab's `subclust()`

```
[C S] = subclust(X, [xy_radii xy_radii 0.4 0.3 0.3], Xbounds);
```

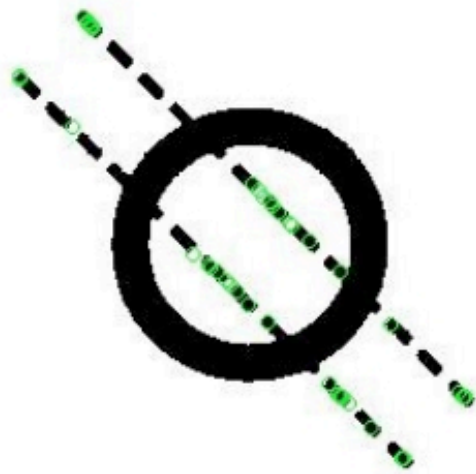
Clustering



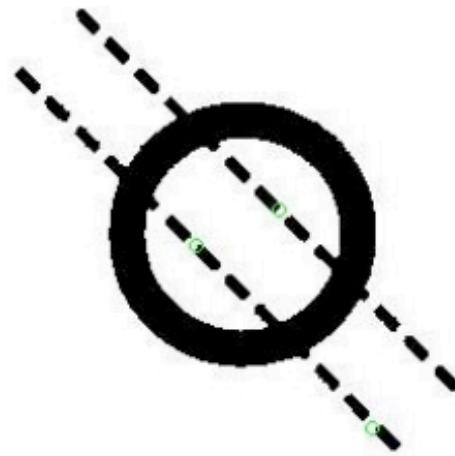
Plot of x and y radii versus number of clusters



Clustering



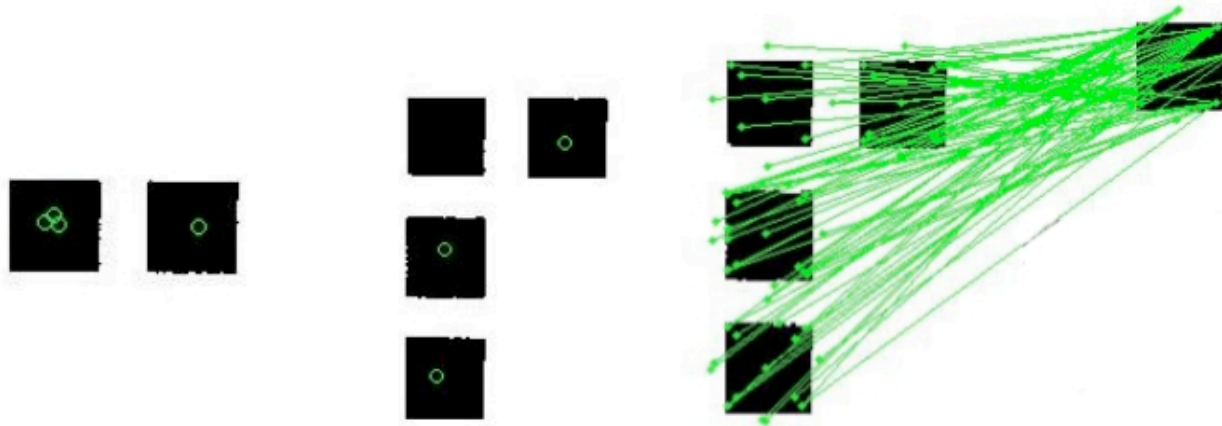
(a) Plot of all model points from a dashed line to the image points



(b) Sample showing 2 clusters found in the image (green dots in the center)



Clustering



(a) Three clusters found in the left square
(b) Top left cluster missing
(c) siftmatch() results between Fig. 8b and the model



Clustering

There are clusters located at (format: [a b c d e]):

156.8772	166.8502	1.0053	-0.9998	0.0211
196.0512	149.6497	1.0226	0.0021	-1.0000
240.3024	252.6831	0.8730	0.9999	-0.0109

where a and b represent the x-y location, c represents the relative scale from the model to the test image, and d and e represent the angle on a unit circle.

$(\cos(\theta), \sin(\theta)) = (d, e)$

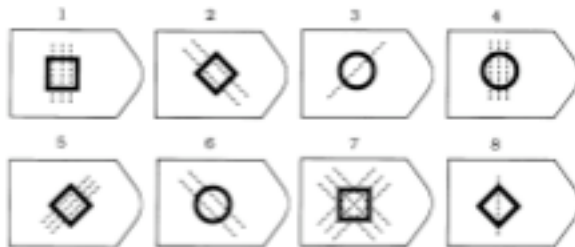
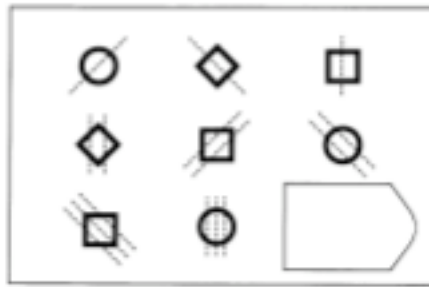


Adaptation to Rasmussen Model

- Basic Idea
 - Apply SIFT/Hough transform on the 8 cells
 - map Rasmussen input with above output
 - If match
 - apply SIFT/Hough transform on answer cells
- theoretical

Adaptation to Rasmussen Model

- Example 1



(a) Raven Matrix 1



(b) Databank of figures



Adaptation to Rasmussen Model

- Cell 1's output from Hough

```
HOUGH IMPLEMENTATION
```

```
=====
```

```
dashed_line_horizontal.jpg located at approximately (173.7823,173.7823)
```

```
scale: 0.99973 times the model
```

```
orientation: 5.4982 radians
```

```
There are clusters located at (format: [a b c d e]):
```

```
173.6544 157.8504 1.0045 0.7079 -0.7063
```

where a and b represent the x-y location, c represents the relative scale from the model to the test image, and d and e represent the angle on a unit circle.

$(\cos(\theta), \sin(\theta)) = (d, e)$



Adaptation to Rasmussen Model

- Cell 1's output from Hough

```
HOUGH IMPLEMENTATION
```

```
=====
```

```
circle.jpg located at approximately (179.9844,179.9844)
```

```
scale: 1.0154 times the model
```

```
orientation: 6.2147 radians
```

```
There are clusters located at (format: [a b c d e]):
```

```
 180.5672 155.7880  1.0035  0.5597  -0.8287
```

```
where a and b ...
```

```
HOUGH IMPLEMENTATION
```

```
=====
```

```
no match for square.jpg
```



Adaptation to Rasmussen Model

- Modified version of Dan's input

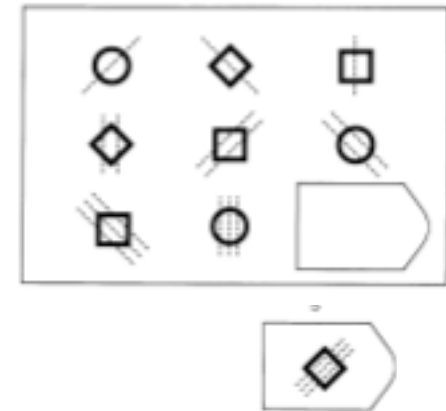
```
#matrix
circle; 1; 45deg
diamond; 1; 135deg
square; 1; 90deg
diamond; 2; 90deg
square; 2; 45deg
circle; 2; 135deg
square; 3; 135deg
circle; 3; 90deg
```

```
#answers
square; 3; 90deg
diamond; 2; 135deg
circle; 1; 45deg
circle; 3; 90deg
diamond; 3; 45deg
circle; 2; 135deg
square; 3; 45deg
diamond; 1; 90deg
```

Adaptation to Rasmussen Model

■ logic

```
if cell 1 contains a circle and 1 dashed line at 45deg &&  
  cell 2 contains a diamond and 1 dashed line at 135deg &&  
  cell 3 contains a square and 1 dashed line at 90deg &&  
  ... &&  
  cell 8 contains a square and 3 dashed line at 90deg  
then  
  apply the Hough transform on the answer cells  
  if Answer 1 contains a diamond and 3 dashed lines at 45deg  
    then output Answer 1  
  else if Answer 2 contains a diamond and 3 dashed lines at 45deg  
    then output Answer 2  
  ...  
  else if Answer 8 contains a diamond and 3 dashed lines at 45deg  
    then output Answer 8  
end
```





Conclusions

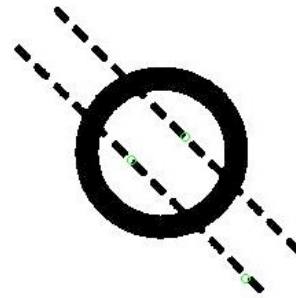
- Is it possible to determine the existence of a model in a given test image?
- **YES!!!**
 - SIFT, Hough transform
 - Clustering
- Theoretically adapt above methods to Rasmussen Model



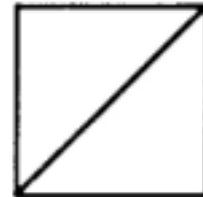
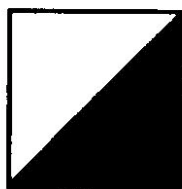
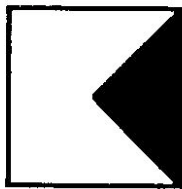
Future Problems

- SIFT has limitations

- Occlusion
- Noise



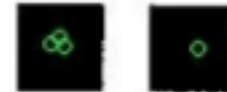
- Non-detectable features (pixel-based)





Future Problems

- Clustering solutions
 - Similarity transform
 - Prior knowledge
 - First bin approach
 - Adjust more parameters
- Training images

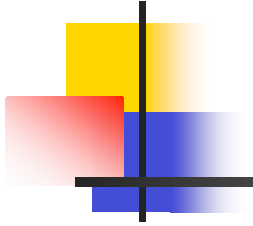




References

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- Vedaldi, A. (2006). *An Implementation of SIFT Detector and Descriptor*. Retrieved March 29, 2010 from <http://www.vlfeat.org/~vedaldi/code/sift.html>

Acknowledgements: Charlie Tang



The End