#### Matching Image Patches



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Many slides from Steve Marschner, Alexei Efros CSC320 Michael Guerzhoy

## **Template matching**

Goal: find **I** in image

#### Main challenge: What is a good similarity or distance measure between two patches?

- Dot product
- (Zero-mean) correlation
- Sum Square Difference
- Normalized Cross Correlation



Slide contents from Derek Hoiem and Alexei Efros

#### Images as vectors

#### Goal: find **Solution** in image Method 0: filter the image with eye patch $h[m,n] = \sum g[k,l] f[m+k,n+l]$



k,l

f = image g = filter

#### What went wrong?

Filtered Image

#### Goal: find in image Method 1: filter the image with zero-mean eye $h[m,n] = \sum_{k,l} (f[k,l] - \overline{f}) (g[m+k,n+l])$ mean of f (here, f is the filter)



Input



Filtered Image (scaled)



# Goal: find in image Method 2: SSD $h[m,n] = \sum_{k,l} (g[k,l] - f[m+k,n+l])^2$





True detections

Input

1- sqrt(SSD)

## Goal: find in image What's the potential downside of SSD? Method 2: SSD $h[m,n] = \sum_{l=l}^{l} (g[k,l] - f[m+k,n+l])^2$



Input

1- sqrt(SSD)

# Goal: find Solution image Method 3: Normalized cross-correlation

$$h[m,n] = \frac{\sum_{k,l} (g[k,l] - \overline{g})(f[m+k,n+l] - \overline{f}_{m,n})}{\left(\sum_{k,l} (g[k,l] - \overline{g})^2 \sum_{k,l} (f[m+k,n+l] - \overline{f}_{m,n})^2\right)^{0.5}}$$

# Goal: find Solution image Method 3: Normalized cross-correlation



Input

Normalized X-Correlation

# Goal: find Solution image Method 3: Normalized cross-correlation



Input

Normalized X-Correlation

#### Q: What is the best method to use?

A: Depends

Zero-mean filter: fastest but not a great matcher

SSD: next fastest, sensitive to overall intensity

Normalized cross-correlation: slowest, invariant to local average intensity and contrast

#### Image half-sizing

This image is too big to fit on the screen. How can we reduce it?

How to generate a halfsized version?

