

Getting Beyond Distance: (Stats) Cross-Correlation & (EE) Correlation

CS 510
Lecture #9
February 18, 2015

Colorado State University

Pearson's Correlation

Recall from the previous lecture ...

$$\frac{\sum_{x,y} (A(x,y) - \bar{A})(B(x,y) - \bar{B})}{\sqrt{\sum_{x,y} (A(x,y) - \bar{A})^2} \sqrt{\sum_{x,y} (B(x,y) - \bar{B})^2}}$$

This is a very important equation...

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Assumptions of Correlation

$$\frac{\sum_{x,y} (A(x,y) - \bar{A})(B(x,y) - \bar{B})}{\sqrt{\sum_{x,y} (A(x,y) - \bar{A})^2} \sqrt{\sum_{x,y} (B(x,y) - \bar{B})^2}}$$

- Two signals vary linearly
 - Constant shift to either signal has no effect.
 - Increased amplitude has no effect.
- This minimizes sensitivity to:
 - changes in (overall) illumination
 - offset or gain.

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Computing Correlation

- A constant added to A does not change its correlation to any other signal, so
 - Let's subtract average A from A()
 - Let's subtract average B from B()
 - The mean of both signals is now zero
 - Then correlation reduces to:

$$A \cdot B$$

Know and love the dot product.

$$\frac{\sum_{x,y} (A(x,y) - \bar{A})(B(x,y) - \bar{B})}{\sqrt{\sum_{x,y} (A(x,y) - \bar{A})^2} \sqrt{\sum_{x,y} (B(x,y) - \bar{B})^2}}$$

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Computing Correlation (II)

- For zero-mean signals, we can scale them without changing their correlation scores
 - Multiply A by the inverse of its length
 - Multiply B by the inverse of its length
 - Both signals are now unit length
 - Then correlation reduces to:

$$A \cdot B$$

Know and love
the dot product.



- Gives rise to 'Correlation Space'.

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Correlation Space

- Why zero-mean & unit-length your images?
- Consider, for example, database retrieval
 - Compare new image A ...
 - with many images in database.
 - When database images are stored in their zero-mean & unit-length form, then
 - Preprocess A (zero-mean, unit-length)
 - Compute dot products

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Correlation Space (II)

- New idea: image as a point in an N dimensional space
 - $N = \text{width} \times \text{height}$
- Zero-mean & unit-length images lie on an $N-1$ dimensional “correlation space” where the dot product equals correlation.
 - This is a highly non-linear projection.
 - Points lie on an $N-1$ surface within the original N dimensional space.
- So consider points in 3-D

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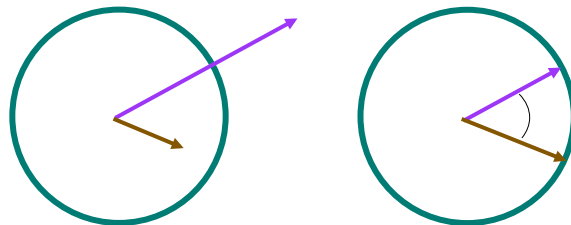
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Correlation Space (II)

- Subtracting mean - translation.
- Length one - project onto sphere.
- Correlation is then:
 - Cosine of angle between vectors (points).



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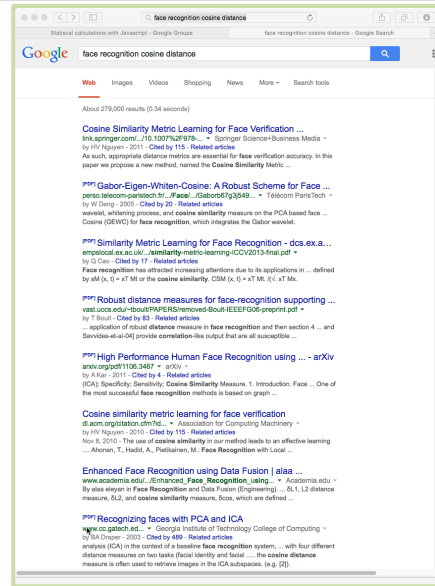
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Useful?

- Yes
- Very commonly used.
- For example
 - Face Rec.
- Google
 - 279,000 hits



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Useful Connection ...

- Euclidean distance is inversely proportional to correlation in correlation space.

$$\begin{aligned}
 \sqrt{\sum_{x,y} (A[x,y] - B[x,y])^2} &= \sqrt{\sum_{x,y} A[x,y]^2 + \sum_{x,y} B[x,y]^2 - 2A[x,y]B[x,y]} \\
 &= \sqrt{1 + 1 - 2 \sum_{x,y} A[x,y]B[x,y]} \\
 &= \sqrt{2 - 2A \cdot B} \\
 &= \sqrt{2 - 2\text{Corr}(A,B)}
 \end{aligned}$$

- Nearest-neighbor classifiers in correlation space maximize correlation / minimized L_2 norm

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Limitations

- To match images this way, they must be
 - The same width & height
 - In correspondence : coordinates match
- More importantly, objects in the scene must
 - Be in the same location
 - Be at the same scale
 - Be at the same orientation
 - Be seen from the same viewpoint

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New Goal: Find a small image within a larger one



- The image above is a small piece of the image to the right. But from where?



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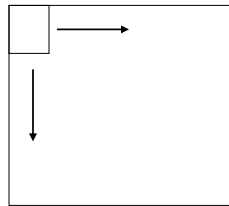
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Brute-Force Translation Invariance

To find a small image in a large one, “slide” the small one across the large, computing Pearson’s correlation at every possible position.



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Statistical Cross-Correlation

- The process of “slide & correlate” is called cross-correlation
- Complexity is $O(nm)$
 - $N = \#$ of pixels in image ($w \times h$)
 - $M = \#$ of pixels in the template ($w \times h$)
- Highly parallel (every position can be computed independently)
- Still sensitive to
 - Rotation
 - in-plane
 - out-of-plane
 - Scale
 - Perspective

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Computing Cross-Correlation

- In cross-correlation, the mask is correlated repeatedly to image windows
 - zero-mean & unit length the mask
 - zero-mean & unit length the image
 - compute the sliding dot product

This is *almost* convolving the image with the mask

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Naming conventions

- In Engineering, convolving a normalized mask with the source image is called correlation
 - Is this exactly the same as Pearson's correlation?
 - Why or why not?
- This is the most common definition of correlation in image processing texts

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Application: Tracking

- Cut picture of a target from the first frame of a video
 - Use it as a template /mask
- Correlate the target in the following frames
 - Move to highest correlation peak and cut new target.

Project 2: Template Tracking Video Stabilization - Due Friday, March 26th. Due Friday, April 2nd.



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Add Scale Correction



ECE 695 Project: OBJECT TRACKING IN VIDEOS, Arun Anbumani, December 18, 2013

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Application: Mosaicing

- Take several, overlapping images from a translating camera
 - Camera cannot move along optical axis
- Correlate the whole images to each other
 - Find location where they match the best
 - Stitch them into a single, larger image

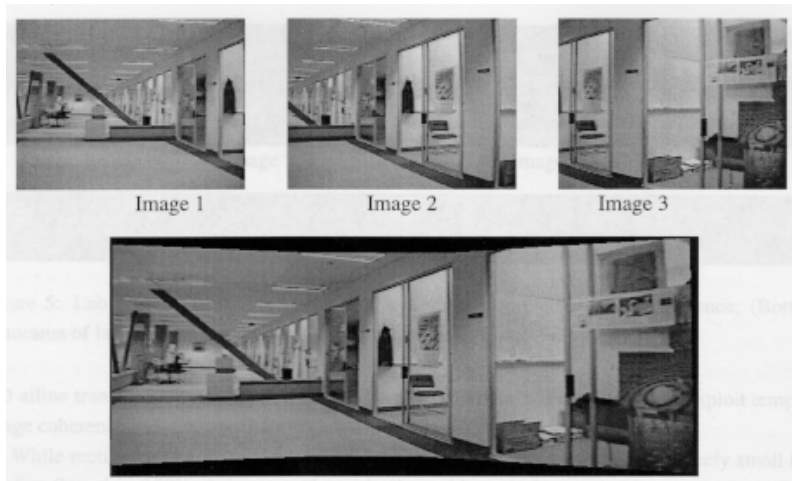
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Mosaicing (II)



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Application: Detection

- Could you use correlation to find...
 - Human faces?
 - heads?
 - Hands?
 - bodies?
- What (if anything) might lead to failures?

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