

Computer Vision

CS311, Spring 2013
David Kauchak

some slides modified from slides obtained from Zach Dodds

+ Admin

- Exam #2 available
 - Take by Sunday at 11:59pm
- No office hours on Friday
- Status reports
 - Keep working on the project ☺
 - Status report 2 due Monday

Every picture tells a story...

**What is going on in this picture?
How did you figure it out?**

+ Computer Vision

What is computer vision?



+ Computer Vision

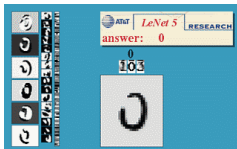
the goal of computer vision is to write computer programs that can interpret images (and videos)



What are some of the challenges?

Applications?

+ Optical character recognition



Long-term work on digits, AT&T labs

deskew



binarize



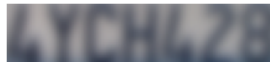
segment



recognize

License plate readers

and ways to get around them!



+ Sports



[machines vs. humans]
"The balls moving so fast these days that sometimes its impossible for anyone to see, even a trained official."
- James Blake



Sportvision

What are some of the problems that these two systems need to handle?


- The system has to know the **orientation of the field with respect to the camera** so that it can paint the first-down line with the correct perspective from that camera's point of view.
- The system has to know, in that same perspective framework, exactly **where every yard line is**.
- Given that the cameraperson can move the camera, the system has to be able to **sense the camera's movement** (tilt, pan, zoom, focus) and **understand the perspective change** that results from the movement.
- Given that the camera can pan while viewing the field, the system has to be able to **recalculate the perspective** at a rate of 30 frames per second as the camera moves.
- A football field is not flat – it crests very gently in the middle to help rainwater run off. So the line calculated by the system has to appropriately **follow the curve of the field**.
- A football game is filmed by **multiple cameras** at different places in the stadium, so the system has to do all of this work for several cameras.
- The system has to be able to **sense when players, referees or the ball cross over the first-down line** so it does not paint the line right on top of them.
- The system also has to be **aware of superimposed graphics** that the network might overlay on the scene.

high-resolution encoders on the cameras

detailed field model ~ crest!

color palettes: in and out

slight delay in the network feed



Sportvision

+ Challenges


- orientation of the field with respect to camera
- handle camera movement
 - recalculate perspective
- where the yard lines are
- multiple cameras
- don't paint over players refs!
- other superimposed graphics

high-resolution encoders on the cameras

detailed field model

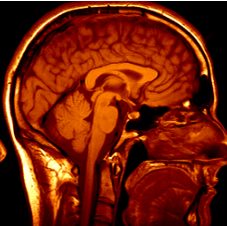
color palettes: in and out

slight delay in the network feed



Sportvision

+ Medical imaging



3D imaging: MRI, CT





Image guided surgery: Eric Grimson @ MIT


+ Face detection

Recognition is more difficult... but products are pushing that way!

http://www.youtube.com/watch?v=N1WC_00L060



on many cameras...



Face Recognition

NATIONAL GEOGRAPHIC
VOL. 92, NO. 6 JUNE 2012

GREY BULL LARVAE: THE BLOODING
U.S. MEDICAL BRIGADE: LIFE ON THE LINE
JAMES WALKER: LIFE ON THE LINE
Alamy, Mechanicals, What's a Fraction?
FIVE SELESTES FOR THE LANTERN GLANCE

THE NATIONAL GEOGRAPHIC CHANNEL BEST VIEWED BY TELEVISION CABLE TV

Smile detection?

The Smile Shutter flow
Imagine a camera smart enough to catch every smile! In Smile Shutter Mode, your Cyber-shot® camera can automatically trip the shutter at just the right instant to catch the perfect expression.

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Technology resources provided by ELSA FORBZ. © 2012 ELSA FORBZ

Sony's 'smile shutter' might make you grimace

Security

Fingerprint scanners can be vision-based devices

Face recognition systems now beginning to appear more widely
www.sensiblevision.com

key drawbacks?!

Entertainment

Shape capture
ESC Entertainment, XYZRGB, NRC

Motion capture
<http://www.ilm.com/thehow/>

Safety

Our Vision. Your Safety.



rear looking camera forward looking camera
side looking camera

EyeQ Vision on a Chip



> read more

Vision Applications

Road, Vehicle, Pedestrian Protection and more



> read more

AWS Advance Warning System



> read more

News

► **Advanced Technologies Power Volvo Car Model From Adaptive Braking With Auto Brake System**

► **Volvo New Collision Warning with Auto Brake: Helps Prevent Rear-end**

Events

► **Holders at Euro Auto Expo, France**

► **Holders at SEMA, Las Vegas, NV**


> read more

MobilEye vision systems currently in high-end BMW, GM, Volvo model: ~70% of car manufacturers use cameras for safety

courtesy of Amnon Shashua


LaneHawk

simpler recognition: products passing by...




"A smart camera is flush-mounted in the checkout lane, continuously watching for items. When an item is detected and recognized, the cashier verifies the quantity of items that were found under the basket, and continues to close the transaction. The item can remain under the basket, and with LaneHawk, you are assured to get paid for it..." ~ Evolution Robotics, Pasadena.


Games and vision-based interaction




Wimotes: infrared images



XBOX 360 Kinect



Digimask: put your face on a 3D avatar.



Camera tracking for crowd interactions...

Exploration in hostile environments (OK, robots!)



NASA's MER "Spirit" captured this westward view from atop a low plateau where Spirit spent the closing months of 2007.

Vision systems (JPL) used for several tasks

- Panorama stitching
- 3D terrain modeling
- Obstacle detection, position tracking
- slip detection on uncertain terrain

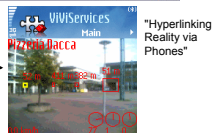
Larry Mathies, CMU

+ Object recognition with mobile phones

Lincoln Microsoft research



also: Point & Find, Nokia



+ Object recognition with mobile phones

Lincoln Microsoft research

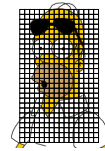


pretty much sums up the state-of-the-art!

+ How is an image represented?



+ How is an image represented?

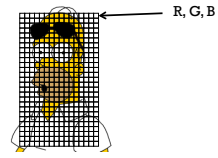


- images are made up of pixels
- for a color image, each pixel corresponds to an RGB value (i.e. three numbers)

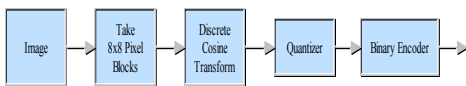
+ Image file formats

- BitMaP
- JPEG
- TIFF
- Gif
- Png
- ...

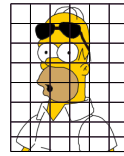
+ Bitmap

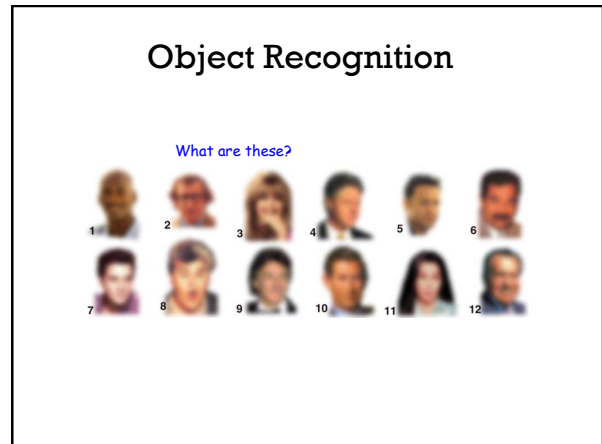
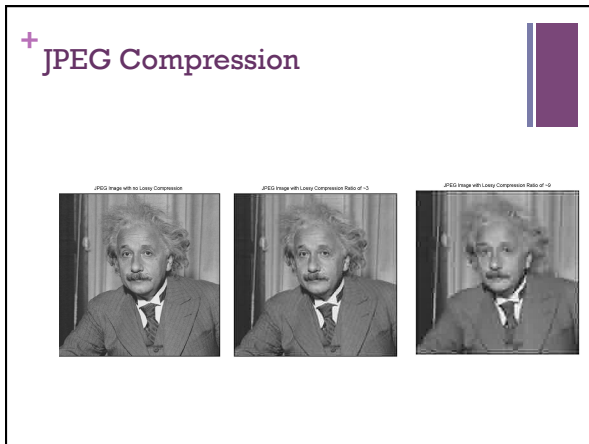
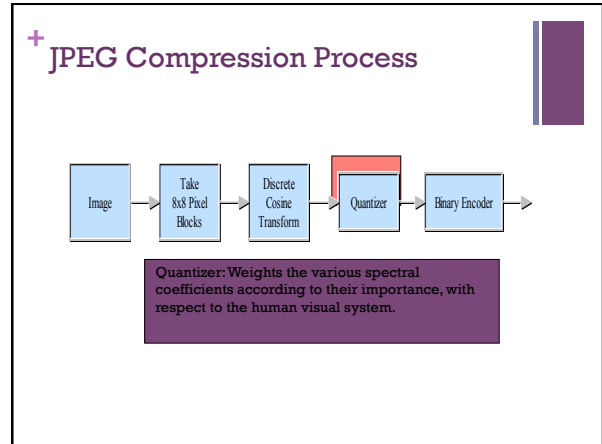
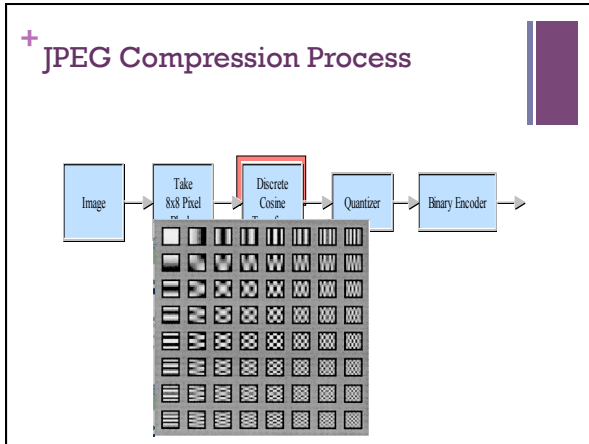


+ JPEG Compression Process




+ JPEG Compression Process





Object Recognition

Do you recognize these people?




order are: Michael Jordan, Woody Allen, Goldie Hawn, Bill Clinton, Tom Hanks, Saddam Hussein, Elvis Presley, Jay Leno, Dustin Hoffman, Prince Charles, Cher, and Richard Nixon.

Different kinds of object recognition?



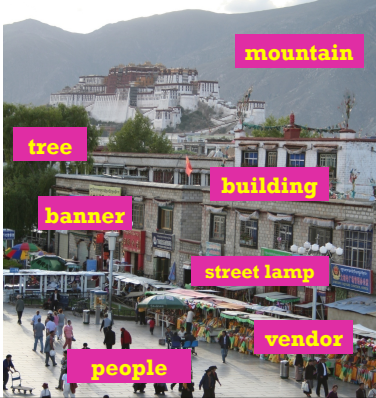
Identification: is that Potala Palace?



Detection: are there people (or faces)?



Object and scene *categorization*



Verification: is that a lamp?



+ Recognition Question(s)

- Identification:** Where is *this particular* object?
- Detection:** Locate *all* instances of a given class
- Content-based image retrieval:** Find something similar
- Categorization:** What *kind* of object(s) is(are) present?
- Verification:** Is this what I think it is?

How might you arrange these, in order of difficulty?

[Ceuzka et al. 2006]:

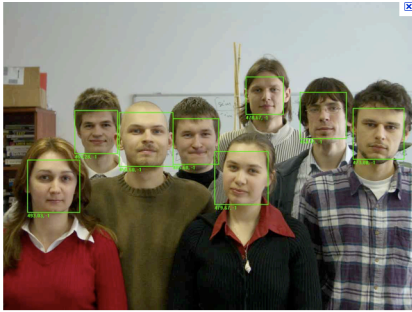
+ Recognition Questions

- Verification:** Is this what I think it is?
- Identification:** Where is *this particular* object?
- Content-based image retrieval:** Find something similar
- Detection:** Locate *all* instances of a given class
- Categorization:** What *kind* of object(s) is(are) present?



Certainly arguable !

+ Today: face recognition



+ Face recognition?

Verification: Is this what I think it is?

Identification: Where is *this particular* object?

Content-based image retrieval: Find something similar

Detection: Locate *all* instances of a given class

Categorization: What *kind* of object(s) is(are) present?

+ Face recognition?

Verification: Is this what I think it is?

Identification: Where is *this particular* object?

Content-based image retrieval: Find something similar

Detection: Locate *all* instances of a given class

Categorization: What *kind* of object(s) is(are) present?

+ Eigenfaces: how do people do it?



The "Margaret Thatcher Illusion", by Peter Thompson

Eigenfaces for recognition

Matthew Turk and Alex Pentland
J. Cognitive Neuroscience, 1991

+ Eigenfaces: how do people do it?



The "Margaret Thatcher Illusion", by Peter Thompson

Eigenfaces for recognition

Matthew Turk and Alex Pentland
J. Cognitive Neuroscience, 1991

+ Image features

We'd like to represent an image as a vector of features

- good for machine learning techniques
- distance/similarity measures
- etc.

What are possible features?

+ Color

How can we represent color?

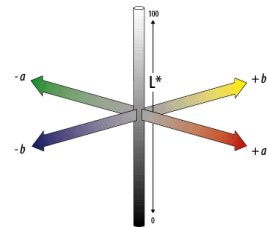
Which is more similar?



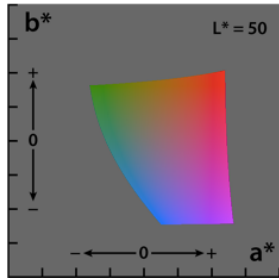
+ L*a*b*

L*a*b* was designed to be uniform in that perceptual "closeness" corresponds to Euclidean distance in the space.

- L - lightness (white to black)
- a - red-greenness
- b - yellowness-blueness



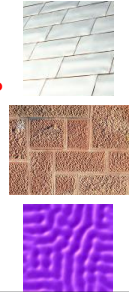
+ $L^*a^*b^*$



Is color useful for face detection/verification?

+ Texture

How is texture different than color?



+ Texture

Texture is not pointwise like color
 Texture involves a local neighborhood

How can we capture texture?



+ Local "response" to feature functions



A "feature" is a particular low-resolution image (intensities)

"convolution"

- matrix dot product
- image portions with similar intensities will have high values

Lots of possible features!

+ Example: Gabor Filters

Gabor filters are Gaussians modulated by sinusoids
They can be tuned in both the scale (size) and the orientation



Scale: 3 at 72°

Scale: 4 at 108°

Scale: 5 at 144°

+ Gabor filters



What would the response look like to a vertical filter?

+ Gabor filters

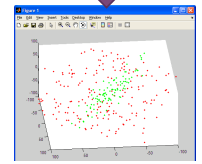


Eigenfaces

Given a face, we can then calculate its response to a number of these filters generating a feature vector



First-thoughts for detection?
First-thoughts for identification?



~10,000 dimensional space

Eigenfaces

Idea: faces have distinctive appearance

There is some *intra-class* variation

But nowhere *near* the *inter-class* variation with "everything else"

non-faces in purple
faces in orange

Only a few dimensions needed

x-y projection x-z projection

but which ones?

Only a few dimensions needed

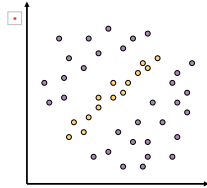
this is a promising view!

+ Learning a projection

We saw data projection when we were looking at machine learning techniques... where?

How did we figure out the projection for clustering?

Dimensionality reduction



How can we find the data's natural coordinate system?

Principal component analysis

Suppose each data point is N-dimensional

What directions maximize variance?

$$\begin{aligned} \text{var}(v) &= \sum_{\mathbf{x}} \|(\mathbf{x} - \bar{\mathbf{x}})^T \cdot v\|^2 \\ &= v^T \mathbf{A} v \quad \text{where } \mathbf{A} = \sum_{\mathbf{x}} (\mathbf{x} - \bar{\mathbf{x}})(\mathbf{x} - \bar{\mathbf{x}})^T \end{aligned}$$

Solution: the eigenvectors of the variance matrix \mathbf{A}

- eigenvector with largest eigenvalue captures the most variation among training vectors \mathbf{x}
- eigenvector with smallest eigenvalue has least variation

We can use only the top few eigenvectors

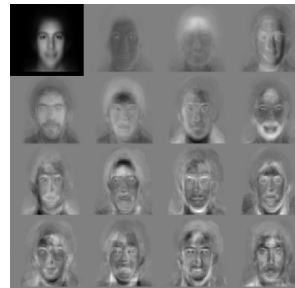
- corresponds to choosing a "linear subspace"
- represent points on a line, plane, or "hyper-plane"
- these eigenvectors are known as the *principal components*

Eigenfaces: pictures!



Does this look like anyone you know?

Eigenfaces: pictures!



What do each of these mean?

Eigenfaces (plus the average face)

Projecting onto low-d eigenspace

The eigenfaces $\mathbf{v}_1, \dots, \mathbf{v}_K$ span the space of faces

- A face is converted to eigenface coordinates by

$$\mathbf{x} \rightarrow ((\mathbf{x} - \bar{\mathbf{x}}) \cdot \mathbf{v}_1, (\mathbf{x} - \bar{\mathbf{x}}) \cdot \mathbf{v}_2, \dots, (\mathbf{x} - \bar{\mathbf{x}}) \cdot \mathbf{v}_K)$$

$$\mathbf{x} \approx \bar{\mathbf{x}} + a_1 \mathbf{v}_1 + a_2 \mathbf{v}_2 + \dots + a_K \mathbf{v}_K$$

The diagram shows a single grayscale face on the left. An arrow points to a row of seven grayscale eigenfaces. A second arrow points from the eigenfaces back to the original face, illustrating the reconstruction process.

Eigenfaces: pictures!

The image shows two sets of face images. The left set is a 5x5 grid of 25 grayscale eigenfaces, which are variations of a face without the average face component. The right set is a 5x5 grid of 25 grayscale faces, showing the result of progressively adding more eigenfaces to the average face to reconstruct the original face.

Eigenfaces (without the average face)

Progressive reconstructions...

How many dimensions?

The graph plots eigenvalues λ_i on the y-axis against the number of eigenfaces i on the x-axis. The curve starts high and decays rapidly, leveling off as i increases towards NM . A vertical dashed line is drawn at $i = K$.

The hope...

How many eigenfaces to use?

Look at the decay of the eigenvalues

- the eigenvalue tells you the amount of variance "in the direction" of that eigenface
- ignore eigenfaces with low variance

How many dimensions?


The graph plots total variance captured on the y-axis (from 0 to 1) against the number of eigenfaces used on the x-axis (from 0 to 100). The curve shows that as the number of eigenfaces increases, the total variance captured approaches 1. A vertical red line is drawn at $K = 10$ and a vertical green line at $K = 25$.

In practice

Total variance captured vs. number of eigenfaces used

- $K = 10$ captures about 36% of the variance
- $K = 25$ captures about 56%


Eigenfaces: recognition (id)




32 test cases
Novel image on left; best-matching image on right

Eigenfaces: recognition (id)

Different lighting conditions




Different facial expressions




On which set do you think eigenfaces will perform better?

Eigenfaces: recognition (id)

Different lighting conditions

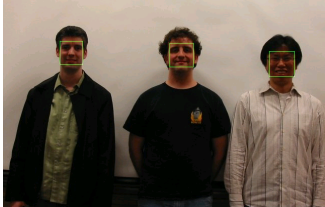


Different facial expressions



9/16 for lighting changes ... 23/26 for expression changes

Eigenfaces: detection



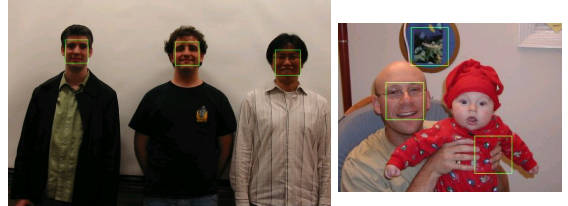
How can we do this using eigenfaces?

Eigenfaces: detection



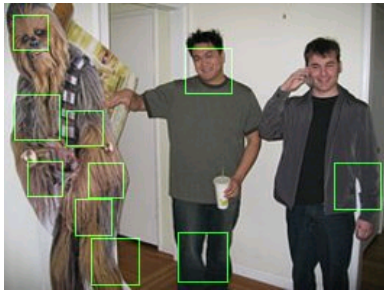
Difficult to avoid false positives...
Top 4

Eigenfaces: detection



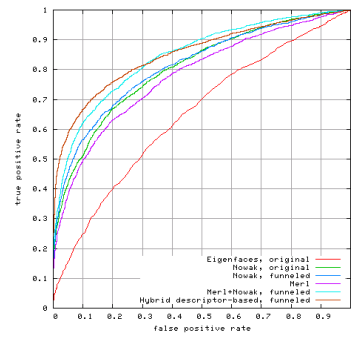
Difficult to avoid false positives...
Top 3

Eigenfaces: detection




Top few

Receiver-operating curve



shows true positive vs. false positive rate


What's wrong with detection?



each top image, projected onto eigenspace

How do we improve this?

What's wrong with detection?



each top image, projected onto eigenspace

Reasonable once we have an image of a face (recognition)

Not so good at finding faces (detection)

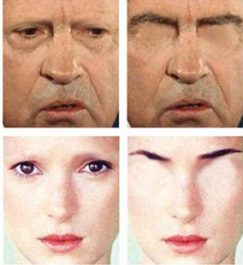
+ What parts are important?



what's missing?

who are these two people?

+ What parts are important?



eyes vs. eyebrows

who are these two people?

+ **What parts parts are important?**



Nixon

Winona Ryder

who are these two people?

+ **Robust** real-time face detection

Paul A. Viola and Michael J. Jones
Intl. J. Computer Vision. 57(2), 137–154, 2004

Learn which "parts" are most important...

+ **Image features**

"Rectangle/box filters"
 ■ Similar to Haar wavelets

Differences between sums of pixels in adjacent rectangles

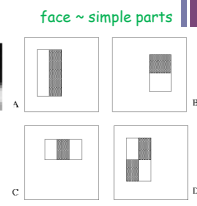
Simple thresholding

$$h_i(x) = \begin{cases} +1 & \text{if } f_i(x) > \theta_i \\ -1 & \text{otherwise} \end{cases}$$

each box filter is present or absent

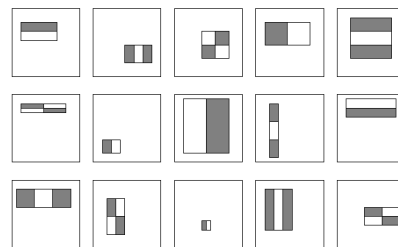


24x24



gray regions are subtracted (after summing)
 white regions are added (after summing)

+ **Huge library of filters**



+ Constructing the classifier

For each round of boosting: (AdaBoost)

- Evaluate each rectangle filter on each example
- Sort examples by filter values
- Select best threshold for each filter (min error)
 - Use sorting to quickly scan for optimal threshold
- Select best filter/threshold combination
- Reweight examples
 - (There are many tricks to make this more efficient.)

+ Characteristics of algorithm

Feature set (...is huge about 16M features)

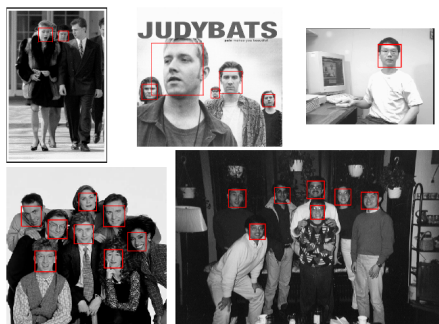
Efficient feature selection using AdaBoost

New image representation

Cascaded Classifier combining simple weak classifiers for rapid detection

➤ Fastest known face detector for gray scale images

Viola and Jones: Results



+ First two filters

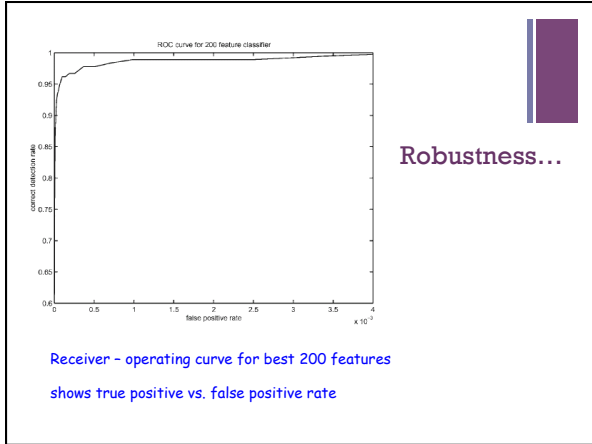
First classifier:

- 2 features
- 100% detection
- 40% false detection



The whole cascade:

- 38 stages
- 6000 features in total
- On dataset with 507 faces and 75 millions sub-windows, faces are detected using 10 feature evaluations on average.
- On average, 10 feature evals/sub-window



+ Summary (Viola-Jones)

Fastest known face detector for gray images

Three contributions with broad applicability:

- ◆ Cascaded classifier yields rapid classification
- ◆ AdaBoost as an extremely efficient feature selector
- ◆ Rectangle Features + Integral Image can be used for rapid image analysis

But, there are better ones out there...

