

# Template-Based Face Detection

CSE 6367 – Computer Vision  
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# Face Detection

- We will make a few assumptions, to simplify the problem:
  - The face is fully visible.



Example cases where the face is not fully visible

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  - The face is fully visible.
  - We see a *frontal* view of the face, i.e., the face is facing towards the camera.



Example cases where the view of the face is not frontal

# Face Detection

- We will make a few assumptions, to simplify the problem:
  - The face is fully visible.
  - We see a *frontal* view of the face, i.e., the face is facing towards the camera.
  - The face is more or less in an upright orientation.

Example of a face not in an upright orientation



# What is a Template?

- A template is an example of how an object looks.
- What example would be appropriate if we are looking for a face?

# What is a Template?

- A template is an example of how an object looks.
- What example would be appropriate if we are looking for a face?
  - An *average* face.

# Computing an Average Face

- We need *aligned* face images:
- In this case, *aligned* means:
  - same size
  - significant features (eyes, nose, mouth), to the degree possible, are in similar pixel locations.



an example set of aligned face images

# Computing an Average Face

```
% Note that filenames is a cell variable (look it up!).
% Using cells is a good way to define a set of strings.
filenames = {
    '4846d101.bmp'
    '4848d101.bmp'
    '4851d101.bmp'
    '4853d101.bmp'
    '4854d101.bmp'
};

number = prod(size(filenames));
image_vertical = 100;
image_horizontal = 100;
total = zeros(image_vertical, image_horizontal);

for index = 1: number
    image = read_gray(filenames{index});
    total = total + image;
    disp(index);
end

average_face = total / number;
```

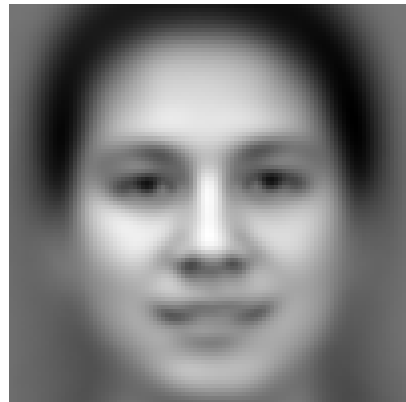


# Result: Average Face



# Defining a Face Template

- Keep the parts of the average face that are most likely to be present in all faces:
  - Exclude background.
  - Exclude forehead (highly variable appearance, due to hair).
  - Exclude lower chin.



average face



face template

# Using a Template to Find Faces

- How can we use a face template to perform face detection in an image?

# Finding Matches for the Template

- How can we find good matches for a given template?

# Using Normalized Correlation

```
function result = vector_normalized_correlation(v1, v2)

% function result = vector_normalized_correlation(v1, v2)

centered_v1 = v1 - mean(v1);
centered_v2 = v2 - mean(v2);
norm1 = norm(centered_v1);
norm2 = norm(centered_v2);

result = centered_v1 .* centered_v2 / (norm1 * norm2);
```

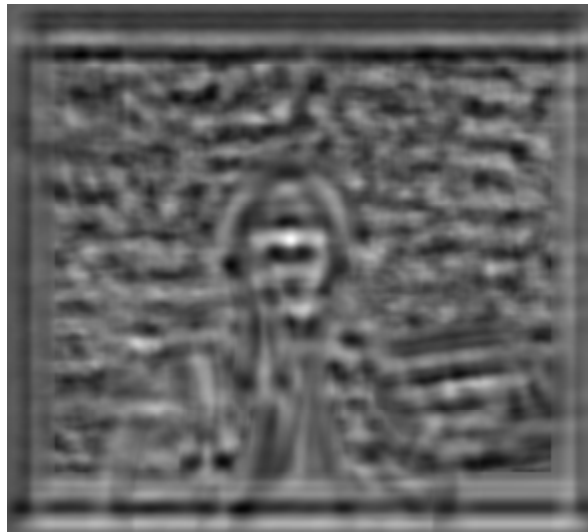
- Function `normxcorr2(template, image)` returns a matrix of normalized correlation scores between the template and each template-sized subwindow of the image.

# Invoking normxcorr2

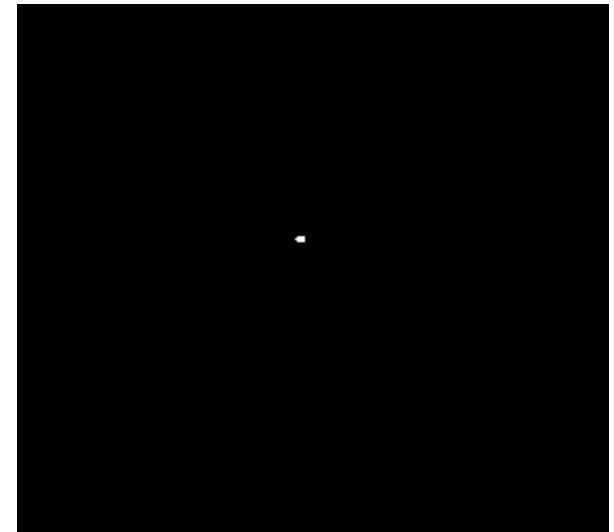
```
photo = read_gray('vassilis1g.bmp');  
result = normxcorr2(face_filter, photo);
```



photo



result



result > 0.6

- It found the face!
- The result of normxcorr2 has larger size than the input.
- The face in photo matched the scale of the template.

# normalized\_correlation

- A wrapper around normxcorr2.
- The result has the same size as the image.
- Border values are zero.

```
function result = normalized_correlation(image, template)

% function result = normalized_correlation(image, template)
%
% Returns a matrix containing normalized correlation results between
% template and all template-sized subwindows of image.

[image_rows, image_columns] = size( image);
[template_rows, template_columns] = size(template);
row_start = floor(template_rows / 2) + 1;
row_end = row_start + image_rows - 1;
col_start = floor(template_columns / 2) + 1;
col_end = col_start + image_columns - 1;

result = normxcorr2(template, image);
[result_rows, result_columns] = size(result);
result(1:template_rows, :) = 0;
result((result_rows-template_rows+1):result_rows, :) = 0;
result(:, 1:template_columns) = 0;
result(:, (result_columns-template_columns+1):result_rows, :) = 0;

result = result(row_start:row_end, col_start:col_end);
```

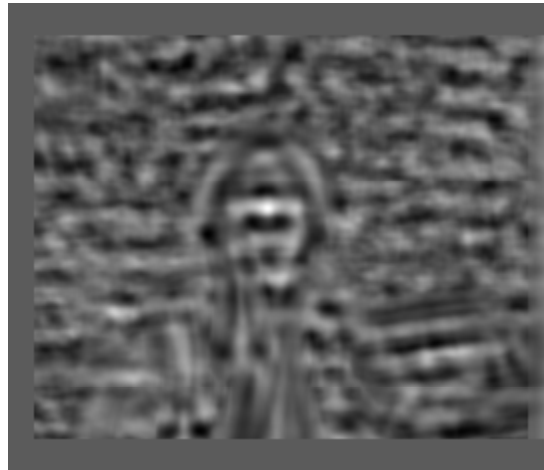


# Invoking normalized\_correlation

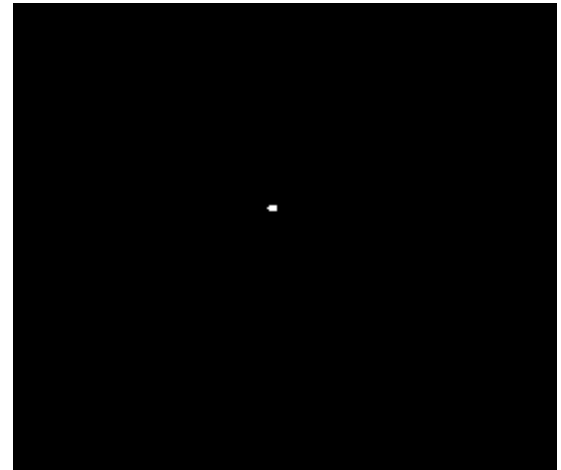
```
photo = read_gray('vassilis1g.bmp');  
result = normalized_correlation(photo, face_filter);
```



photo



result



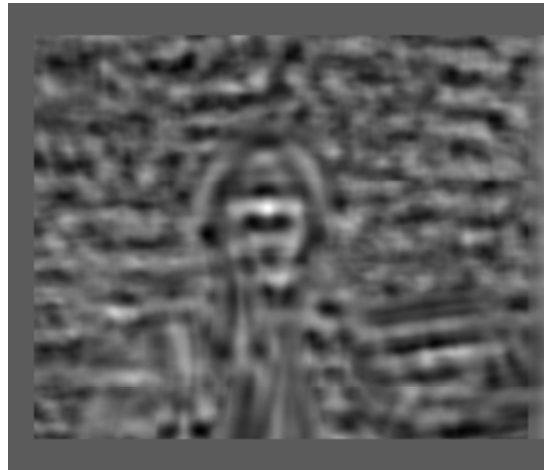
result > 0.6

# A Trick for Visualizing Results

```
photo = read_gray('vassilis1g.bmp');  
result = normalized_correlation(photo, face_filter);  
visualization = max((result > 0.6)*255, photo * 0.7);
```



photo



result



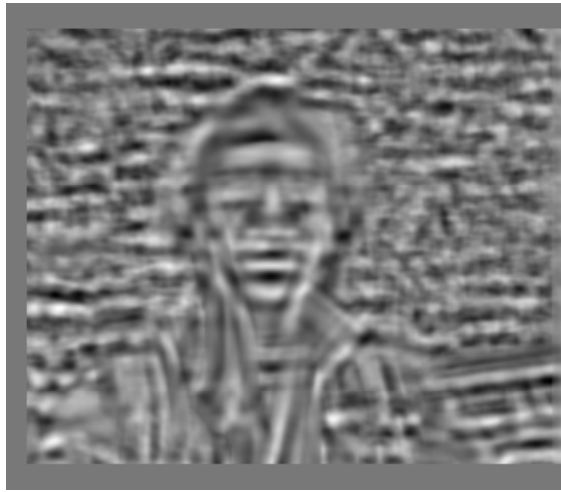
visualization

# Problem: Different Scales

```
photo = read_gray('vassilis1.bmp');  
result = normalized_correlation(photo, face_filter);  
visualization = max((result > 0.35)*255, photo * 0.7);
```



photo



result



visualization

# Solution: Multiscale Search

```
function ...  
[result, max_scales] = multiscale_correlation(image, template, scales)  
  
% function result = multiscale_correlation(image, template, scales)  
%  
% for each pixel, search over the specified scales, and record:  
% - in result, the max normalized correlation score for that pixel  
%   over all scales  
% - in max_scales, the scale that gave the max score
```

# Solution: Multiscale Search

```
function ...
[result, max_scales] = multiscale_correlation(image, template, scales)

result = ones(size(image)) * -10;
max_scales = ones(size(image)) * -10;

for scale = scales;
    % for efficiency, we either downsize the image, or the template,
    % depending on the current scale
    if scale >= 1
        scaled_image = imresize(image, 1/scale, 'bilinear');
        temp_result = normalized_correlation(scaled_image, template);
        temp_result = imresize(temp_result, size(image), 'bilinear');
    else
        scaled_image = image;
        scaled_template = imresize(template, scale, 'bilinear');
        temp_result = normalized_correlation(image, scaled_template);
    end

    higher_maxes = (temp_result > result);
    max_scales(higher_maxes) = scale;
    result(higher_maxes) = temp_result(higher_maxes);
end
```

# Results of multiscale\_correlation

```
photo = read_gray('vassilis1e.bmp');  
scales = 0.5:0.1:3;  
[result2, max_scales] = ...  
    multiscale_correlation(photo, face_filter, scales);  
visualization2 = max((result2 > 0.6)*255, photo * 0.7);
```



photo



result2



visualization2

# Handling Rotations

```
load face_filter;  
photo = read_gray('vassilis2b.bmp');  
[result, boxes] = ...  
    template_detector_demo(photo, face_filter, 0.5:0.1:3., 0, 1);
```



photo



result

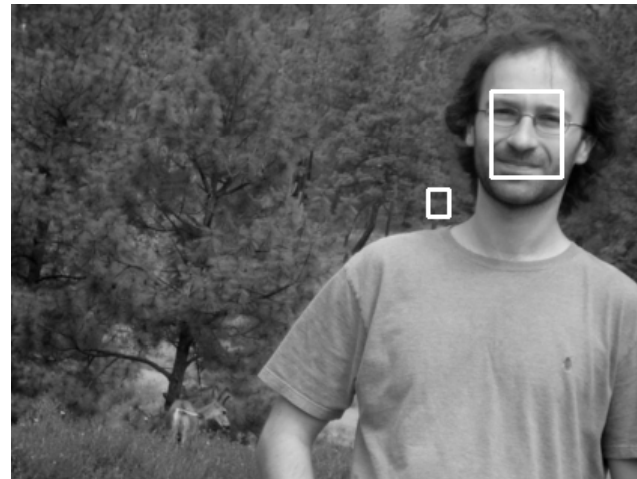


# Handling Rotations

```
load face_filter;  
photo = read_gray('vassilis2b.bmp');  
[resultb, boxes] = ...  
    template_detector_demo(photo, face_filter, 0.5:0.1:3., 0, 2);
```



photo



resultb



# Code for Template Search

- Useful functions:
  - `template_search`
  - `find_template`
  - `template_detector_demo`

```
function [max_responses, max_scales, max_rotations] = ...
    template_search(image, template, scales, rotations, result_number)

% function [result, max_scales, max_rotations] = ...
%     template_search(image, template, scales, rotations, result_number)
%
% for each pixel, search over the specified scales and rotations,
% and record:
% - in result, the max normalized correlation score for that pixel
%   over all scales
% - in max_scales, the scale that gave the max score
% - in max_rotations, the rotation that gave the max score
%
% clockwise rotations are positive, counterclockwise rotations are
% negative.
% rotations are specified in degrees
```

```

function [max_responses, max_scales, max_rotations] = ...
    template_search(image, template, scales, rotations, result_number)

max_responses = ones(size(image)) * -10;
max_scales = zeros(size(image));
max_rotations = zeros(size(image));

for rotation = rotations
    rotated = imrotate(image, -rotation, 'bilinear', 'crop');
    [responses, temp_max_scales] = ...
        multiscale_correlation(rotated, template, scales);
    responses = imrotate(responses, rotation, 'nearest', 'crop');
    temp_max_scales = imrotate(temp_max_scales, rotation, ...
        'nearest', 'crop');
    higher_maxes = (responses > max_responses);
    max_responses(higher_maxes) = responses(higher_maxes);
    max_scales(higher_maxes) = temp_max_scales(higher_maxes);
    max_rotations(higher_maxes) = rotation;
end

```

```

function [result, boxes] = ...
    template_detector_demo(image, template, ...
                           scales, rotations, result_number)

% function [result, boxes] =
%     template_detector_demo(image, template, ...
%                             scales, rotations, result_number)
%
% returns an image that is a copy of the input image, with
% the bounding boxes drawn for each of the best matches for
% the template in the image, after searching all specified
% scales and rotations.

boxes = find_template(image, template, scales, ...
                     rotations, result_number);
result = image;

for number = 1:result_number
    result = draw_rectangle1(result, boxes(number, 1), ...
                           boxes(number, 2), ...
                           boxes(number, 3), boxes(number, 4));
end

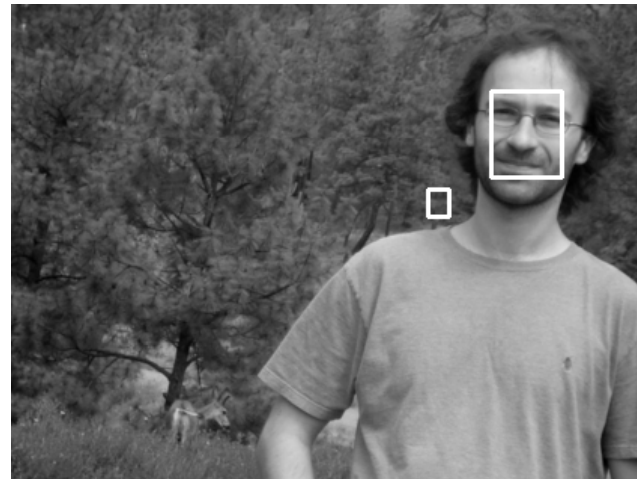
```

# Handling Rotations

```
load face_filter;  
photo = read_gray('vassilis2b.bmp');  
[resultb, boxes] = ...  
    template_detector_demo(photo, face_filter, 0.5:0.1:3., 0, 2);
```



photo



resultb

# Handling Rotations

```
load face_filter;  
photo = read_gray('vassilis2b.bmp');  
scales = 0.5:0.1:3;  
rotations = -10:5:10;  
[result2, boxes] = template_detector_demo(photo, face_filter, ...,  
                                          scales, rotations, 1);
```



photo



result2