

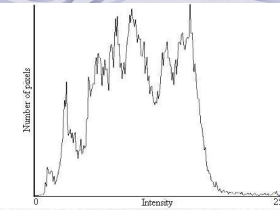
Histograms

- 1D Histograms
- 3D Histograms
- Equalisation
- Histogram Comparison
- Back Projection

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Slide 1

1D Histograms

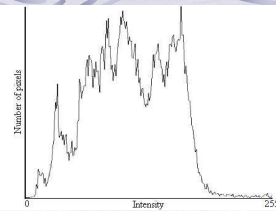


- Is this useful?
 - o Global information
 - o Useful for classification ?
 - o Not unique

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Slide 3

1D Histograms



- Determine the frequency of brightness values

- Initialise:
 - $h(z) = 0$ for all values of z
- Compute:
 - For all pixels (i,j) : $h(f(i,j))++$

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Slide 2

1D Histograms

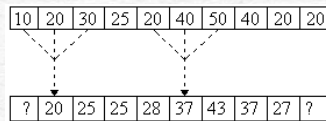
```
Mat display_image;
MatND gray_histogram;
const int* channel_numbers = { 0 };
float channel_range[] = { 0.0, 255.0 };
const float* channel_ranges = channel_range;
int number_bins = 64;
calcHist( &gray_image, 1, channel_numbers, Mat(),
          gray_histogram, 1, &number_bins, &channel_ranges );
OneDHistogram::Draw1DHistogram( &gray_histogram, 1,
                                display_image );
imshow("Greyscale histogram", display_image);
```

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Slide 4

1D Histograms – Smoothing

- Local minima and maxima are useful.
- But how can we deal with noise though?



- Smooth the histogram...
 - For all values v : $h_{\text{new}}(v) = (h(v-1) + h(v) + h(v+1)) / 3$
- What do we do at the ends?
 - Do not compute OR Wraparound OR Duplicate values OR Reflect values OR Constant values ??

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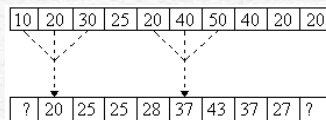
1D Histograms – Colour Histograms

- Determine histograms for each channel independently...
- Choice of colour space...

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Slide 7

1D Histograms – Smoothing

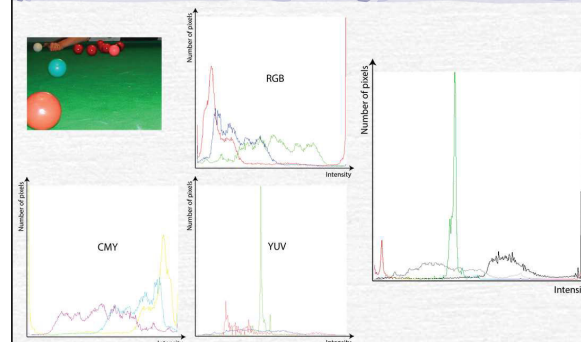


```
MatND smoothed_histogram = histogram[channel].clone();
for(int i = 1; i < histogram[channel].rows - 1; ++i)
    smoothed_histogram[channel].at<float>(i) =
        (histogram.at<float>(i-1) + histogram.at<float>(i) +
         histogram.at<float>(i+1)) / 3;
```

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1D Histograms – Colour Histograms



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Slide 8

1D Histograms – Colour Histograms

```
MatND* histogram = new MatND[image.channels()];
vector<Mat> channels(image.channels());
split( image, channels );
const int* channel_numbers = { 0 };
float channel_range[] = { 0.0, 255.0 };
const float* channel_ranges = channel_range;
int number_bins = 64;
for (int chan=0; chan < image.channels(); chan++)
    calcHist( &(channels[chan]), 1, channel_numbers, Mat(),
             histogram[chan], 1, &number_bins, &channel_ranges );
OneDHistogram::Draw1DHistogram(histogram,
                                image.channels(), display_image );
```

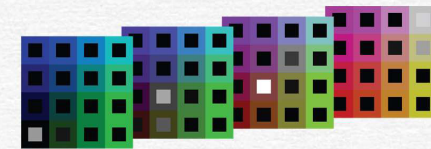
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3D Histograms

Reduce quantisation

- 6 bits = 262,144
- 4 bits = 4,096
- 2 bits = 64



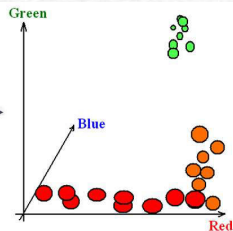
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3D Histograms

- Channels are not independent
- Better discrimination comes from considering all channels simultaneously
- Number of cells?

- 8 bits = 16,777,216



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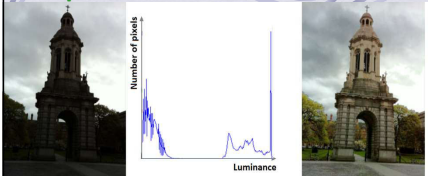
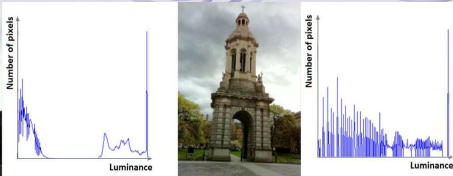
3D Histograms

```
MatND histogram;
int channel_numbers[] = { 0, 1, 2 };
int* number_bins = new int[image.channels()];
for (ch=0; ch < image.channels(); ch++)
    number_bins[ch] = 64;
float ch_range[] = { 0.0, 255.0 };
const float* channel_ranges[] = {ch_range, ch_range, ch_range};
calcHist( &image, 1, channel_numbers, Mat(), histogram,
          image.channels(), a_number_bins, channel_ranges );
```

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Slide 12

Equalisation

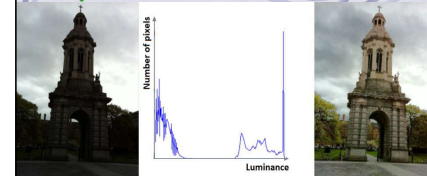




- ☞ If an image has insufficient contrast
- ☞ Human can distinguish 700-900 greyscales
- ☞ Evenly distribute the greyscales...
 - Result has missing greyscales
- ☞ Normally equalise only the greyscales / luminance

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Equalisation

```
vector<Mat> channels( hls_image.channels() );
split(hls_image, channels);
equalizeHist( channels[1], channels[1] );
merge( channels, hls_image );
```

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Equalisation

```
// Create a lookup table to map the luminances
// h[x] = histogram of luminance values image f(i,j).
pixels_so_far = 0
num_pixels = image.rows * image.cols
output = 0
for input = 0 to 255
  pixels_so_far = pixels_so_far + h[ input ]
  new_output = (pixels_so_far*256) / (num_pixels+1)
  LUT[ input ] = (output+1+new_output) / 2
  output = new_output
// Apply the lookup table LUT(x) to the image:
for every pixel f(i,j)
  f'(i,j) = LUT[ f(i,j) ]
```

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Histogram Comparison

- ☞ To find similar images...
 - Use metadata
 - Compare images
- ☞ Compare the colour distributions
- ☞ Need a metric for comparisons...

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Histogram Comparison



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Histogram Comparison – Earth Mover

- An alternative to the metrics is to compute the Earth Mover's Distance...
 - Minimum cost for turning a distribution into another distribution
 - Compare images
- 1D solution:
 - $EMD(-1) = 0$
 - $EMD(i) = h_1(i) + EMD(i-1) - h_2(i)$
 - Earth Mover's Distance = $\sum_i |EMD(i)|$
- Colour EMD is harder to compute...

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Histogram Comparison – Metrics

- $D_{Correlation}(h_1, h_2) = \frac{\sum_i (h_1(i) - \bar{h}_1)(h_2(i) - \bar{h}_2)}{\sqrt{\sum_i (h_1(i) - \bar{h}_1)^2 \sum_i (h_2(i) - \bar{h}_2)^2}}$
- $D_{Chi-Square}(h_1, h_2) = \sum_i \frac{(h_1(i) - h_2(i))^2}{(h_1(i) + h_2(i))}$
- $D_{Intersection}(h_1, h_2) = \sum_i \min(h_1(i), h_2(i))$
- $D_{Bhattacharyya}(h_1, h_2) = \sqrt{1 - \frac{1}{\sqrt{h_1 \cdot h_2 \cdot N^2}} \sum_i \sqrt{h_1(i) \cdot h_2(i)}}$
- where
 - N is the number of bins in the histograms,
 - $\bar{h}_k = \frac{\sum_i (h_k(i))}{N}$

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Histogram Comparison

```
normalize( histogram1, histogram1, 1.0);
normalize( histogram2, histogram2, 1.0);
double matching_score = compareHist( histogram1,
                                     histogram2, CV_COMP_CORREL);
```

We can also use Chi-Square (CV_COMP_CHISQR), Intersection (CV_COMP_INTERSECT) or Bhattacharyya (CV_COMP_BHATTACHARYYA) metrics or alternatively can use the Earth Mover's Distance function (EMD())

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Histogram Back Projection

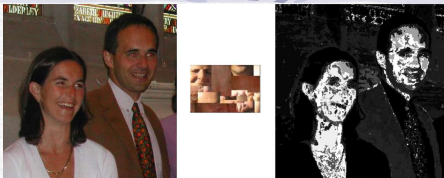
- A better approach to selecting colours (based on samples):
- 1. Obtain a representative sample set of the colours.
- 2. Histogram those samples.
- 3. Normalize that histogram so that the maximum value is 1.0.
- 4. Back project the normalized histogram onto any image $f(i,j)$.
- 5. This will provide a 'probability' image $p(i,j)$ which indicates the similarity between $f(i,j)$ and the sample set.



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Histogram Back Projection



```
calcHist( &hls_samples_image, 1, channel_numbers, Mat(),  
         histogram,image.channels(),number_bins,channel_ranges);  
normalize( histogram, histogram, 1.0);  
Mat probabilities = histogram.BackProject( hls_image );
```

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