

Digital Video Processing

Outline

- Video Compression
- Segmentation in Video
- Foreground and Background Segmentation Using Motion and Color
- Layered Video
- Digital Watermarking
- Video Streaming

Video Compression

- Problem:

- Raw video contains an immense amount of data.
- Communication and storage capabilities are limited and expensive.

- Example HDTV video signal:

- 720x1280 pixels/frame, progressive scanning at 60 frames/s:

$$\left(\frac{720 \times 1280 \text{ pixels}}{\text{frame}} \right) \left(\frac{60 \text{ frames}}{\text{sec}} \right) \left(\frac{3 \text{ colors}}{\text{pixel}} \right) \left(\frac{8 \text{ bits}}{\text{color}} \right) = 1.3 \text{ Gb/s}$$

- 20 Mb/s HDTV channel bandwidth

- Requires compression by a factor of 70 (equivalent to .35 bits/pixel)

Video Compression (Contd..)

Bandwidth Reduction

Application	Data Rate	
	Uncompressed	Compressed
Video Conference <i>352 x 240 @ 15 fps</i>	30.4 Mbps	64 - 768 kbps
CD-ROM Digital Video <i>352 x 240 @ 30 fps</i>	60.8 Mbps	1.5 - 4 Mbps
Broadcast Video <i>720 x 480 @ 30 fps</i>	248.8 Mbps	3 - 8 Mbps
HDTV <i>1280 x 720 @ 60 fps</i>	1.33 Gbps	20 Mbps

Video Compression (Contd..)

Standards

STANDARD	APPLICATION	BIT RATE
JPEG	Continuous-tone still-image compression	Variable
H.261	Video telephony and teleconferencing over ISDN	p x 64 kb/s
MPEG-1	Video on digital storage media (CD-ROM)	1.5 Mb/s
MPEG-2	Digital Television	> 2 Mb/s
H.263	Video telephony over PSTN	< 33.6 kb/s
MPEG-4	Object-based coding, synthetic content, interactivity	Variable
H.264	From Low bitrate coding to HD encoding, HD-DVD, Surveillance, Video conferencing.	Variable

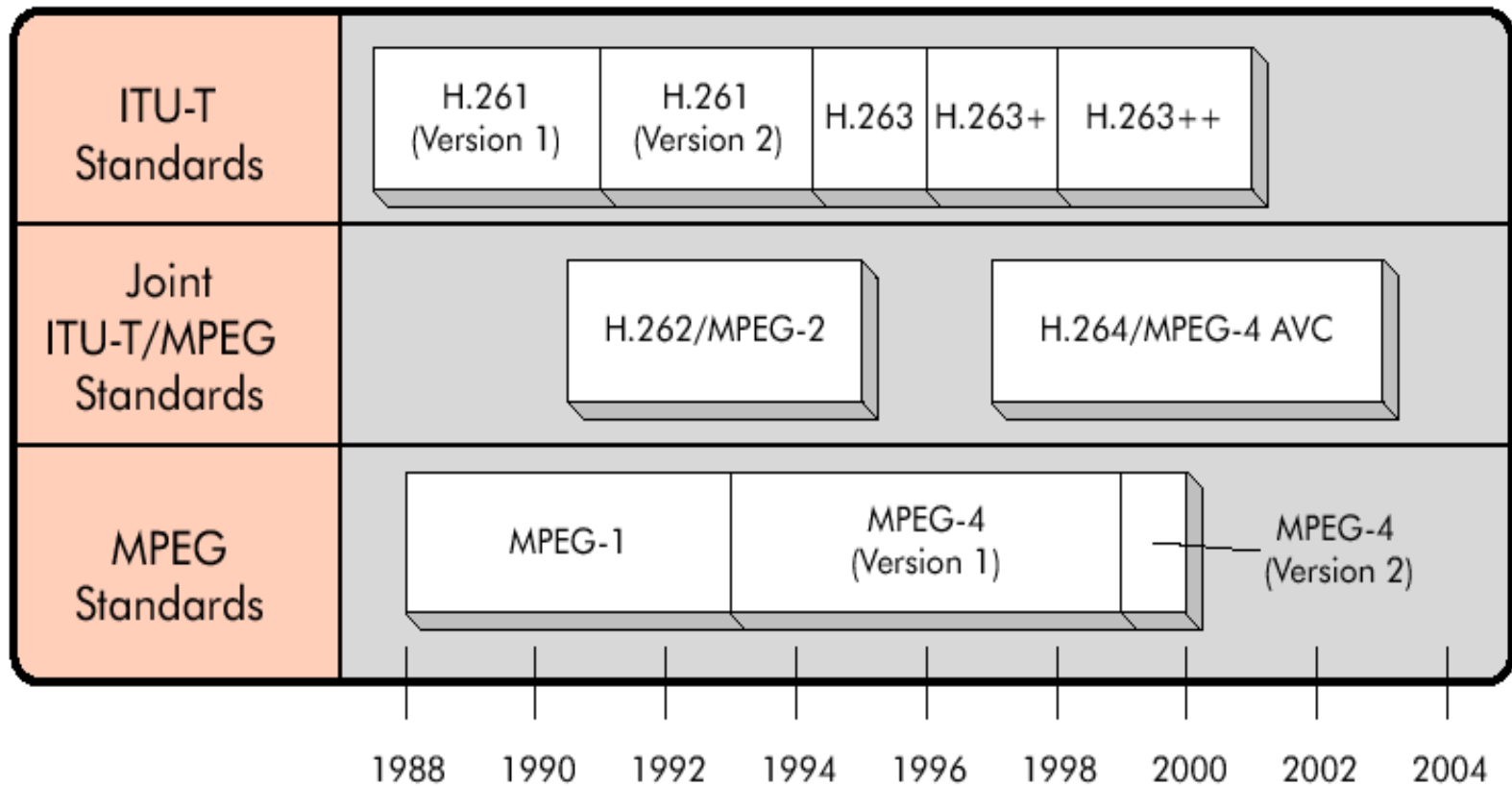
Video Compression (Contd..)

Goal of standards:

- Ensuring interoperability – Enabling communication between devices made by different manufacturers.
- Promoting a technology or industry.
- Reducing costs.

Video Compression (Contd..)

History of Video Standards



Video Compression (Contd..)

What Do the Standards Specify?

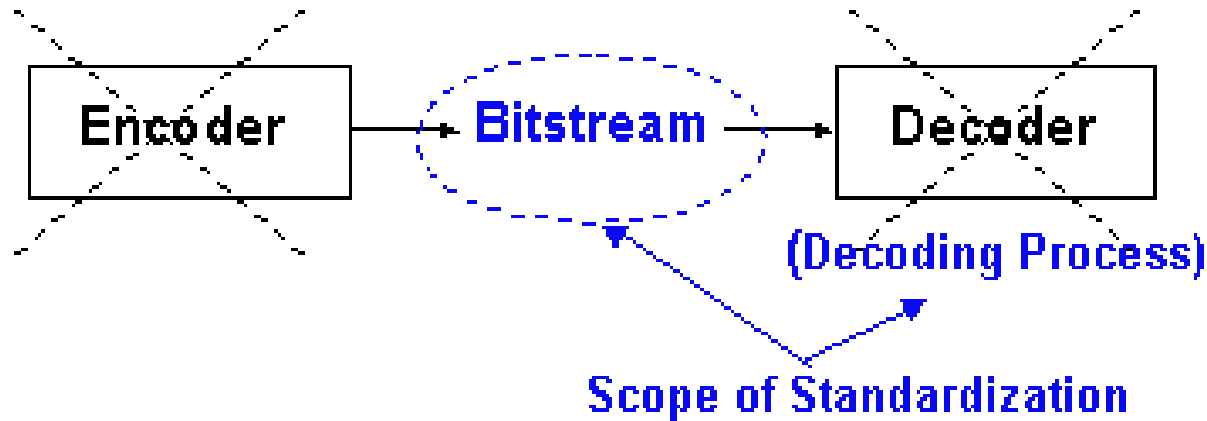
- A video compression system consists of the following:
 - An encoder
 - Compressed bit-streams
 - A decoder

- What parts of the system do the standards specify?

Video Compression (Contd..)

What Do the Standards Specify? (Contd..)

- Not the encoder, not the decoder.



Video Compression (Contd..)

What Do the Standards Specify? (Contd..)

- Just the bit-stream syntax and the decoding process, for example it tells to use IDCT, but not how to implement the IDCT.
- Enables improved encoding and decoding strategies to be employed in a standard-compatible manner.

Video Compression (Contd..)

Achieving Compression

- Reduce redundancy and irrelevancy.
- Sources of redundancy:
 - Temporal – Adjacent frames highly correlated.
 - Spatial – Nearby pixels are often correlated with each other.
 - Color space – RGB components are correlated among themselves.
- Irrelevancy – Perceptually unimportant information.

Video Compression (Contd..)

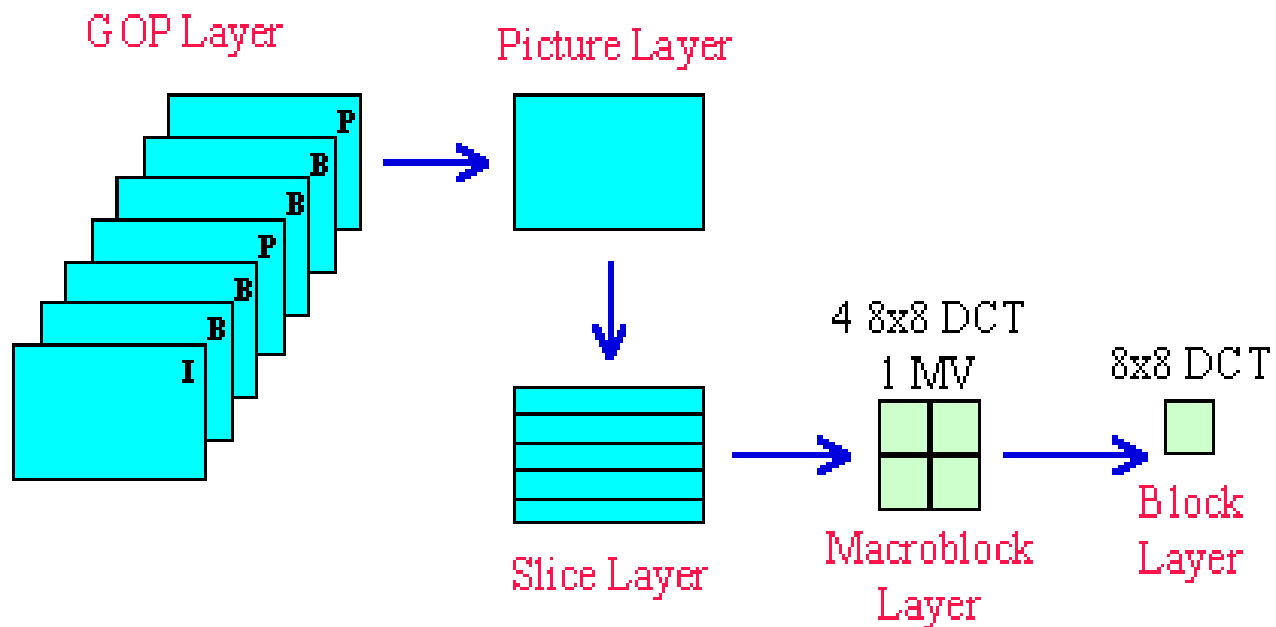
Basic Video Compression Architecture

- Exploiting the redundancies
 - Temporal – MC-prediction and MC-interpolation
 - Spatial – Block DCT
 - Color – Color space conversion
- Scalar quantization of DCT coefficients
- Run-length and Huffman coding of the non-zero quantized DCT coefficients

Video Compression (Contd..)

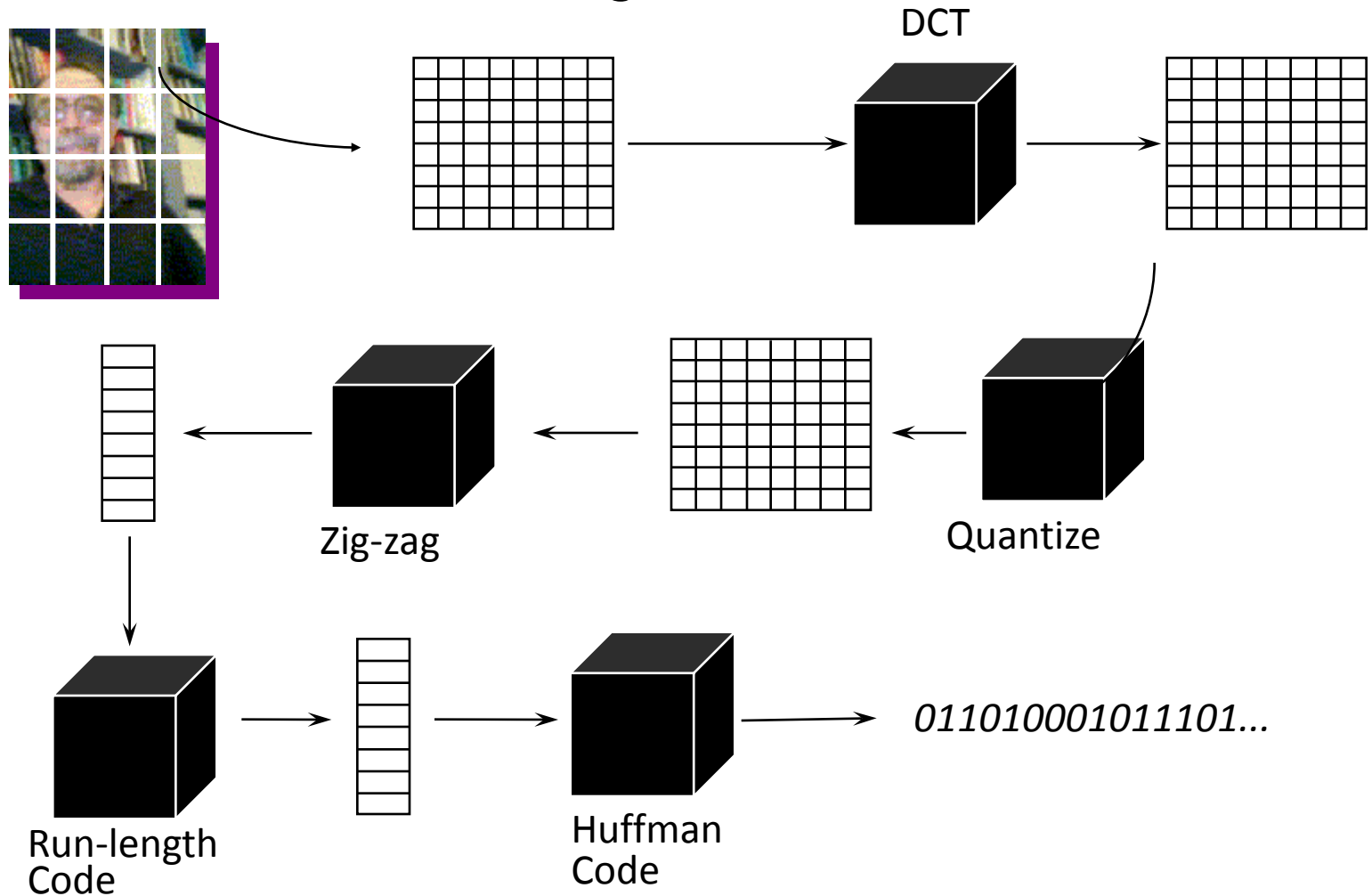
Video Structure

● MPEG Structure



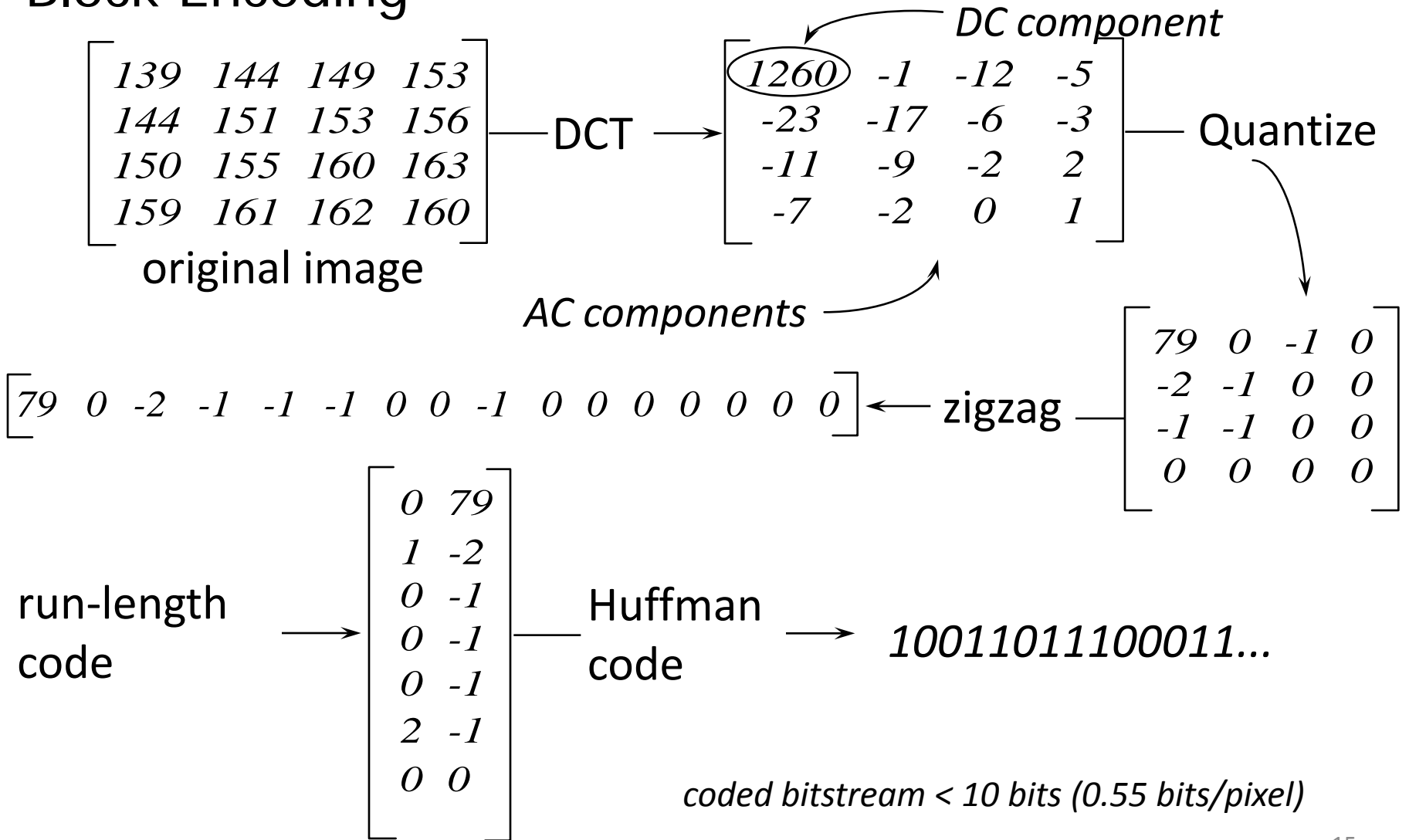
Video Compression (Contd..)

Block Transform Encoding



Video Compression (Contd..)

Block Encoding



Video Compression (Contd..)

Result of Coding/decoding

$$\begin{bmatrix} 139 & 144 & 149 & 153 \\ 144 & 151 & 153 & 156 \\ 150 & 155 & 160 & 163 \\ 159 & 161 & 162 & 160 \end{bmatrix}$$

original block

$$\begin{bmatrix} 144 & 146 & 149 & 152 \\ 156 & 150 & 152 & 154 \\ 155 & 156 & 157 & 158 \\ 160 & 161 & 161 & 162 \end{bmatrix}$$

reconstructed block

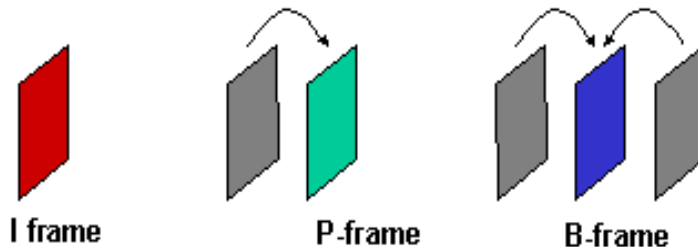
$$\begin{bmatrix} -5 & -2 & 0 & 1 \\ -4 & 1 & 1 & 2 \\ -5 & -1 & 3 & 5 \\ -1 & 0 & 1 & -2 \end{bmatrix}$$

errors

Video Compression (Contd..)

Video Compression

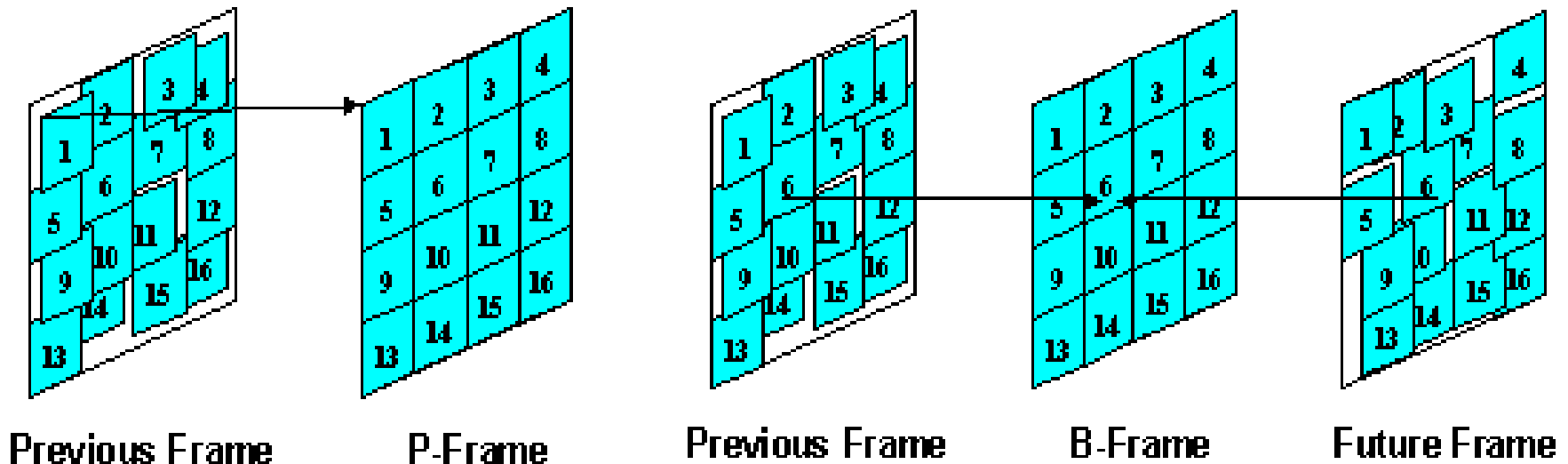
- Main addition over image compression
 - Exploit the temporal redundancy
- Predict current frame based on previously coded frames
- Types of coded frames:
 - I-frame – Intra-coded frame, coded independently of all other frames
 - P-frame – Predictively coded frame, coded based on previously coded frame
 - B-frame – Bi-directionally predicted frame, coded based on both previous and future coded frames



Video Compression (Contd..)

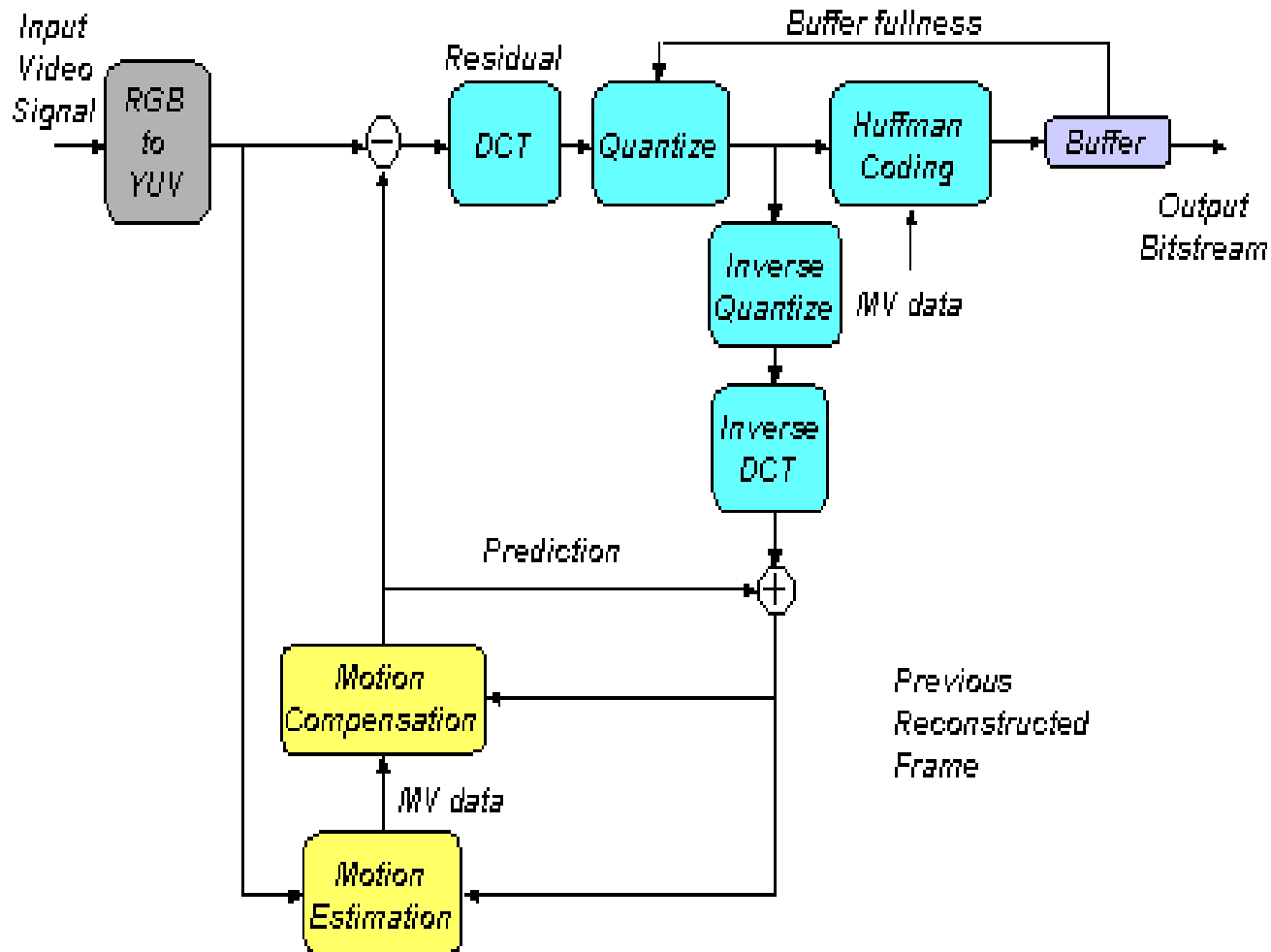
Motion Compensated Prediction (P and B Frames)

- Motion compensated prediction – predict the current frame based on a reference frame while compensating for the motion
- Examples of block-based motion-compensated prediction



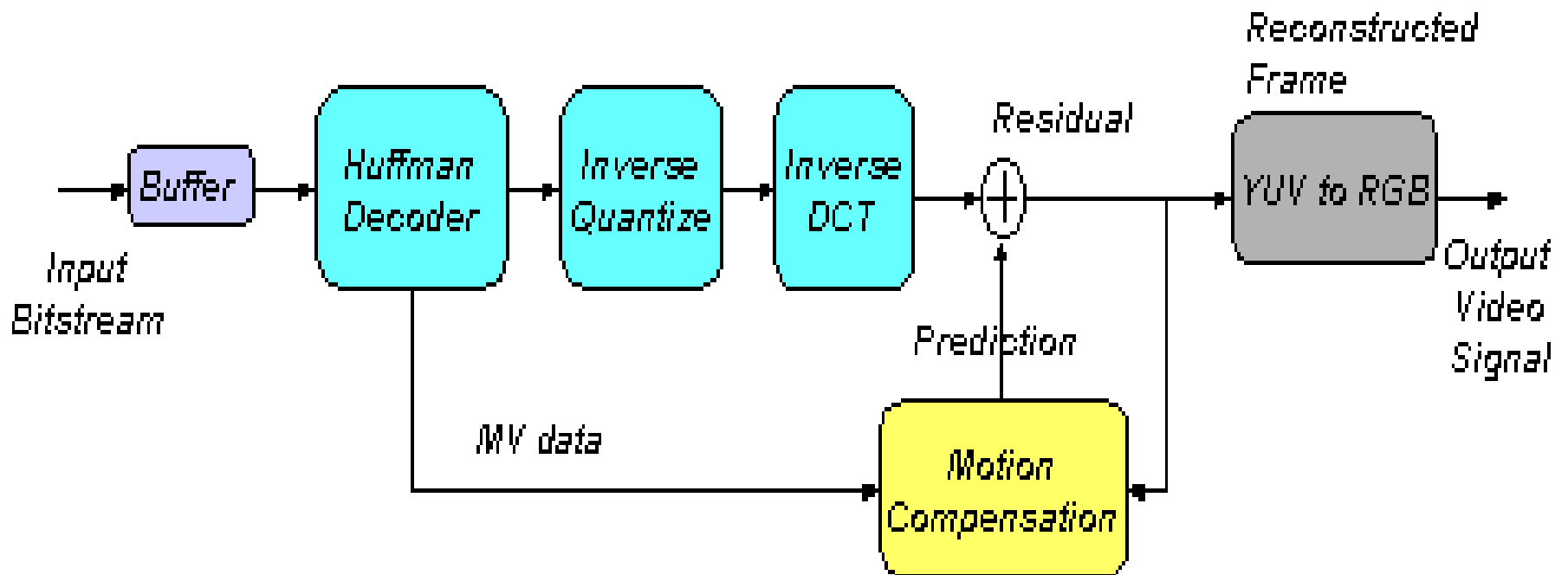
Video Compression (Contd..)

Example Video Encoder



Video Compression (Contd..)

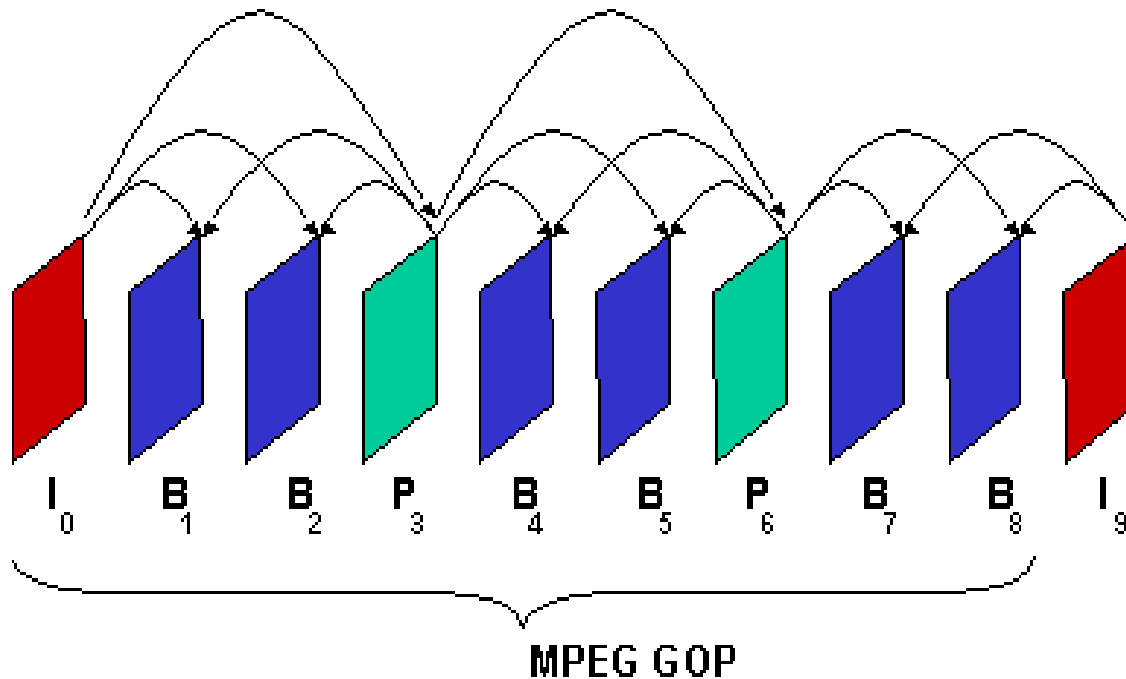
Example Video Decoder



Video Compression (Contd..)

Group of Pictures (GOP) Structure

- Enables random access into the coded bit-stream.
- Number of B frames and impact on search range.



Video Compression (Contd..)

Current Video Compression Standards

- Classification & Characterization of different standards
 - Based on the same fundamental building blocks
 - Motion-compensated prediction and interpolation
 - 2-D Discrete Cosine Transform (DCT)
 - Color space conversion
 - Scalar quantization, run-length, and Huffman coding
 - Other tools added for different applications
 - Progressive or interlaced video
 - Improved compression, error resilience, scalability, etc

Video Compression (Contd..)

Compressed Video

- To manage storage and transmission
- Several standards but MPEG is widely used
- Different MPEG versions
 - MPEG-1
 - MPEG-2
 - MPEG-4 “Coded representation of media objects”
 - MPEG-7 “Multimedia Content Description Interface”

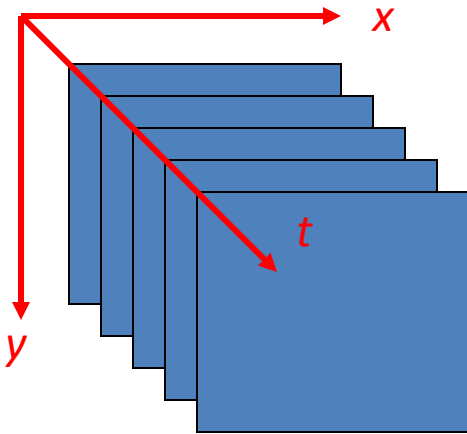
Outline

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- Video Streaming

Segmentation in Video

Videos

- Videos are **Image Sequences over Time**



- An image is a function over x and y :

$$f(x, y)$$

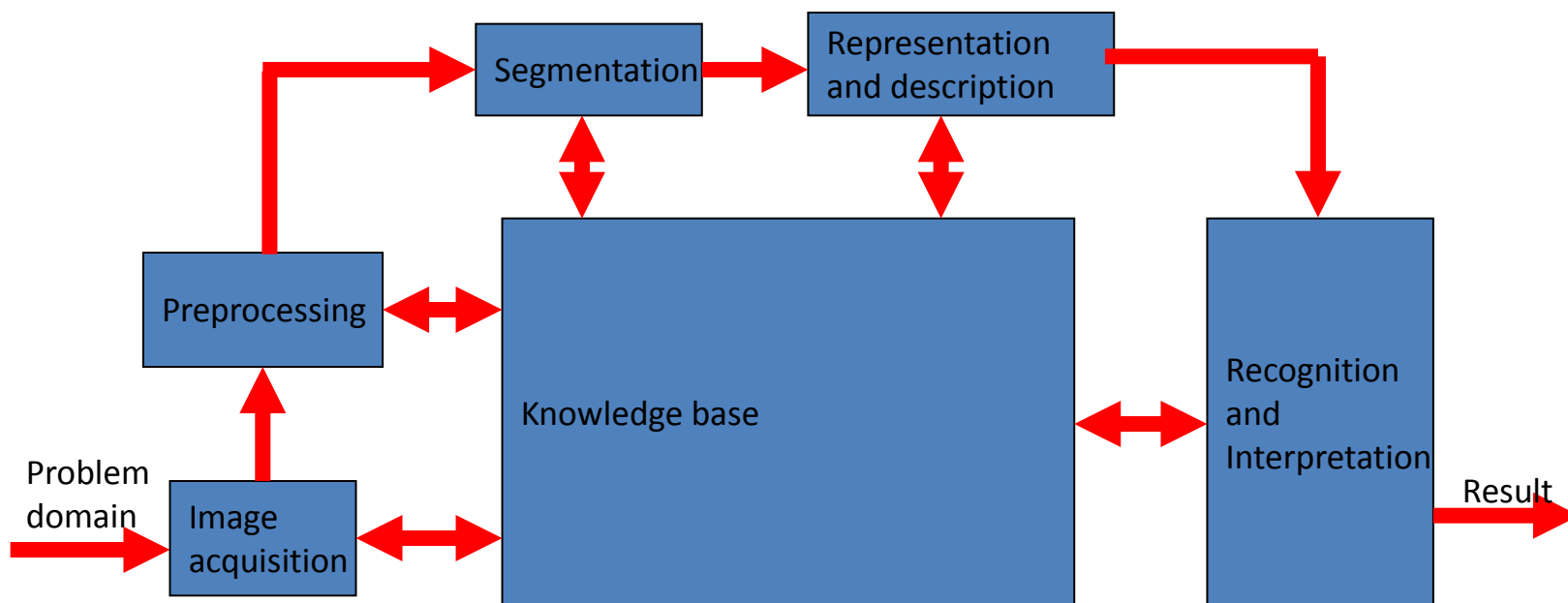
- A video is a function of images over time t :

$$f(x, y, t) = f_t(x, y)$$

- At each time step t we have an image
- Framerate = the number of images per second, e.g., 25 Images/s

Segmentation in Video (Contd..)

- General Segmentation is **application-dependent!**
Knowledge base!!
- Finding the object(s)
 - Preprocessing, Segmentation
- Tracking = follow the object(s) over time:
 - being able at each time t to give, e.g, position, color, orientation, etc.
 - Representation, Description



Segmentation in Video (Contd..)

Lots of applications!! One is:

- Separation of **Foreground** (object) and **Background** (everything else = noise)
- Result could be a
 - Binary image, containing foreground only
 - Probability image, containing the likelihood of each pixel being foreground
- Useful for further processing, such as using silhouettes, etc.
- Approaches
 - Motion-based
 - Color-based
 - Some approaches can learn!

Foreground-Background Segmentation using Motion and Color

- Motion-based
 - Image differencing
 - Model-free
 - No learning
- Color-based
 - Background subtraction
 - Background used as a model
 - No learning
 - Advanced background subtraction
 - Background is learned
 - Very advanced background subtraction
 - Background is learned

Foreground-Background Segmentation using Motion and Color (Contd..)

Image Differencing

- The motion in an image can be found by subtracting the current image from the previous image
- Algorithm
 1. Save image in last frame
 2. Capture current camera image
 3. Subtract image (= difference = motion)
 4. Threshold
 5. Delete noise

Foreground-Background Segmentation using Motion and Color (Contd..)

3. Subtract Image

- Compute pixel-wise
- Subtract previous image from input image:

$$F(x, y) = I(x, y) - B(x, y)$$

- Usually the absolute distance is applied

$$|F(x, y)|$$

1. Save image in last frame
2. Capture camera image
3. **Subtract image**
4. Threshold
5. Delete noise

Foreground-Background Segmentation using Motion and Color (Contd..)

4. Threshold

- Decide, when a pixel is supposed to be considered as a background pixel, or when it is to be considered as a foreground pixel:
- Pixel (x, y) is foreground pixel, if
$$|F(x, y)| > TH$$
- Pixel (x, y) is background pixel, if
$$|F(x, y)| \leq TH$$
- Problem: What TH??

1. Save image in last frame
2. Capture camera image
3. Subtract image
4. **Threshold**
5. Delete noise

Foreground-Background Segmentation using Motion and Color (Contd..)

5. Deleting Noise

- Singular pixels are likely to appear:
 - Pixel-noise!!
- Apply Median filter:
 - Depending on filter size, bigger spots can be erased
- Alternative: Morphologic

1. Save image in last frame
2. Capture camera image
3. Subtract image
4. Threshold
5. **Delete noise**

Foreground-Background Segmentation using Motion and Color (Contd..)

Background Subtraction

- Foreground is moving, background is stable
- Algorithm
 - 1. Capture image containing background**
 2. Capture camera image
 3. Subtract image (= difference = motion)
 4. Threshold
 5. Delete noise

Foreground-Background Segmentation using Motion and Color (Contd..)

Advanced Background Subtraction

- What if we have small motion in the background?
 - Bushes, leaves, etc. and noise in the camera/lighting
 - (show histo patch)
- **Learn** the background
- Capture N images and calculate the average background image (no object present)

1. **Calculate average background image**
2. Capture camera image
3. Subtract image (= motion)
4. Threshold
5. Delete noise

Foreground-Background Segmentation using Motion and Color (Contd..)

Very Advanced Background Subtraction

- **Use Neighborhood relation!!**
 - Compare pixel with its neighbors!!
 - **Weight them!!**
- **Learn the background and its variations!!**
 - E.g. Gaussian models (mean,var) for each pixel!!!
 - E.g. a Histogram for each Pixel
 - The more images you train on the better!!
 - Idea:
 - Some pixel may vary more than other pixels
 - Algorithm:
 - Consider each pixel (x,y) in the input image and check, how much it varies with respect to the mean and variance of the learned Gaussian models?

1. Calculate mean and variance for each pixel
2. Capture camera image
3. Subtract image (= motion)
4. **Weight the distances (new)**
5. **Threshold according to variance**
6. Delete noise

Foreground-Background Segmentation using Motion and Color (Contd..)

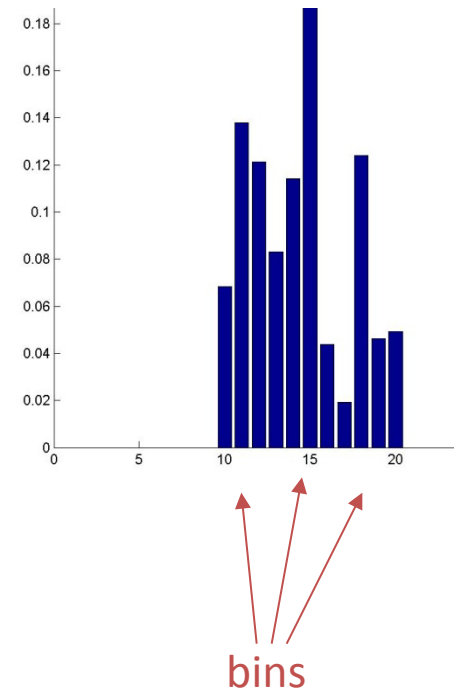
Segmentation using Color Histogram

- Given an object, defined by its color Histogram
- Algorithm:
 - Segment each pixel in the input image according to the histogram
- The result: binary or probability image.
 - White pixels: pixel in input image had color as defined by the histogram
 - Black pixel: pixel in input image did not have color defined by the histogram

Foreground-Background Segmentation using Motion and Color (Contd..)

Learning the Color Histogram

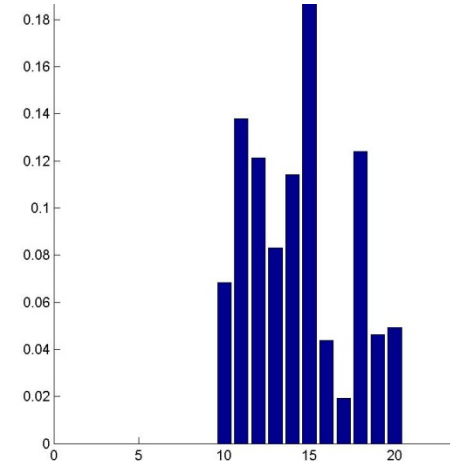
- The color histogram summarizes the color distribution in an image (region)
- There is a histogram bin for each possible color!
- The columns of the histogram are high for those colors appearing often in the image (region) and low for those appearing seldom in the image.
- The more images you train on the better!!



Foreground-Background Segmentation using Motion and Color (Contd..)

Color Segmentation with Histograms

- Algorithm:
 - For each pixel in the input $I(x,y)$
 - Go to the bin having the same color as the pixel $H(I(x,y))$
 - Assign the probability value at the bin to the output image: $O(x,y)$
 - $O(x,y) = H(I(x,y))$
 - Using the above leads to a probability image
 - Run a threshold on the output image to get a binary image



Foreground-Background Segmentation using Motion and Color (Contd..)

Segmentation with Gaussian Distribution

- Given an object, calculate the mean and standard deviation of its color
 - Represent each color component (e.g., R,G,B) by a **Gaussian model**
 - That is, a Gaussian distribution: (mean,sigma)
- The same principle as in “very advanced background subtraction” !!
- Algorithm:
 - Given an input image, segment each pixel by comparing it to the Gaussian models of the object color
- For example:
 - If $Th_{min} < dist < Th_{max} \Rightarrow$ object pixel
 - $Th_{min} = mean - 2s$
 - $Th_{max} = mean + 2s$

Foreground-Background Segmentation using Motion and Color (Contd..)

Weight the Distances, Correlation between Pixel values

- If one pixel is considered to be a foreground pixel
 - Its neighbor is also likely to be a foreground pixel
 - If its neighbor is not considered to be a foreground pixel, one of the two might be wrong
 - Neighboring pixels are highly **correlated** (similar)

1. Calculate mean and variance for each pixel
2. Capture camera image
3. Subtract image (= motion)
4. **Weight the distances (new)**
5. Threshold according to variance
6. Delete noise

Foreground-Background Segmentation using Motion and Color (Contd..)

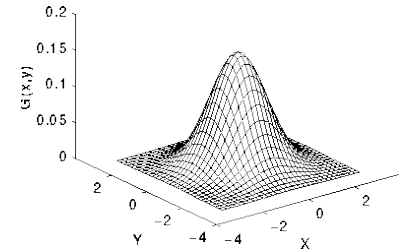
Weight the Distances

- What does a pixel say about a foreground pixel, that is further away?
 - Pixels with an increasing distance to each other are saying less about each other
 - The correlation between pixels decreases with distance

1. Calculate mean and variance for each pixel
2. Capture camera image
3. Subtract image (= motion)
4. **Weight the distances (new)**
5. Threshold according to variance
6. Delete noise

Foreground-Background Segmentation using Motion and Color (Contd..)

- Use a Gaussian for Weighting
- To test a pixel $I(x,y)$
 - Center a Gaussian on this pixel and weight the neighboring pixels accordingly => Convolution!
- 1D example:



Signal:

60	48	2	3	1	222	100
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Gaussian filter:

0.25	0.5	0.25
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Output:

	39.5	13.75	2.25	56.75	136.3	
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Foreground-Background Segmentation using Motion and Color (Contd..)

- **Likelihood image** or **Probability image**, containing the likelihood for each pixel of being a background/foreground pixel

1. Calculate mean and variance for each pixel
2. Capture camera image
3. Subtract image (= motion)
4. **Weight the distances (new)**
5. Threshold according to variance
6. Delete noise