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9 Video Compression

9.1 Perception of motion

Perception of motion: Human visual system is specifically sensitive to motion. Eves follow motion automatically. Some distortions are not as percoivable as in

image coding (would be if we froze frame). No good psycho-visual model avaivable. Vusal perception is limited to < 24 Hz. Asuccession of images will be perceived as continuous if frequency is sufficiely high, Cinema 2424 Hz, TV 25 Hz or 50 Hz. We still nee to avoid alias- 16 × 16 pixels ing (wheel effect). High-rendering frame- 2 Describe motion of each block rates desired in computer games (needed due to absence of motion blur). Flicker good, robust performance. can be perceived up to > 60 Hz in particular in periphery. Issue addressed by 100 Hz TV.

9.2 Interlaced video format

Two temporarlly shifted half images, in- mation crease of frequency 25 Hz → 50 Hz. Reduction of spatial resolution. Full image representation: progressive.

9.3 Why compress video?

Raw HD TV signal 720p@50 Hz: 1280 · 720 · 50 · 24 bits/s

= 1105920000 bits/s > 1 Gb/s

Only 20 Mb/s HDTV channel bandwdith requires compression of factor of 60 (0.4 bits/pixel on average)

9.4 Lossy video compression

Take advantage of redundancy, Spatial corpercentuall unimportant details

tage of similarity between successive frames

Usually high frame rate: Significant tem- $n_1, n_2 \in Block$ poral redundancy. Possible representations along temporal dimension:

1 Transform/subband methods: Good for textbook case of constant velocity uni- = $\sum |f(n_1, n_2, k_{cur})|$ form global motion. Inefficient for nonuni- $n_1, n_2 \in Block$ form motion, i.e. real-world motion, Requires large number of frame stores which leads to delay. (Memory cost may alse be an issue.) Is ineffective for many scene changes or high motion.

2 Prodictive methods: Good performance using only 2 frame stores. How-

Full search: Examine all candidate ever, simple frame differencing is not blocks

Exploit the temporal redundancy rent frame based on previously coded frames

 I-frame: Intra-coded frame, coded independently of all other frames.

P-frame: Predictively coded frame, coded based on previously coded frame I or P. Can send motion vector plus changes

3 B-frame: Bi-directionally predicted frame, coded based on both previous and future coded frames I and P. In case something is uncovered.

Simple frame differencing fails when there is motion. Must account for motion. → Motion-compensated (MC) prediction. MC-prediction generally provides significant improvements. Questions: How can we estimate motion? How can we form MC-prediction?

1 Partition video into moving objects 2 describe object motion → Generally

Block-Matching Motion Estimation

Partition each frame into blocks, e.g.

→ No object identification required and

9.5 Block-matching motion esti-

Translational motion

 $f(n_1, n_2, k_{\text{cur}})$

 $= f(n_1 - mv_1, n_2 - mv_2, k_{ref}).$

rithm 1 Divide current frame into non-overlapping $N_1 \times N_2$ blocks

For each block, find the best matching block in reference frame

9.5.1 Determining the best matching

relation between neighboring pixels. Tem- For each block in the current frame, search poral correlation between frames. Drop for best matching block in the reference

ral Redundancy Take advan- Metrics for determining "best match": 9.7 Block Matching Algorithm

 $-f(n_1 - mv_1, n_2 - mv_2, k_{\text{ref}})|$.

Partial (fast) search: Examine a

Motion vector Estimate of motion for

9.6 Motion vector and motion

relative horizontal and vertical offsets

 (mv_1, mv_2) , or motion, of a given

tion vectors for all the blocks in a frame.

Motivation: Motion is not limited to in-

· Fractional MVs are used to represent

Improved performance (extra complex-

· Half-pixel ME used in most standards:

· Why are half-pixel motion vectors bet-

ter? They can capture half-pixel motion. Averaging effect (from spatial interpo-

lation) reduces prediction error -> Im-

proved prediction. For noisy sequences,

averaging effect reduces noise -> Im-

1 Coarse step: Perform integer mo-

tion estimation on blocks; find best inte-

b Compare current block to interpolated

Choose the integer or half-pixel offset that provides best match Typically, bilin-

ear interpolation is used for spatial inter-

spatially interpolated.

the sub-pixel motion.

ity is worthwhile)

proved compression

mation Algorithm

ger-pixel MV

half-pixel MV

in reference frame

reference frame block.

MPEG-1/2/4

otion vector field Collection of mo-

Example: (±7,±7)search area
 Dots represent search points
 Search performed in 3 steps (coarse-to-fine):

Step 1: (+4 nixels) Step 2: • (±2pixels) Step 3: • (±1pixels)

· Best match is found at each step

Next step: Search is centered around the best match of prior st

tion vector Expresses

block from one frame to another

carefully selected subset.

best matching block.

vector field

 $= \sum [f(n_1, n_2, k_{\text{cur}})]$ vector accuracy

e blocks All blocks in, e.g.

Done typically only from lu- $-f(n_1 - mv_1, n_2 - mv_2, k_{\text{ref}})]^2$

es 1 Good, robust performance for compression.

2 Resulting motion vector field is easy to represent (one MV per block) and useful for compression

Block size, search range, motion

3 Simple, periodic structure, easy VLSI implementations

 Assumes translational motion model → Breaks down for more complex mo-

Ofter produces blocking artifacts (OK) for coding with Block DCT)

ional MC prediction is uset to estimate a block in the current frame from a block in:

Previous frame

 Future frame 3 Average of ablock from the previous

frame and a block from the future frame 4 Neither, i.e. code current block without prediction

Example: Prediction with P- and B-frames Motion compensated prediction: Predict the current frame based on reference frame(s) while compensating for the mo-

Examples of block-based motion-compensated prediction (P-frame) and bi-directional prediction (B-frame).

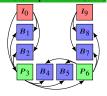
9.8 Frame types

Main addition over image compression: Exploit the temporal redundancy. Predict current frame based on previously coded frames. Three types of coded frames:

 I-frame: Intra-coded frame, coded independently of all other frames

2 P-frame: Predictively coded frame. teger-pixel offsets. However, video is only coded based on previously coded frame known at discrete pixel locations. To es-

3 B-frame: Bi-directionally predicted timate sub-pixel motion, frames must be frame, coded based on both previous and future coded frames



Starts with an I-frame, ends with frame 9.6.1 Practical Half-Pixel Motion Estiright before next I-frame. "Open" ends in B-frame, "closed" in P-frame. MPEG Encoding a parameter, but "typical": Half-pixel ME (coarse-fine) algorithm:

IBBPBBPBBI,

IBBPBBPBBPBBI.

Why not all P and B frames after initial I? 2 Fine step: Refine estimate to find best (Because then the whole movie depends on the accuracy of the first frame. Data Spatially interpolate the selected region loss possible etc.)

I: $\frac{1}{7}$, P: $\frac{1}{20}$, B: $\frac{1}{50}$, Average: $\frac{1}{27}$.

9.9 Summary of Temporal Processing

Use MC-prediction (P and P frames) to

reduce temporal redundancy

MC-prediction usually performs well; In compression have a second changce to recover when it performs badly.

3 MC-prediction yields a Motion vectors

MC-prediction error or residual → Code error with conventional image coder Sometimes MC-prediction may per-

a Examples: complex motion, new imagery (occlusions)

Approach: 1. Identify frame or indi-

vidual blocks where prediction fails 2. Code without prediction

9.10 Basic Video Compression Architecture

Exploiting the reduncancies: a Temporal: MC-prediction (P and B frames)

b Spatial: Block DCT

Color: color space conversion 2 Scalar quantization of DCT coefficients

3 Zigzag scanning, runlength and Huffman coding of the nonzero quantized DCT coefficients

10 Ouestions

$$I_{\text{comp}} = I_{\alpha}I_{\alpha} + (1 - I_{\alpha})I_{b}$$

- · MAP. Maximum a posteriori detec-
- graph cuts
- Solve MRFs with graph cuts
- impulse response t(-x, -y)
- Canny nonmaxima suppression
- Entropy Coding (Huffman code)
- Aperture problem: normal flow
- · Lucas-Kanade: Iterative refinement/local gradient method
- Coarse-to-fine-estimation

A Big equations

$$\mathcal{F}[h](u,v) = \frac{1}{2\ell} \int_{-\ell}^{\ell} dx_1 \exp(-i2\pi u x_1) \cdot \int_{-\infty}^{\infty} dx_2 \, \delta(x_2) \exp(-i2\pi v x_2)$$

$$= \operatorname{sinc}(2\pi u \ell)$$

$$E = \iint dx dy \left[\left(\frac{\partial I}{\partial x} \frac{dx}{dt} + \frac{\partial I}{\partial y} \frac{\partial y}{\partial t} + \frac{\partial I}{\partial t} \right)^{2} + \alpha^{2} (\|\nabla \dot{x}\| + \|\nabla \dot{y}\|)^{2} \right]$$

$$\mathbf{v} = \left(\frac{\sum_{i} w_{i} I_{X}(q_{i})^{2}}{\sum_{i} w_{i} I_{X}(q_{i}) I_{Y}(q_{i})} \sum_{i} w_{i} I_{Y}(q_{i}) I_{Y}(q_{i})\right)^{-1} \cdot \left(-\sum_{i} w_{i} I_{X}(q_{i}) I_{I}(q_{i})\right)$$