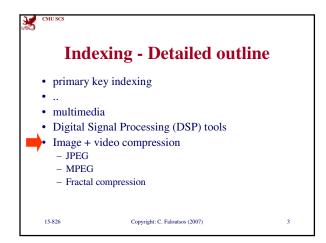


15-826: Multimedia Databases and Data Mining

Compression: JPEG, MPEG, fractal C. Faloutsos





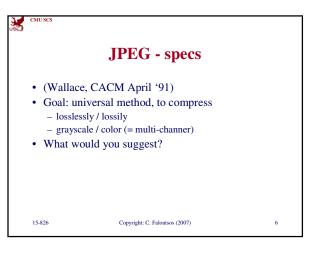


Motivation

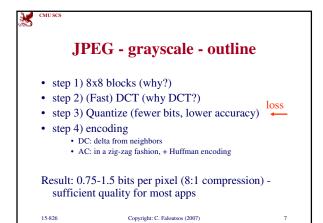
• Q: Why study (image/video) compression?

• A1: feature extraction, for multimedia data mining

• A2: (lossy) compression = data mining!



15-826



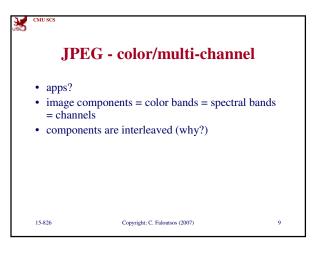
JPEG - grayscale - lossless

• Predictive coding: X = f(A, B, C)eg. X = (A+B)/2, or?

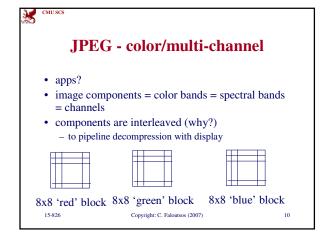
• Then, encode prediction errors

Result: typically, 2:1 compression

15-826 Copyright: C. Faloutsos (2007) 8



C. Faloutsos 15-826





JPEG - color/multi-channel

- · tricky issues, if the sampling rates differ
- · Also, hierarchical mode of operation: pyramidal structure
 - sub-sample by 2
 - interpolate
 - compress the diff. from the predictions

15-826

Copyright: C. Faloutsos (2007)

11

12

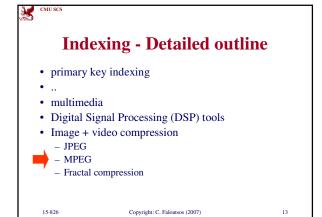


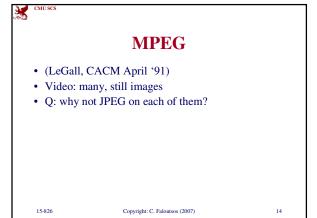
JPEG - conclusions

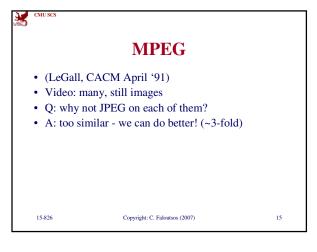
- grayscale, lossy: 8x8 blocks; DCT; quantization and encoding
- · grayscale, lossless: predictions
- color (lossy/lossless): interleave bands

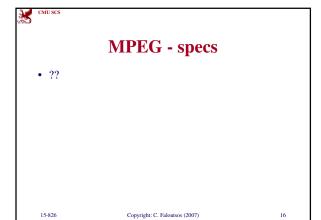
15-826

1









3 °

CMU SCS

MPEG - specs

- · acceptable quality
- asymmetric/symmetric apps (#compressions vs #decompressions)
- Random access (FF, reverse)
- audio + visual sync
- error tolerance
- variable delay / quality
- editability

15-826

Copyright: C. Faloutsos (2007)

17

18

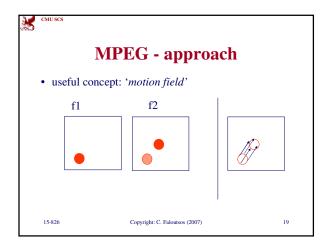


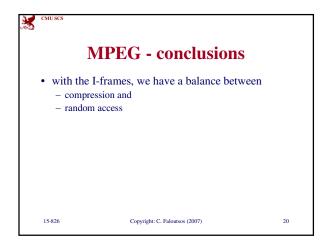
CMU SCS

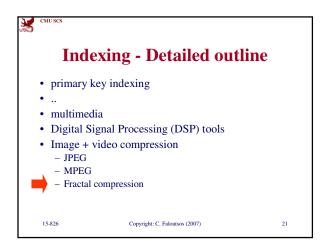
MPEG - approach

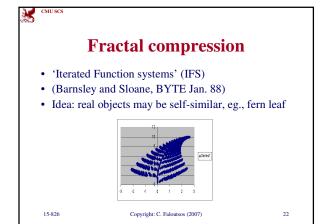
- main idea: balance between inter-frame compression and random access
- thus: compress some frames with JPEG (I-frames)
 - rest: prediction from motion, and interpolation
 - P-frames (predicted pictures, from I- or P-frames)
 - B-frames (interpolated pictures never used as reference)

15-826









Fractal compression

• simpler example: Sierpinski triangle.

- has details at every scale -> DFT/DCT: not good

- but is easy to describe (in English)

• There should be a way to compress it very well!

• Q: How??

Copyright: C. Faloutsos (2007)

15-826

Fractal compression

• simpler example: Sierpinski triangle.

- has details at every scale -> DFT/DCT: not good

- but is easy to describe (in English)

• There should be a way to compress it very well!

• Q: How??

• A: several, affine transformations

• Q: how many coeff. we need for a (2-d) affine transformation?

25

27



Fractal compression

- A: 6 (4 for the rotation/scaling matrix, 2 for the translation)
- $(x,y) \rightarrow w((x,y)) = (x', y')$
 - x' = a x + b y + e
 - y' = c x + d y + f
- for the Sierpinski triangle: 3 such transformations which ones?

15-826

Copyright: C. Faloutsos (2007)

Fractal compression prob (~ fraction • A: of ink) $d \quad e \quad f$ b w10.5 0 w2 0.5 0 0 0.5 1 0 1/3 w30.5 0 0 0.5 0.5 0.5 1/3 15-826 Copyright: C. Faloutsos (2007) 26



CMU SCS

Fractal compression

- The above transformations 'describe' the Sierpinski triangle is it the only one?
- ie., how to de-compress?

15-826



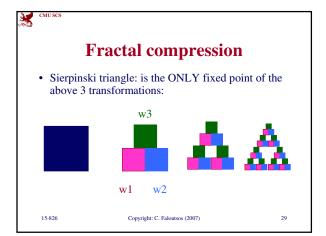
Fractal compression

- The above transformations 'describe' the Sierpinski triangle is it the only one?
- A: YES!!!
- ie., how to de-compress?
- A1: Iterated functions (expensive)
- A2: Randomized (surprisingly, it works!)

15-826

Copyright: C. Faloutsos (2007)

28





CMU SCS

Fractal compression

- We'll get the Sierpinski triangle, NO MATTER what image we start from! (as long as it has at least one black pixel!)
- thus, (one, slow) decompression algorithm:
 - start from a random image
 - apply the given transformations
 - union them and
 - repeat recursively
- · drawback?

15-826

Copyright: C. Faloutsos (2007)

30



CMU SC

Fractal compression

- A: Exponential explosion: with 3 transformations, we need 3^{**k} sub-images, after k steps
- Q: what to do?

15-826

Copyright: C. Faloutsos (2007)



MU SCS

Fractal compression

- A: PROBABILISTIC algorithm:
 - pick a random point (x0, y0)
 - choose one of the 3 transformations with prob. p1/p2/p3
 - generate point (x1, y1)
 - repeat
 - [ignore the first 30-50 points why??]
- Q: why on earth does this work?
- A: the point (x_n, y_n) gets closer and closer to Sierpinski points (n=1, 2, ...), ie:

15-826

Copyright: C. Faloutsos (2007)

32

31



CMU SCS

Fractal compression

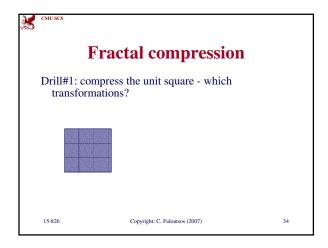
- ... points outside the Sierpinski triangle have no chance of attracting our 'random' point (x_n, y_n)
- Q: how to compress a real (b/w) image?
- A: 'Collage' theorem (informally: find portions of the image that are miniature versions, and that cover it completely)

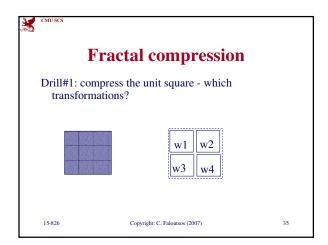
Drills:

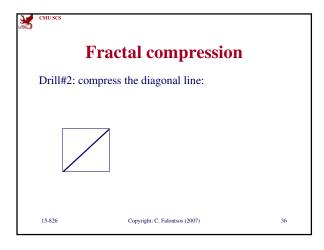
15-826

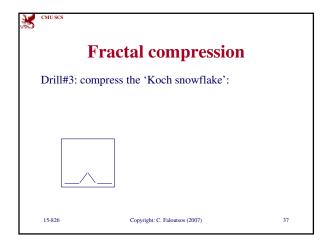
Copyright: C. Faloutsos (2007)

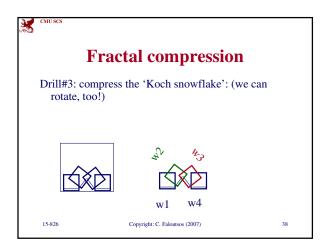
33

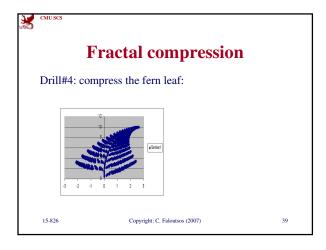


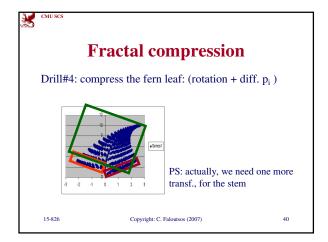














Fractal compression

- How to find self-similar pieces automatically?
- A: [Peitgen+]: eg., quad-tree-like decomposition

15-826

Copyright: C. Faloutsos (2007)

41

42

3 CMUSCS

Fractal compression

- Observations
 - may be lossy (although we can store deltas)
 - can be used for color images, too
 - can 'focus' or 'enlarge' a given region, without JPEG's 'blockiness'

15-826

C. Faloutsos 15-826

43



Conclusions

- JPEG: DCT for images
- MPEG: I-frames; interpolation, for video
- IFS: surprising compression method

15-826

Copyright: C. Faloutsos (2007)



Resources/ References

- IFS code: www.cs.cmu.edu/~christos/SRC/ifs.tar
- Gregory K. Wallace, The JPEG Still Picture Compression Standard, CACM, 34, 4, April 1991, pp. 31-44

15-826

Copyright: C. Faloutsos (2007)



References

- D. Le Gall, MPEG: a Video Compression Standard for Multimedia Applications CACM, 34, 4, April 1991, pp. 46-58
- M.F. Barnsley and A.D. Sloan, A Better Way to Compress Images, BYTE, Jan. 1988, pp. 215-223
- Heinz-Otto Peitgen, Hartmut Juergens, Dietmar Saupe: Chaos and Fractals: New Frontiers of Science, Springer-Verlag, 1992

15-826

Copyright: C. Faloutsos (2007)

45