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15-826: Multimedia Databases and Data Mining

Lecture #30: Conclusions
C. Faloutsos

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Outline

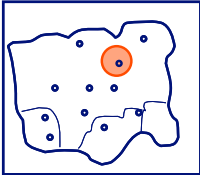
Goal: 'Find similar / interesting things'

- Intro to DB
- Indexing - similarity search
 - Points
 - Text
 - Time sequences; images etc
 - Graphs
- Data Mining

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Indexing - similarity search



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Indexing - similarity search

- R-trees
- z-ordering / hilbert curves
- M-trees
- (DON'T FORGET ...)

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Indexing - similarity search

- R-trees
- z-ordering / hilbert curves
- M-trees
- beware of high intrinsic dimensionality

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Text searching

- ‘find all documents with word *bla*’

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Text searching

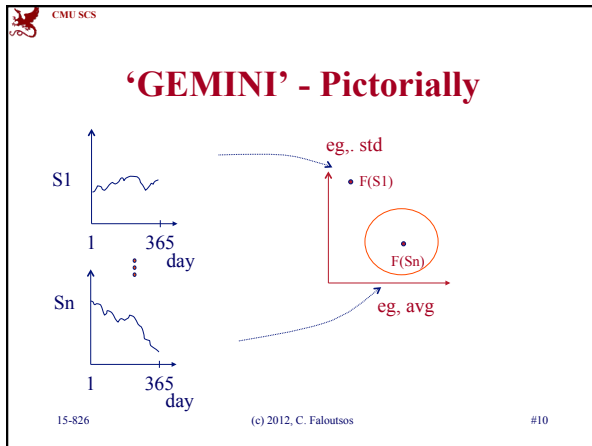
- Full text scanning (‘grep’)
- Inversion (B-tree or hash index)
- (signature files)
- Vector space model
 - Ranked output
 - Relevance feedback
- String editing distance (-> dynamic prog.)

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Multimedia indexing

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- Multimedia indexing**
- Feature extraction for indexing (GEMINI)
 - Lower-bounding lemma, to guarantee no false alarms
 - MDS/FastMap
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Time series & forecasting

Goal: given a signal (eg., sales over time and/or space)
 Find: patterns and/or compress

lynx caught per year

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Wavelets

- Q: baritone/silence/soprano - DWT?

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Time series + forecasting

- Fourier; Wavelets
- Box/Jenkins and AutoRegression
- non-linear/chaotic forecasting (fractals again)
 - 'Delayed Coordinate Embedding' ~ nearest neighbors

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Graphs

- Real graphs: surprising patterns
 - ??

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Graphs

- Real graphs: surprising patterns
 - 'six degrees'
 - **Skewed** degree distribution ('rich get richer')
 - Super-linearities (2x nodes -> 3x edges)
 - Diameter: **shrinks** (!)
 - Might have **no** good cuts

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Data Mining - DB

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Data Mining - DB

- Association Rules ('diapers' -> 'beer')
- [OLAP (DataCubes, roll-up, drill-down)]
- [Classifiers]

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Taking a step back:


We saw some fundamental, recurring concepts and tools:

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Powerful, recurring tools

- Fractals/ self similarity
 - Zipf, Korcak, Pareto’s laws
 - intrinsic dimension (Sierpinski triangle)
 - correlation integral
 - Barnsley’s IFS compression
 - (Kronecker graphs)




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Powerful, recurring tools

- Fractals/ self similarity
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• ‘Take logarithms’
• AVOID ‘avg’

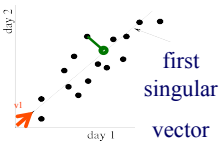


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Powerful, recurring tools

- SVD (optimal L2 approx)
 - LSI, KL, PCA, ‘eigenSpokes’, (& in ICA)
 - HITS (PageRank)



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Powerful, recurring tools

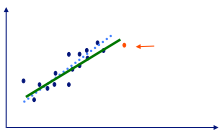
- Discrete Fourier Transform
- Wavelets

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Powerful, recurring tools

- Matrix inversion lemma
 - Recursive Least Squares
 - Sherman-Morrison(-Woodbury)




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Summary

- **fractals / power laws** probably lead to the most startling discoveries ('the mean may be meaningless')
- **SVD**: behind PageRank/HITS/tensors/...
- **Wavelets**: Nature seems to prefer them
- **RLS**: seems to achieve the impossible
- (approximate counting – NOT in the exam)

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Thank you!

- Feel free to contact me:
 - [christos@cs](mailto:christos@cs.cmu.edu) GHC 8019
- Reminder: faculty course eval's:
 - www.cmu.edu/hub/fce/
- Final: Thu. Dec 13, 5:30-8:30pm @DH A302
 - Double-check at www.cmu.edu/hub/docs/final-exams.pdf
- Have a great break!

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