

Why data-driven?

Hongxin Zhang
zhx@cad.zju.edu.cn

State Key Lab of CAD&CG, ZJU
2015-03-10





Outline

- Background
- What is data-driven about?
- Is it really useful for computer science and technology?

The largest challenge of Today's CS



- Big Data
- Big companies are collecting data!!!
 - Google, Apple, Facebook, IBM, Microsoft, Amazon, ...
 - In china, Baidu, Alibaba, Tecent, Sina

The largest challenge of Today's CS



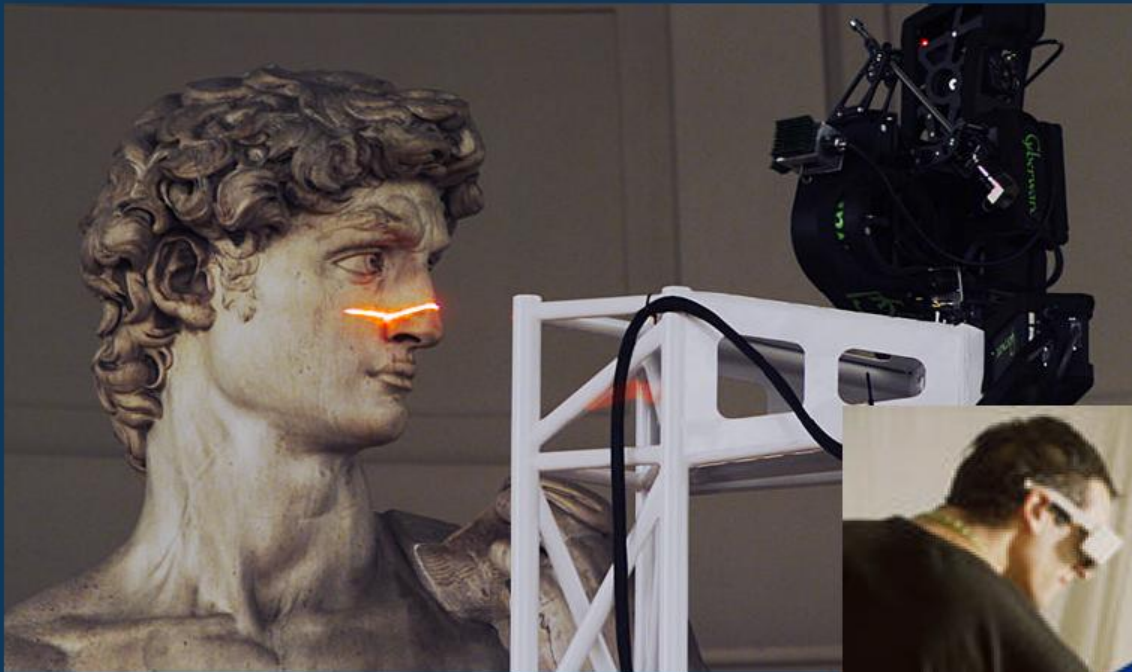
- Data, Data, Data ...
 - The tedious effort required to create digital worlds and digital life.
 - Finding new ways to communicate and new kinds of media to create.
 - Experts are expensive: scientists, engineers, filmmakers, graphic designers, fine artists, and game designers.
- Process existing data and then create new ones from them.

Computers are really fast

- If you can create it, you can render it



How do you create it?



Digital Michelangelo Project



Steven Schkolne

Pure procedural synthesis vs. Pure data



- Creating motions for a character in a movie
 - Pure procedural synthesis.
 - compact, but very artificial, rarely used in practice.
 - “By hand” or “pure data”.
 - higher quality but lower flexibility.
- the best of both worlds: hybrid methods?!?

Everything but Avatar



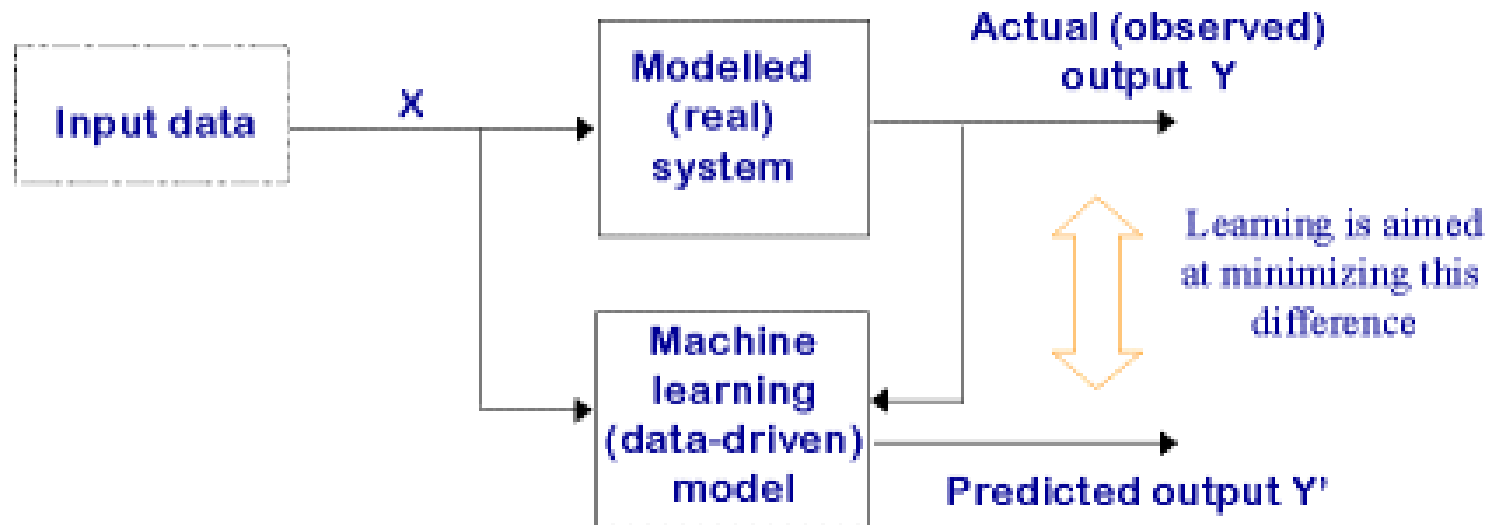


Bayesian Reasoning

- ❖ Principle modeling of uncertainty.
- ❖ General purpose models for unstructured data.
- ❖ Effective algorithm for data fitting and analysis under uncertainty.
- But currently it is always used as a black box.

Belief v.s. Probability

Data driven modeling



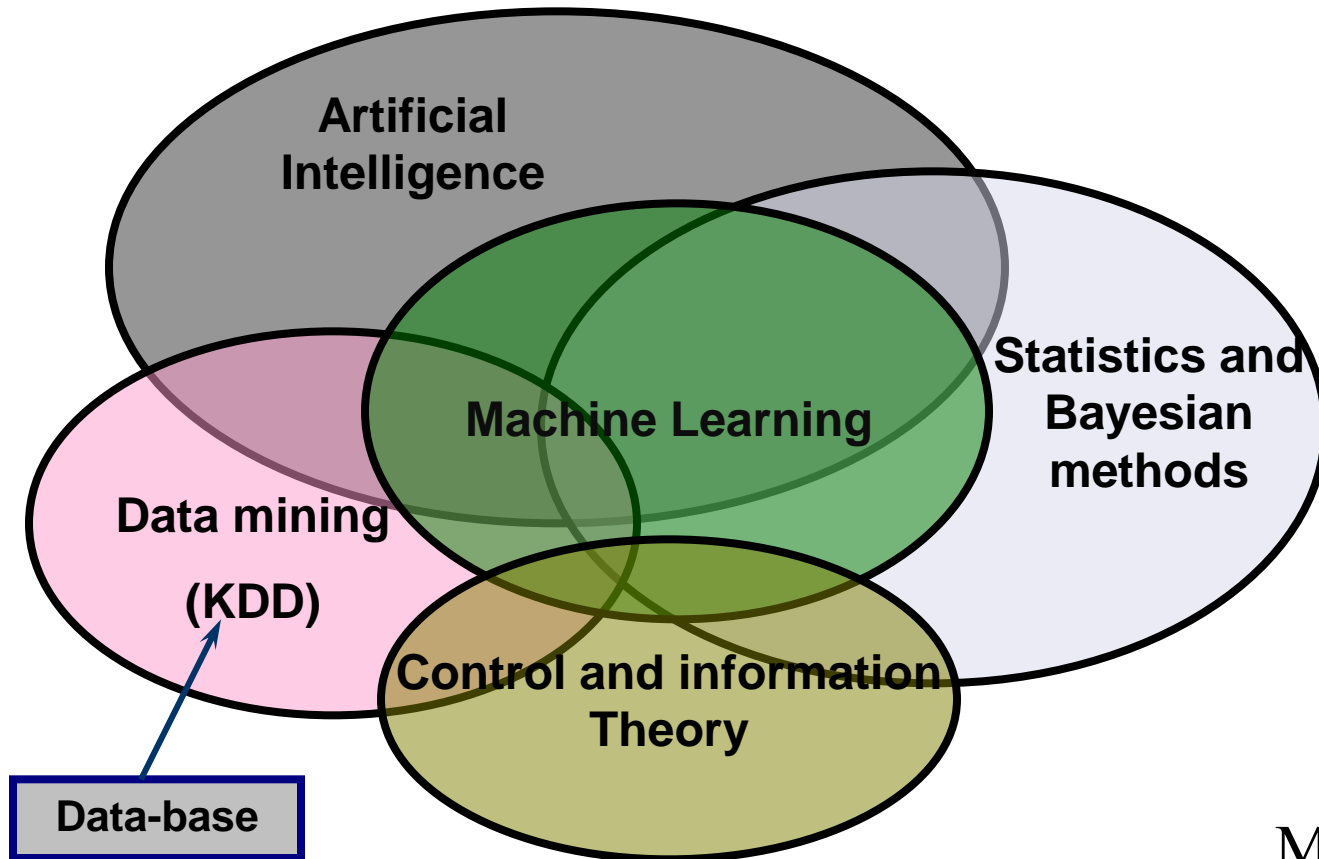


Data-driven vocabulary

- Data
 - data-driven, data mining
- Learning
 - machine learning, statistical learning
- Uncertainty
 - probability, likelihood
- Intelligent
 - Inference, decision, detection, recognition



Data-driven related techniques



Computer Vision	Multi-media	Bio-informatics	Computer Graphics	Information retrieval
-----------------	-------------	-----------------	-------------------	-----------------------



Data-driven system

- Learning systems are not directly programmed to solve a problem, instead develop own program based on:
 - examples of how they should behave
 - from trial-and-error experience trying to solve the problem

Different from standard CS: want to implement unknown function, only have access to sample input-output pairs (training examples)

Main categories of learning problems



Learning scenarios differ according to the available information in training examples

- **Supervised**: correct output available
 - **Classification**: 1-of-N output (speech recognition, object recognition, medical diagnosis)
 - **Regression**: real-valued output (predicting market prices, temperature)
- **Unsupervised**: no feedback, need to construct measure of good output
 - **Clustering** : Clustering refers to techniques to segmenting data into coherent “clusters.”
 - **Novelty-detection**: detecting new data points that deviate from the normal.
- **Reinforcement**: scalar feedback, possibly temporally delayed



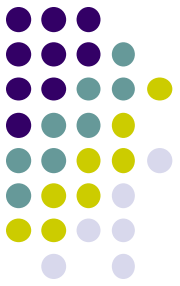
Main class of learning problems

Learning scenarios differ according to the available information in training examples

- **Supervised**: correct output available
 - ...
- **Semi-Supervised**: only a part of output available
 - **Ranking**:
- **Unsupervised**: no feedback, need to construct measure of good output
 - ...
- *Reinforcement*: scalar feedback, possibly temporally delayed

And more ...

- Time series analysis.
- Dimension reduction.
- Model selection.
- Generic methods.
- Graphical models.



Why data driven methods?



- **Develop enhanced computer systems**
 - automatically adapt to user, customize
 - often difficult to acquire necessary knowledge
 - discover patterns offline in large databases (*data mining*)
- **Improve understanding of human, biological learning**
 - computational analysis provides concrete theory, predictions
 - explosion of methods to analyze brain activity during learning
- **Timing is good**
 - growing amounts of data available
 - cheap and powerful computers
 - suite of algorithms, theory already developed

Is it really useful for computer science and technology?



- Con: Everything is machine learning or everything is human tuning?
 - Sometimes, this may be true.
- Pro: more understanding of learning, but yields much more powerful and effective algorithms.
 - Problem taxonomy.
 - General-purpose models.
 - Reasoning with probabilities.
- ❖ I believe the mathematic magic.

What will be a successful D-D algorithm?



- Computational efficiency
- Robustness
- Statistical stability

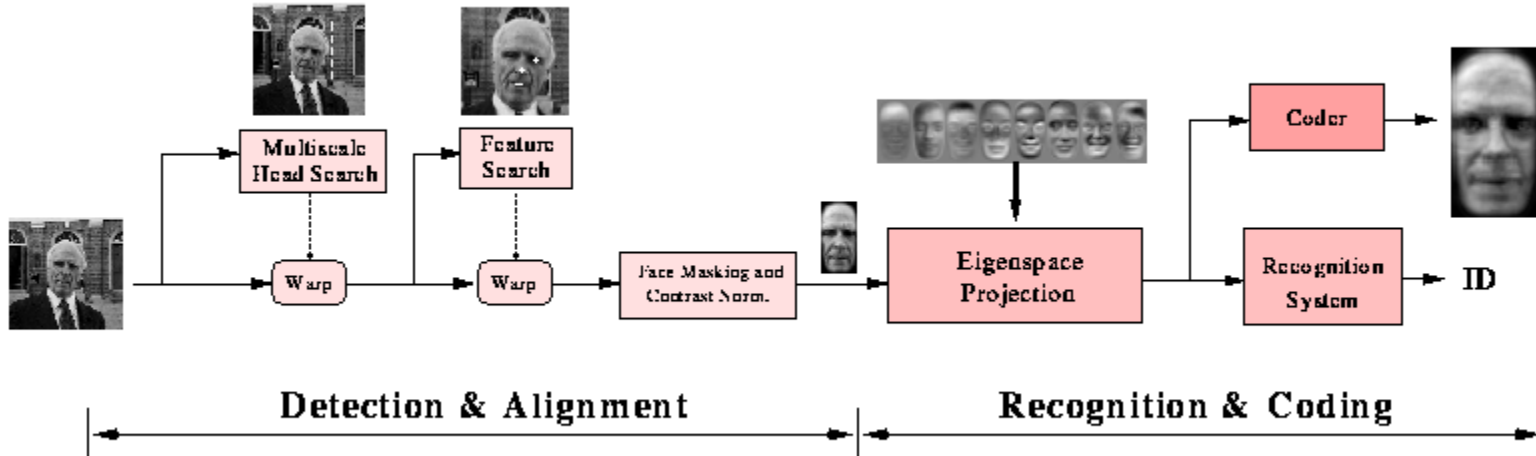


The First Example: Google!



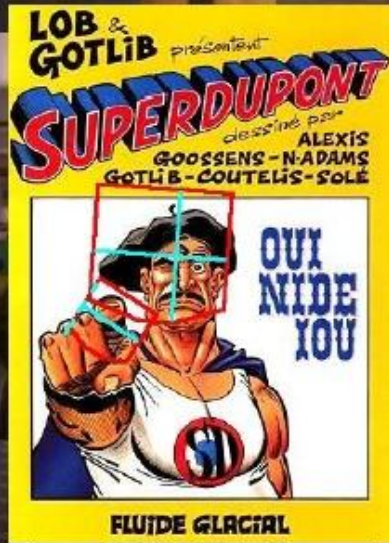
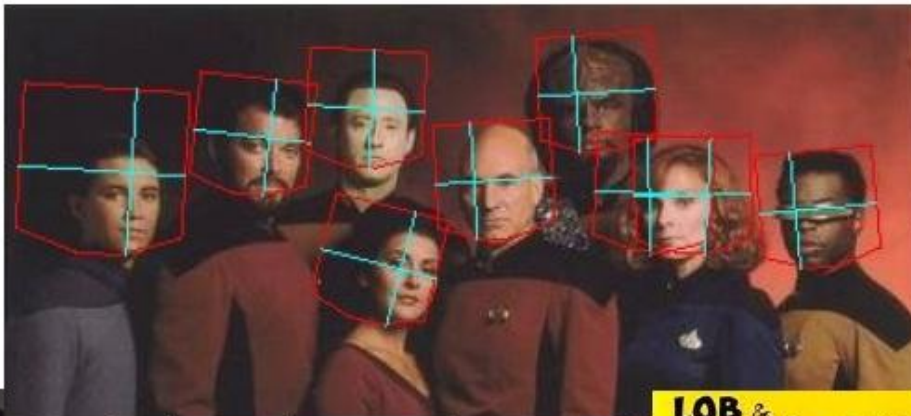
- 每天过滤200亿个网页
- 每天追踪300亿个的独立URL
- 每月接受1000亿次搜索请求

Object detection and recognition - the power of DD



The image is copied from
<http://vismod.media.mit.edu/vismod/demos/facerec/>

Object detection and recognition



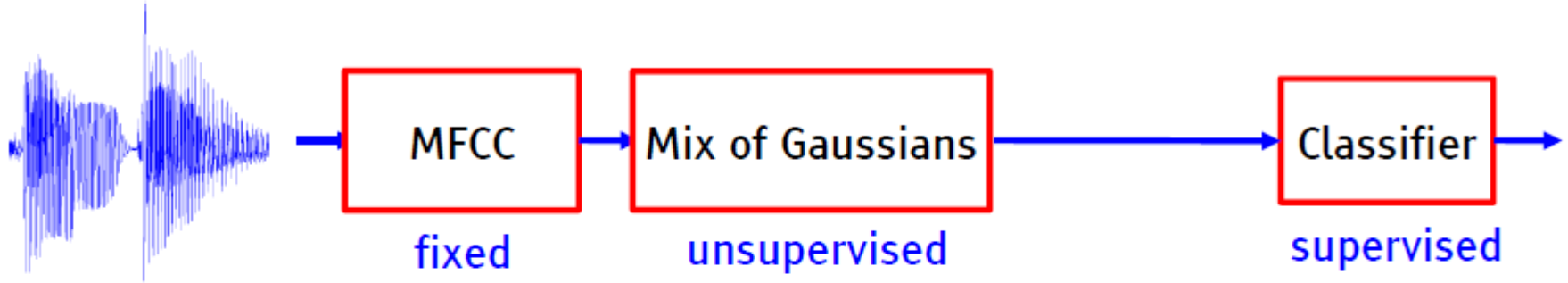
Face [Vaillant et al IEE 1994] [Garcia et al PAMI 2005] [Osadchy et al JMLR 2007]
Pedestrian: [Kavukcuoglu et al. NIPS 2010] [Sermanet et al. CVPR 2013]



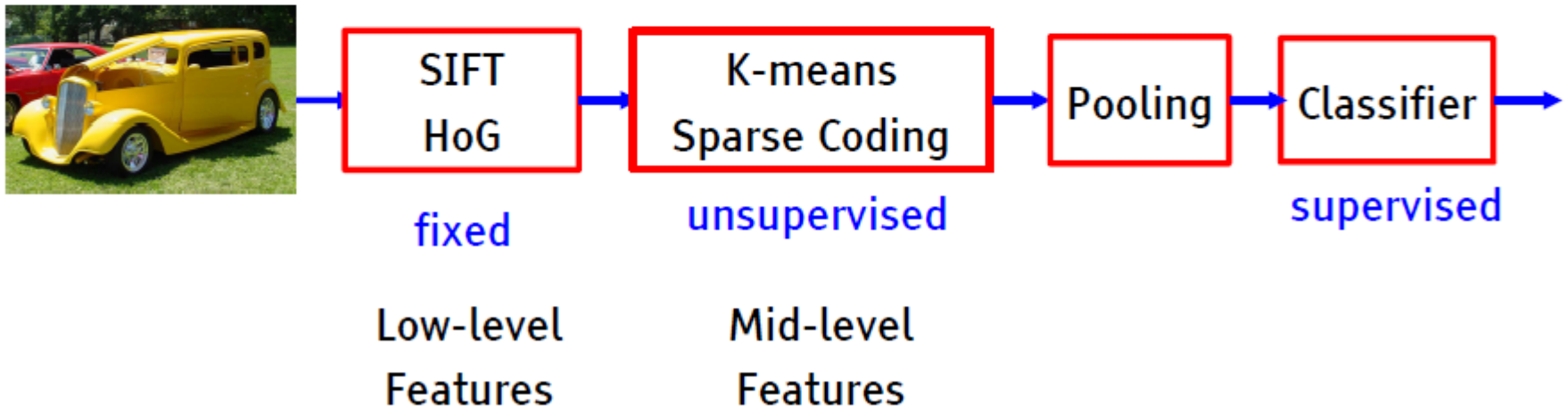
Speech recognition

Modern architecture for pattern recognition

▶ Speech recognition: early 90's – 2011



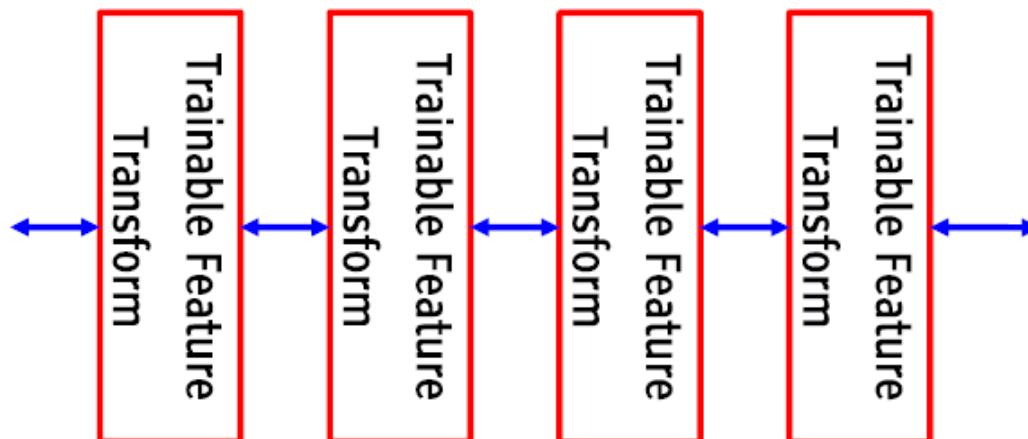
▶ Object Recognition: 2006 - 2012





Speech recognition

- Hierarchy of representations with increasing level of abstraction
- Each stage is a kind of trainable feature transform
- Image recognition
 - ▶ Pixel → edge → texon → motif → part → object
- Text
 - ▶ Character → word → word group → clause → sentence → story
- Speech
 - ▶ Sample → spectral band → sound → ... → phone → phoneme → word →



Document processing – Bayesian classification



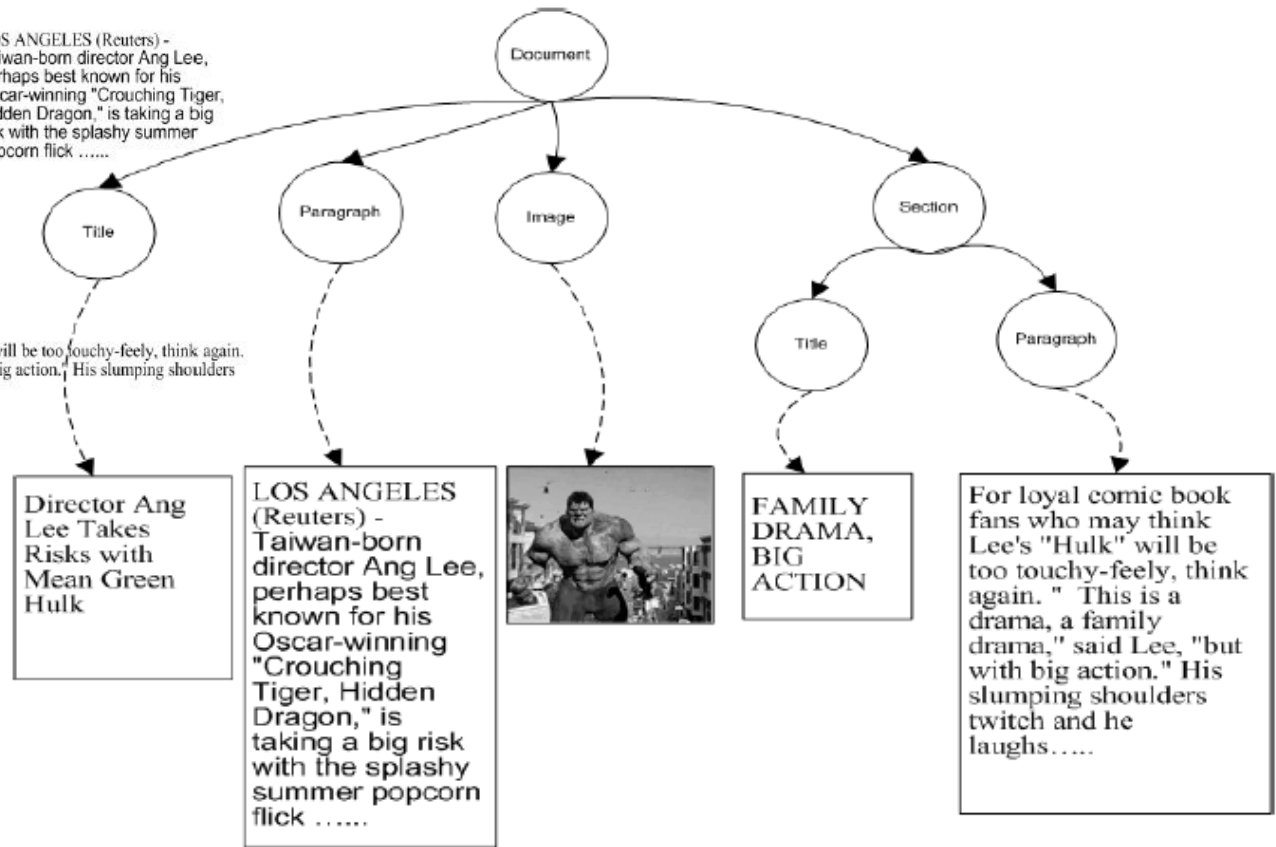
Director Ang Lee Takes Risks with Mean Green 'Hulk'



LOS ANGELES (Reuters) - Taiwan-born director Ang Lee, perhaps best known for his Oscar-winning "Crouching Tiger, Hidden Dragon," is taking a big risk with the splashy summer popcorn flick

FAMILY DRAMA, BIG ACTION

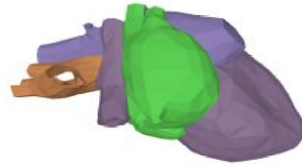
For loyal comic book fans who may think Lee's "Hulk" will be too touchy-feely, think again. " This is a drama, a family drama," said Lee, "but with big action." His slumping shoulders twitch and he laughs.....



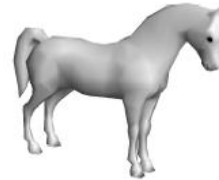
Mesh Processing – Data clustering/segmentation



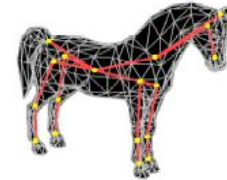
(c) mechanical part – 1270 faces
7 patches



(d) heart – 1619 faces
4 patches



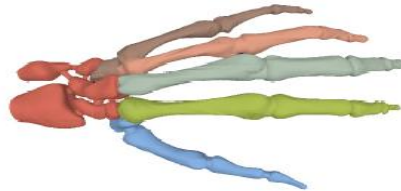
(a) object



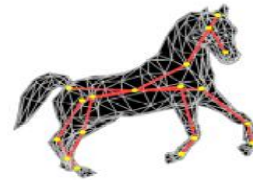
(b) skeleton



(e) Venus – 67,170 faces
3 patches



(f) skeleton hand – 654,666 faces
6 patches



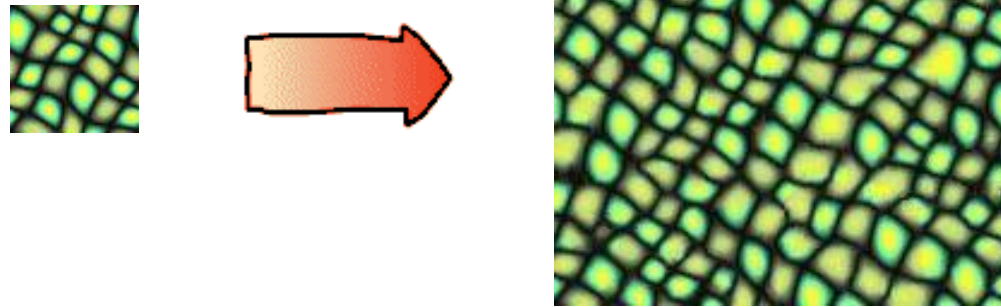
(c) deformed skeleton



(d) deformed object

- *Hierarchical Mesh Decomposition using Fuzzy Clustering and Cuts.*
By Sagi Katz and Ayellet Tal, SIGGRAPH 2003

Texture synthesis and analysis – Hidden Markov Model



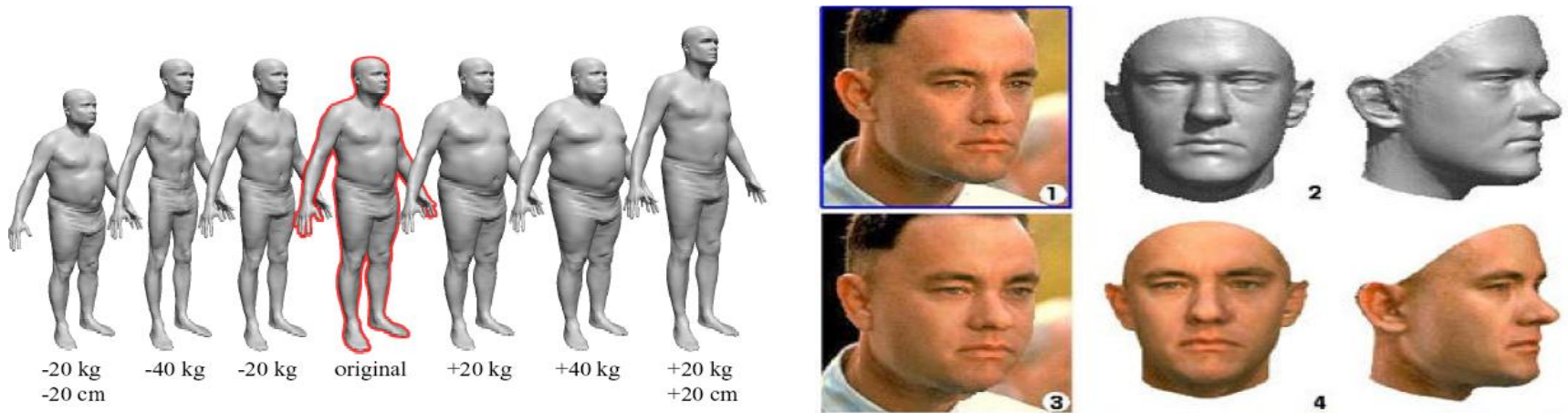
- *Texture Synthesis over Arbitrary Manifold Surfaces.* Li-Yi Wei and Marc Levoy. SIGGRAPH 2001.
- *Fast Texture Synthesis using Tree-structured Vector Quantization.* Li-Yi Wei and Marc Levoy. SIGGRAPH 2000.

Reflectance texture synthesis – Dimension reduction



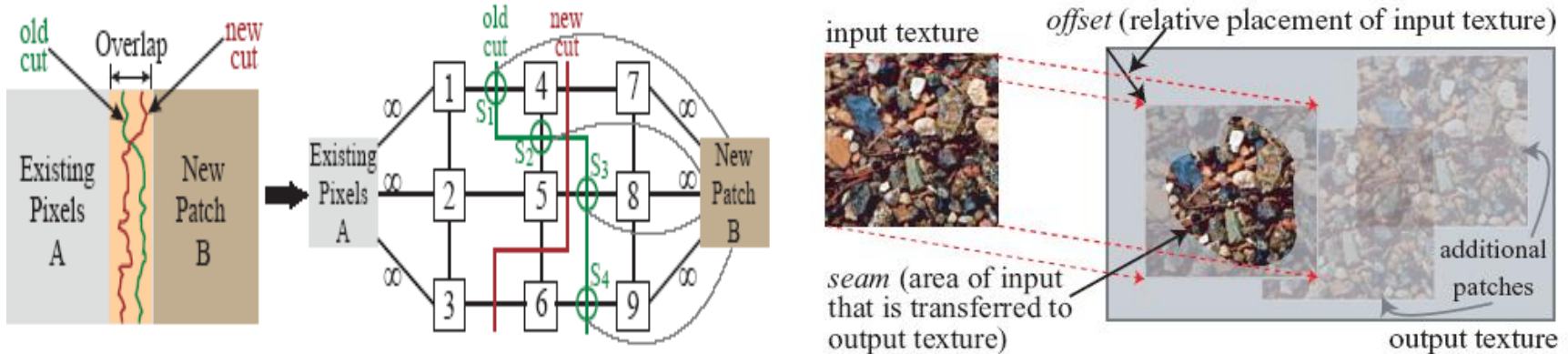
- *Synthesizing Bidirectional Texture Functions for Real-World Surfaces.* Xinguo Liu, Yizhou Yu and Heung-Yeung Shum. SIGGRAPH 2001.
- More recent papers...

Human shapes - Dimension reduction



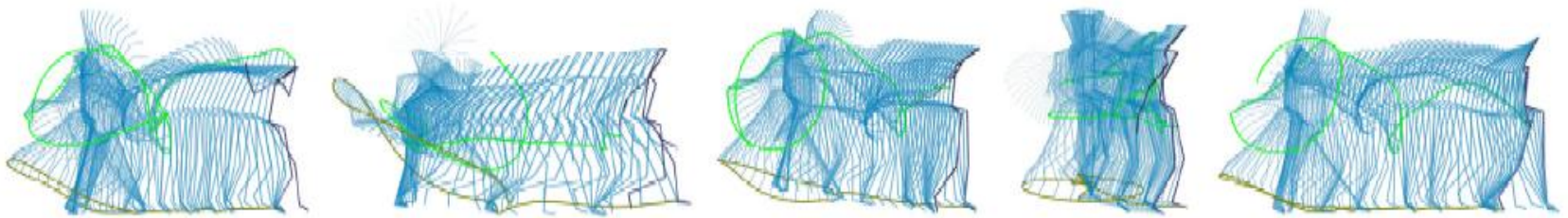
- *The Space of Human Body Shapes: Reconstruction and Parameterization From Range Scans.* Brett Allen, Brian Curless, Zoran Popovic. SIGGRAPH 2003.
- *A Morphable Model for the Synthesis of 3D Faces.* Volker Blanz and Thomas Vetter. SIGGRAPH 1999.

Image processing and synthesis - Graphical model



- *Image Quilting for Texture Synthesis and Transfer.* Alexei A. Efros and William T. Freeman. SIGGRAPH 2001.
- *Graphcut Textures: Image and Video Synthesis Using Graph Cuts.* V Kwatra, I. Essa, A. Schödl, G. Turk, and A. Bobick. SIGGRAPH 2003.

Human Motion - Time series analysis



A pirouette and promenade in five synthetic styles drawn from a space that contains ballet, modern dance, and different body types. The choreography is also synthetic. Streamers show the trajectory of the left hand and foot.

- *Style Machines*. M. Brand and A. Hertzmann. SIGGRAPH 2000.
- *A Data-Driven Approach to Quantifying Natural Human Motion*. L. Ren, A. Patrick, A. Efros, J. Hodgins, J. Rehg. SIGGRAPH 2005

Video Textures - Reinforcement Learning



- [Video textures](#). Arno Schödl, Richard Szeliski, David H. Salesin, and Irfan Essa. *SIGGRAPH 2000*.



Summary

- Learning (from Data) is a nut-shell, :-D
 - Keywords
 - Noun: data, models, patterns, features;
 - Adj.: probabilistic, statistical;
 - Verb: fitting, reasoning, mining.



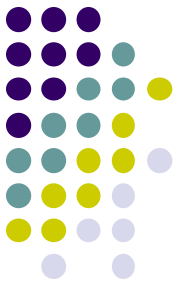
Homework

- Try to find potential learning based (data driven) applications in your research area



Reference

- Reinforcement learning: A survey



The End

新浪微博： @浙大张宏鑫

微信公众号：

