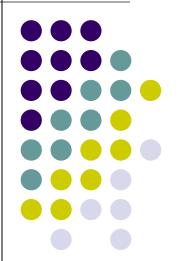
### Why data-driven?

Hongxin Zhang zhx@cad.zju.edu.cn

State Key Lab of CAD&CG, ZJU 2014-02-27



#### **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?



## Pure procedural synthesis vs. Pure data



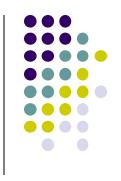
- Creating motions for a character in a movie
  - Pure procedural synthesis.
    - compact, but very artificial, rarely used in practice.
  - "By hand" or "pure data".
    - higher quality but lower flexibility.
  - the best of both worlds: hybrid methods?!?

### **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



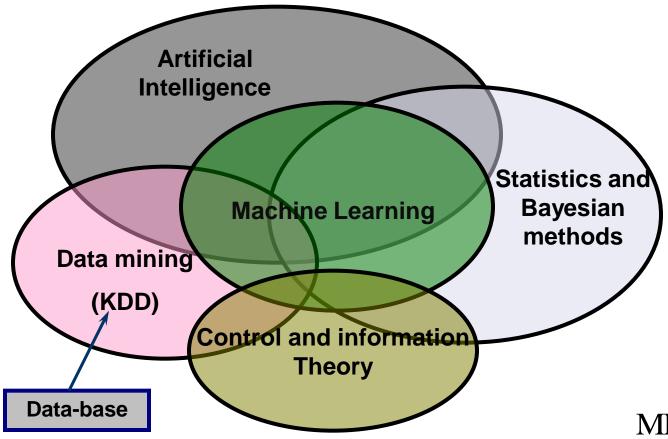
#### **Data-driven vocabulary**



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

### Data-driven related techniques





 $ML \neq AI$ 

**Computer** Vision

Multi-media

**Bio-informatics** 

**Computer Graphics** 

Information retrieval

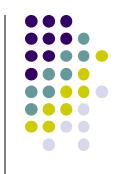
#### **Data-driven system**



- 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





Learning scenarios differ according to the available information in training examples

- Supervised: correct output available
  - ...
- Semi-Supervised: only a part of output available
  - Ranking:

#### 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!





<u>广告计划</u> - <u>Google 大全</u> - **中国 Google** 

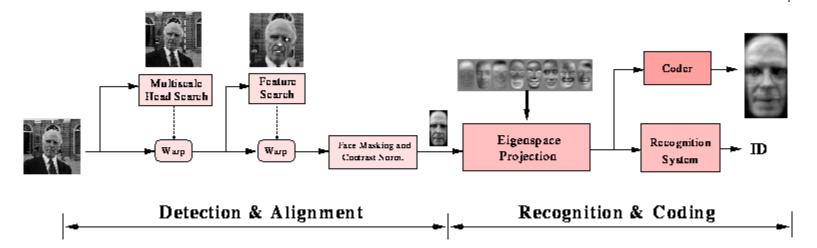
将 Google 设为首页!

©2007 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

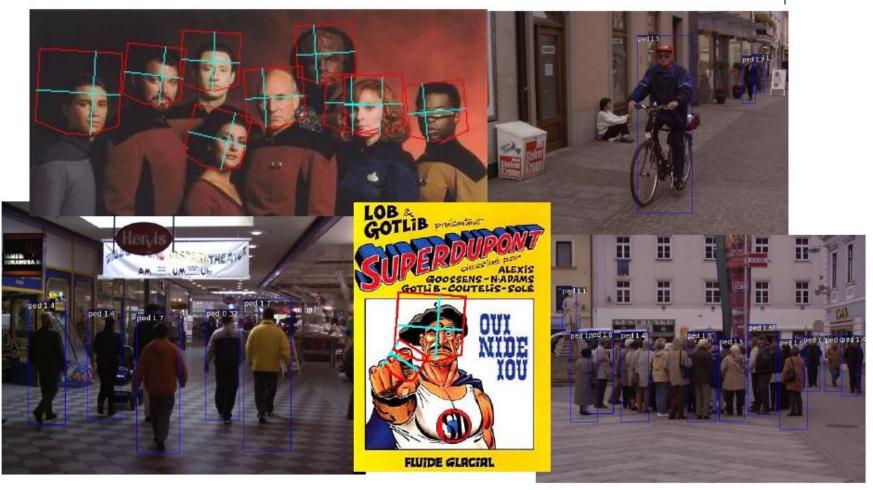




The image is copied from http://vismod.media.mit.edu/vismod/demos/facerec/

# Object detection and recognition





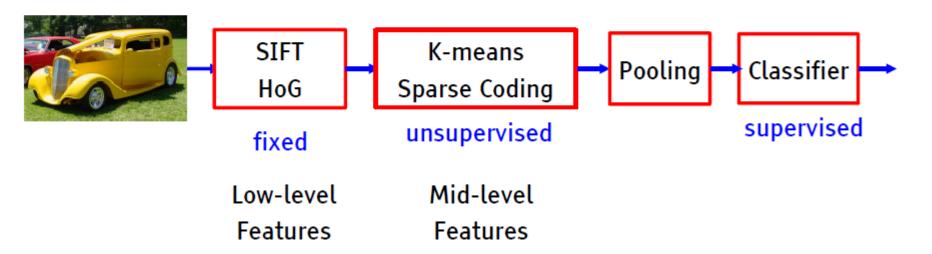
Face [Vaillant et al IEE 1994] [Garcia et al PAMI 2005] [Osadchy et al JMLR 2007] Pedestrian: [Kavukcuoglu et al. NIPS 2010] [Sermanet et al. CVPR 2013]



- Modern architecture for pattern recognition
  - Speech recognition: early 90's 2011

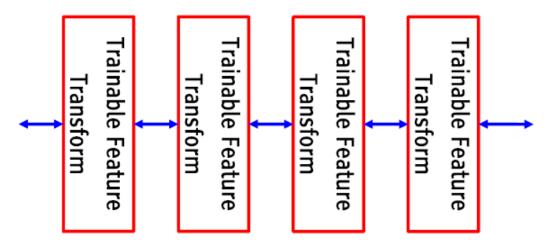


Object Recognition: 2006 - 2012



### Speech recognition

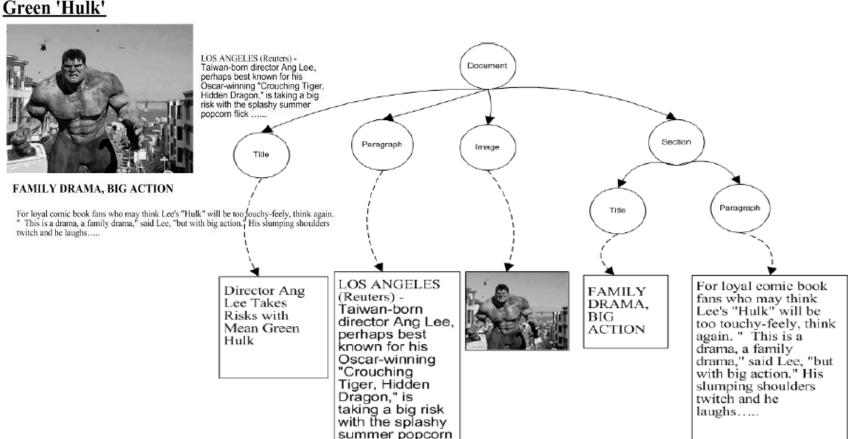
- Hierarchy of representations with increasing level of abstraction
- Each stage is a kind of trainable feature transform
- Image recognition
  - Pixel → edge → texton → motif → part → object
- Text
  - Character → word → word group → clause → sentence → story
- Speech
  - ▶ Sample  $\rightarrow$  spectral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



# Document processing – Bayesian classification



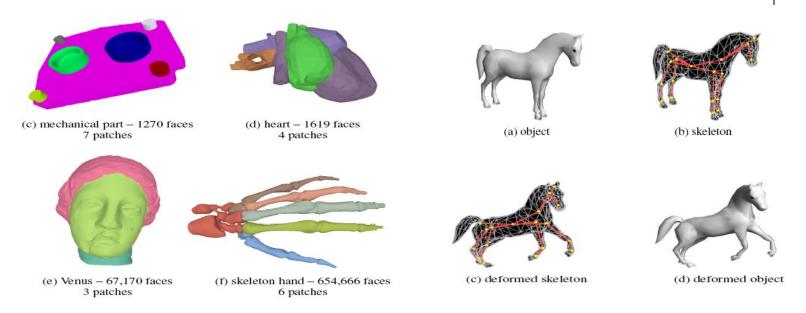
#### <u>Director Ang Lee Takes Risks with Mean</u> <u>Green 'Hulk'</u>



flick ......

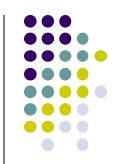
## Mesh Processing – Data clustering/segmentation

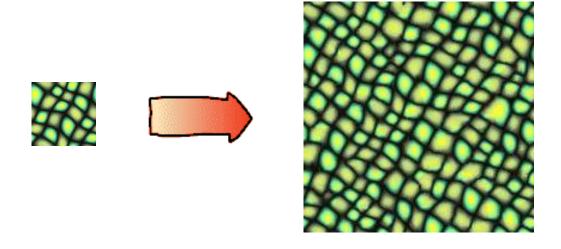




Hierarchical Mesh Decomposition using Fuzzy Clustering and Cuts.
 By Sagi Katz and Ayellet Tal, SIGGRAPH 2003

### 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

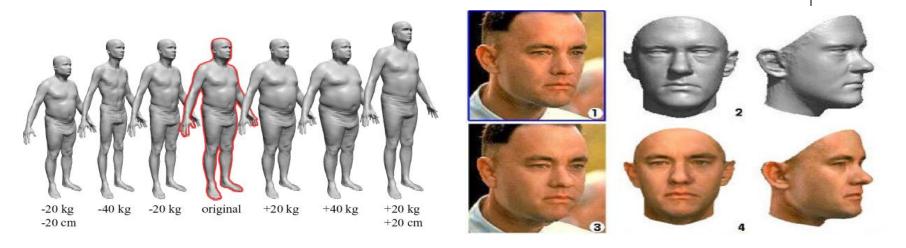




- Synthesizing Bidirectional Texture Functions for Real-World Surfaces. Xinguo Liu, Yizhou Yu and Heung-Yeung Shum. SIGGRAPH 2001.
- More recent papers...

### Human shapes - Dimension reduction

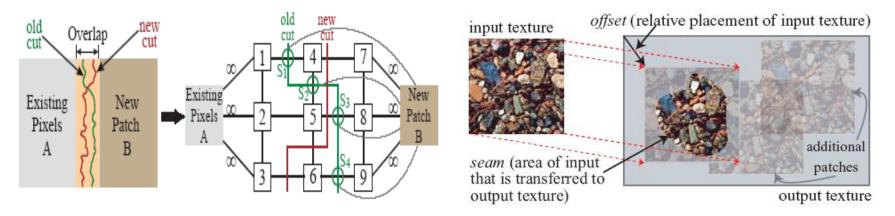




- The Space of Human Body Shapes: Reconstruction and Parameterization From Range Scans. Brett Allen, Brian Curless, Zoran Popovic. SIGGRAPH 2003.
- A Morphable Model for the Synthesis of 3D Faces. Volker Blanz and Thomas Vetter. SIGGRAPH 1999.

## Image processing and synthesis - Graphical model

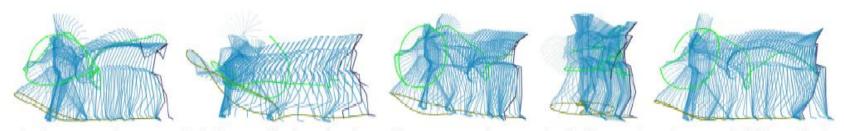




- Image Quilting for Texture Synthesis and Transfer. Alexei A. Efros and William T. Freeman. SIGGRAPH 2001.
- Graphcut Textures: Image and Video Synthesis Using Graph Cuts.
  V Kwatra, I. Essa, A. Schödl, G. Turk, and A. Bobick. SIGGRAPH 2003.

## **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







 <u>Video textures</u>. 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.

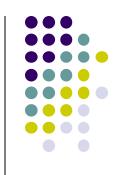
#### Homework



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



#### Reference



Reinforcement learning: A survey