# Why data-driven?

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# 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?



**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



# What is machine learning? (Cont.)



- Definition by Mitchell, 1997
  - A program learns from experience E with respect to some class of tasks T and performance measure P, if its performance at task T, as measured by P, improves with experience E.
  - 机器学习乃于某类任务兼性能度量的经验中学习之程序;若其作用于任务,可由度量知其于已知经验中获益。
- Comments from Hertzmann, 2003
  - For the purposes of computer graphics, machine learning should really be viewed as a set of techniques for leveraging data. Given some data, we can model the process that generated the data.

## **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

# **Growth of Machine Learning**

- Machine learning is preferred approach to
  - Speech recognition, Natural language processing
  - Computer vision
  - Medical outcomes analysis
  - Robot control
  - ...

#### This trend is accelerating

- Improved machine learning algorithms
- Improved data capture, networking, faster computers
- Software too complex to write by hand
- New sensors I / O devices
- Demand for self-customization to user, environment



# 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</u> 大全 - **中国 Google** 

<u>将 Google 设为首页!</u>

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- •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]



#### **Object Detection**

#### (Prof. H. Schneiderman)



Example training images for each orientation



# **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  $\rightarrow$  edge  $\rightarrow$  texton  $\rightarrow$  motif  $\rightarrow$  part  $\rightarrow$  object
- Text
  - ▶ Character  $\rightarrow$  word  $\rightarrow$  word group  $\rightarrow$  clause  $\rightarrow$  sentence  $\rightarrow$  story

#### Speech

▶ Sample → spectral band → sound → ... → phone → phoneme → word →



# Document processing – Bayesian classification

Director Ang Lee Takes Risks with Mean Green 'Hulk'



### 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.

# Learning a Probabilistic Latent Space of Object Shapes - GANs





# Learning a Probabilistic Latent Space of Object Shapes - GANs





We combine the encoder with 3D-GAN for reconstruction and generation.

### 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.

#### Motion texture -Linear dynamic system





 Motion Texture: A Two-Level Statistical Model for Character Motion Synthesis. Yan Li, Tianshu Wang, and Heung-Yeung Shum. SIGGRAPH 2002.

# 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





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