

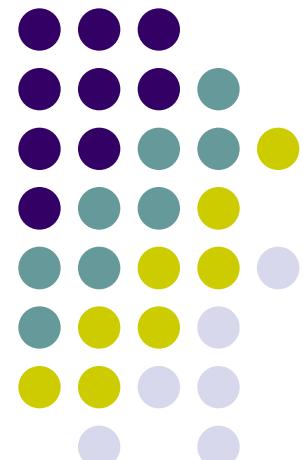
# Why data-driven?

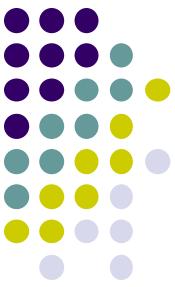
Hongxin Zhang

[zhx@cad.zju.edu.cn](mailto:zhx@cad.zju.edu.cn)

State Key Lab of CAD&CG, ZJU

2022-02-22

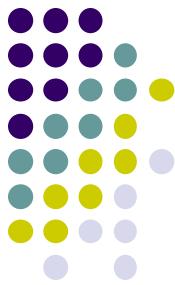




# Outline

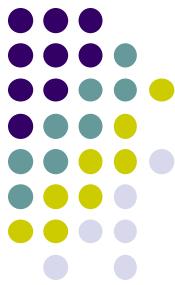
- Background
- What is data-driven about?
- Is it really useful for computer science and technology?

# The largest challenge of Today's CS



- Big Data
- All big companies are collecting data!!!
  - Google, Apple, Facebook, IBM, Microsoft, Amazon, ...
- In China, Baidu, Alibaba, Tencent, 360, DiDi, Netease, Xiaomi, Sina, Huawei

# The largest challenge of Today's CS



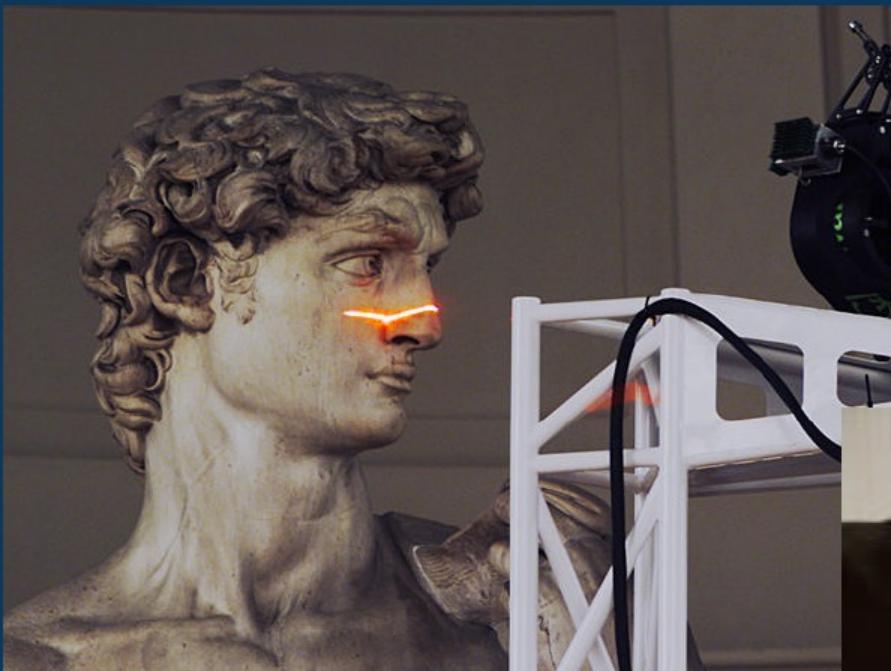
- Data, Data, Data ... (in computer graphics)
  - The tedious effort required to create digital worlds and digital life.
    - Finding new ways to communicate and new kinds of media to create.
    - Experts are expensive: scientists, engineers, filmmakers, graphic designers, fine artists, and game designers.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

# Computers are really fast

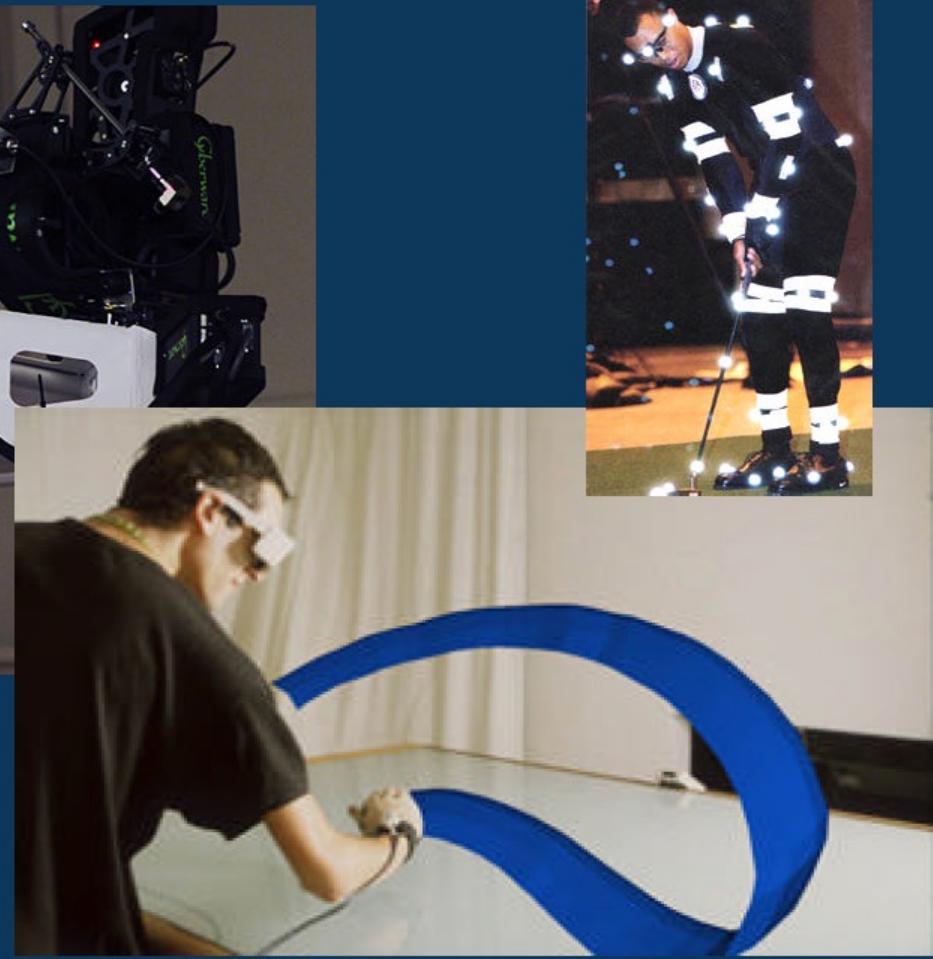
- If you can create it, you can render it



# How do you create it?



Digital Michaelangelo Project

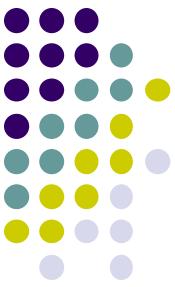


Steven Schkolne

# Pure procedural synthesis vs. Pure data



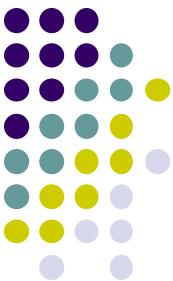
- Creating motions for a character in a movie
  - Pure procedural synthesis (**model**)
    - compact, but very artificial, rarely used in practice.
  - “By hand” or “pure data” (**data**)
    - higher quality but lower flexibility.
  - the best of both worlds: hybrid methods?!?



# Everything but Avatar



Make it easy and true



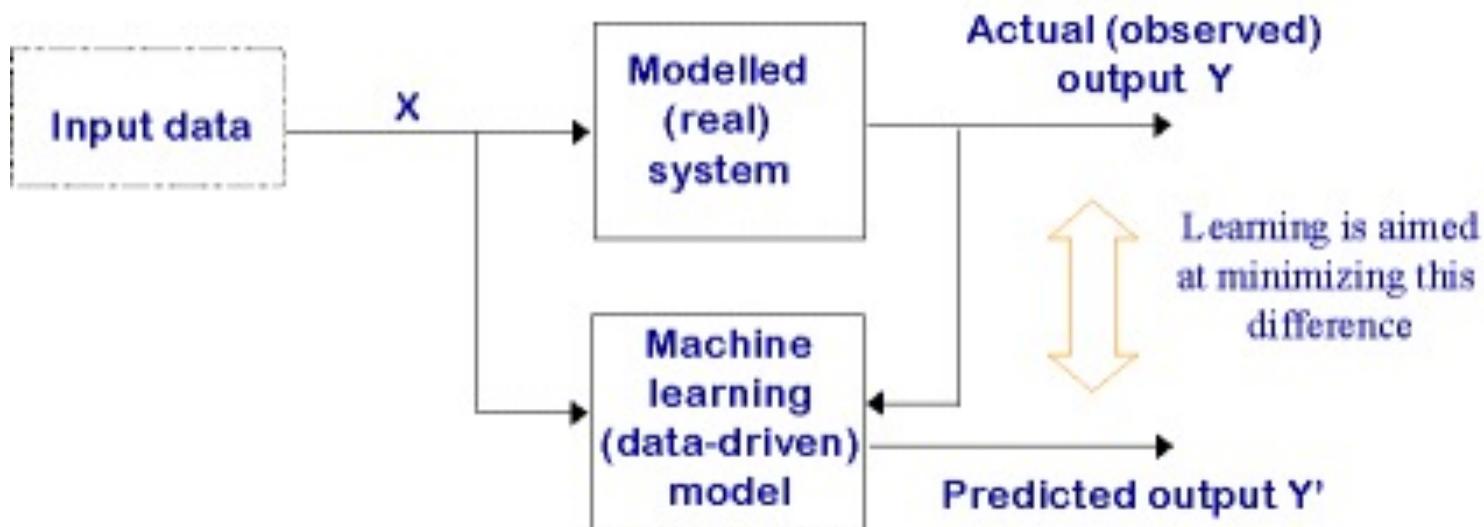
# Bayesian Reasoning

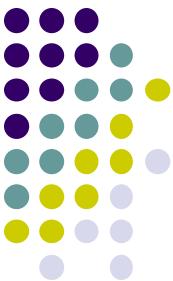
- ❖ Principle modeling of uncertainty.
  - ❖ General purpose models for unstructured data.
  - ❖ Effective algorithm for data fitting and analysis under uncertainty.
- 
- But currently it is always used as a black box.

Belief v.s. Probability



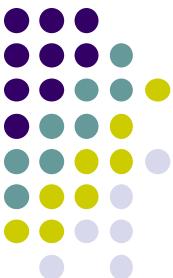
# Data driven modeling



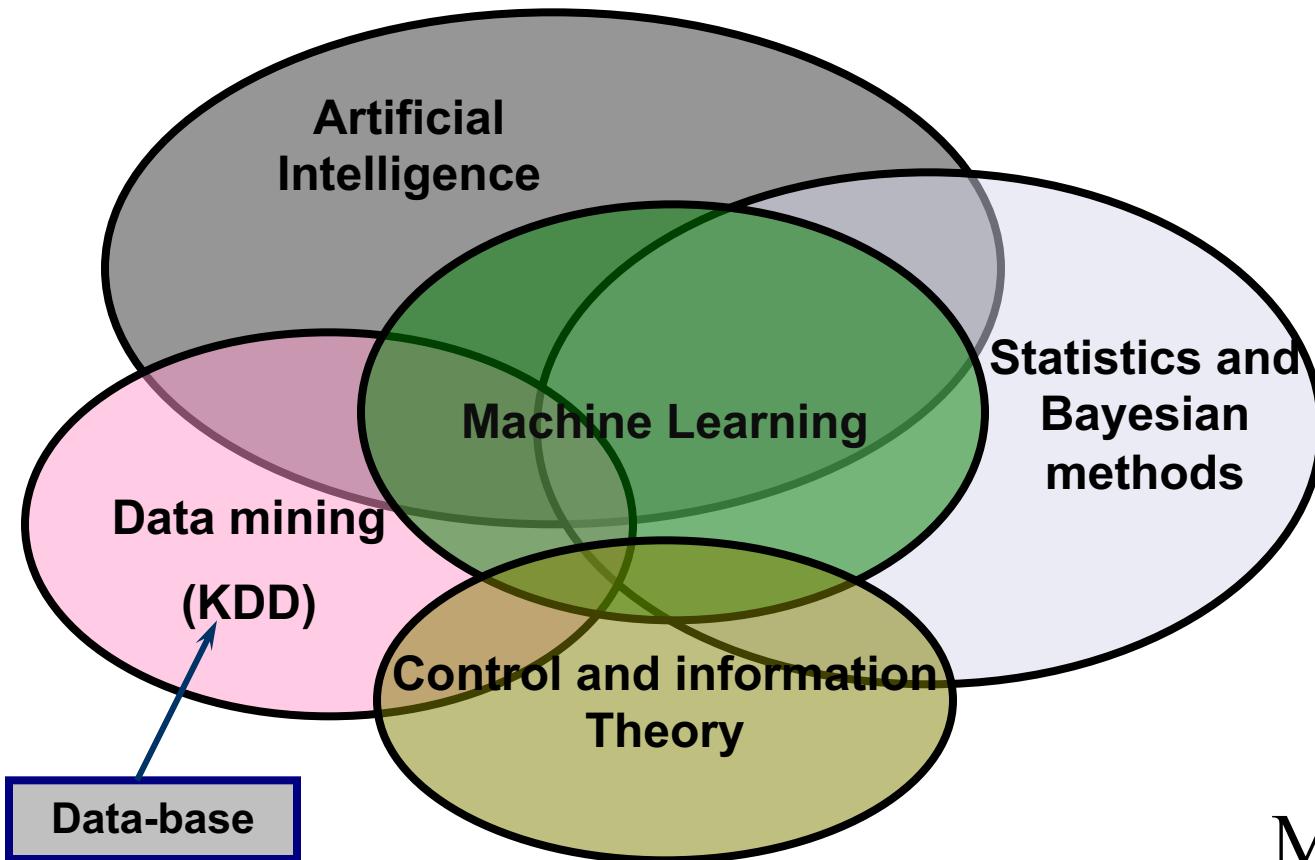


# Data-driven vocabulary

- 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

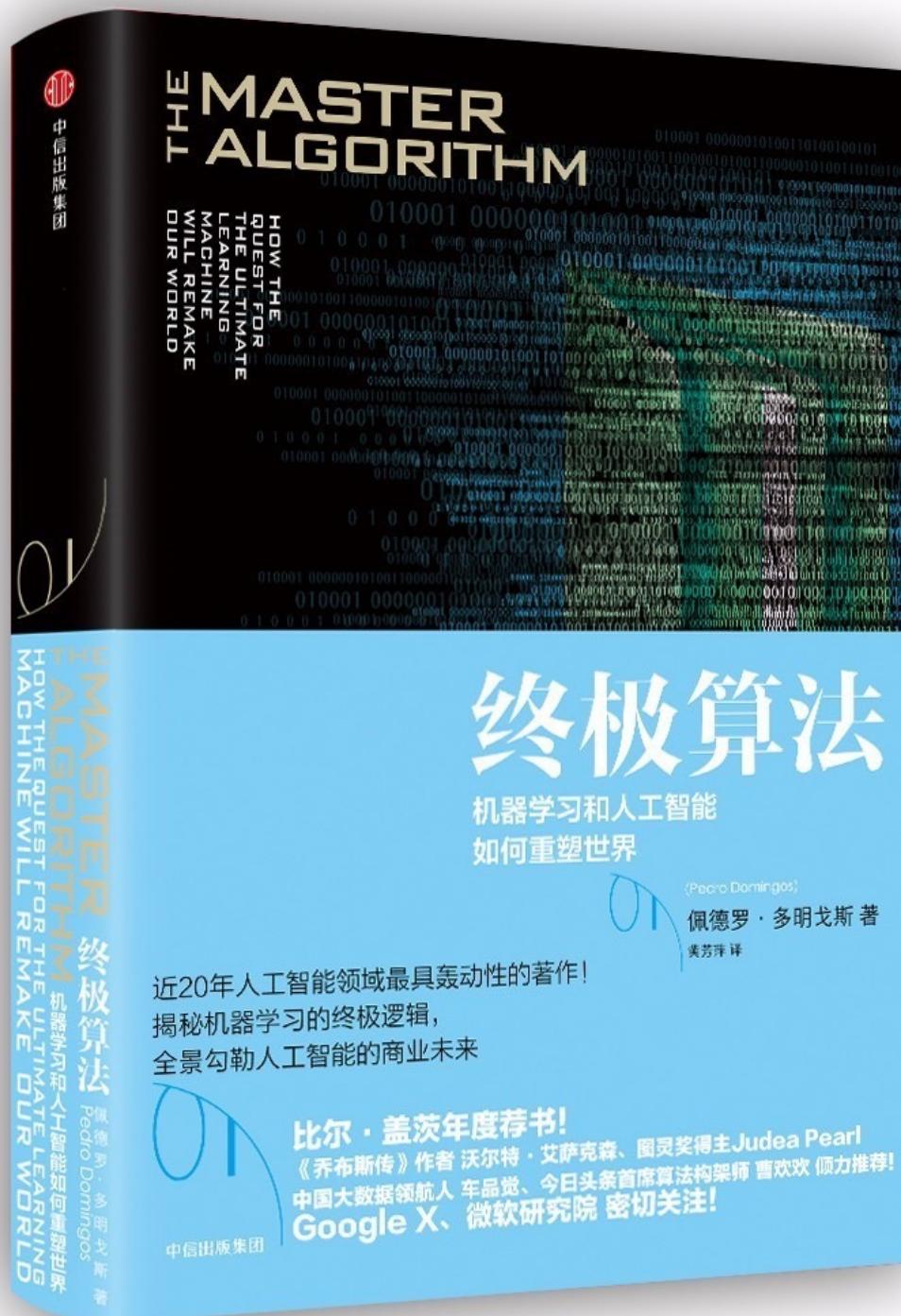
Information  
retrieval

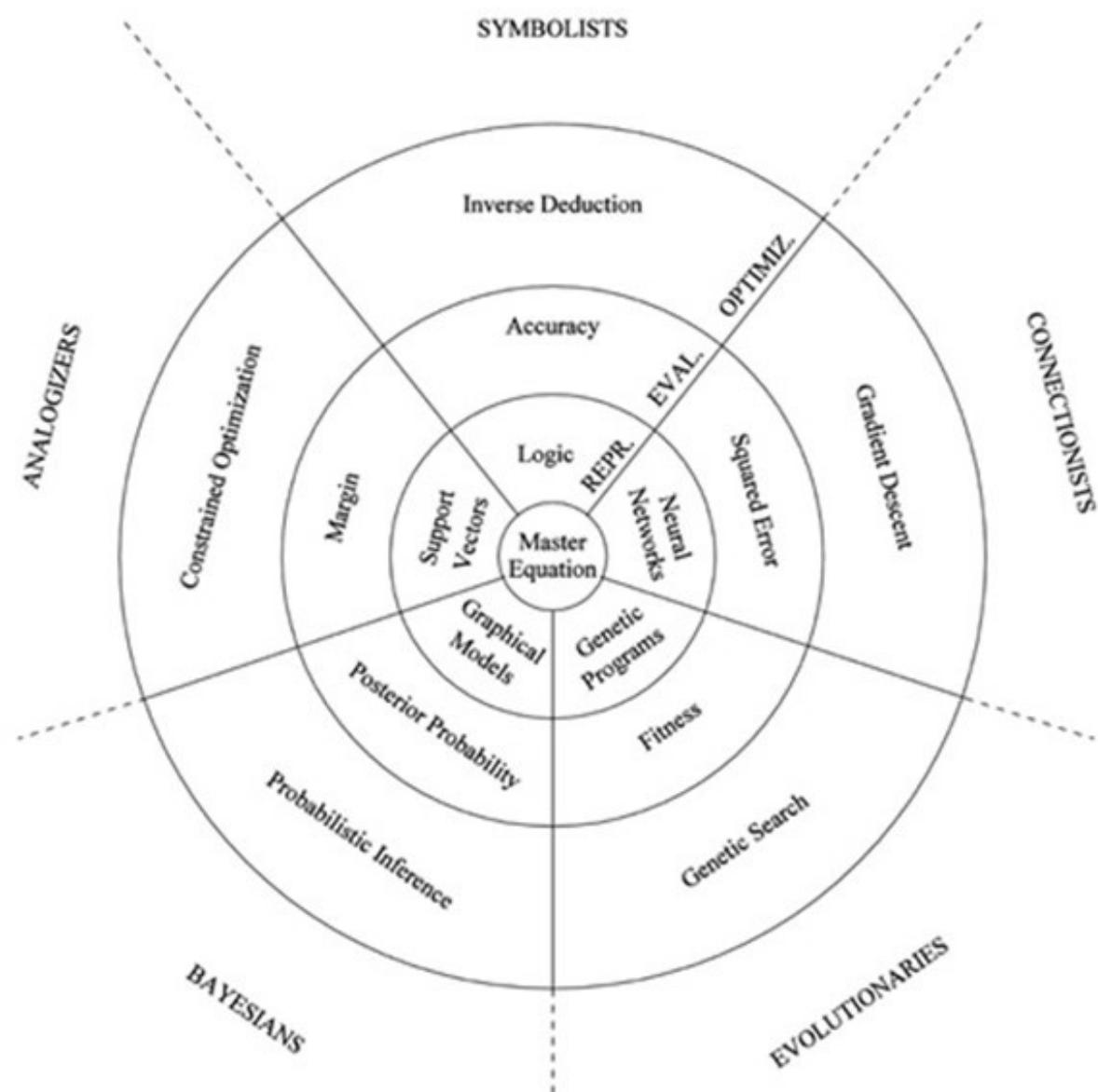
"PEDRO DOMINGOS DEMYSTIFIES MACHINE LEARNING AND SHOWS HOW WONDROUS  
AND EXCITING THE FUTURE WILL BE." —WALTER ISAACSON

# THE MASTER ALGORITHM

HOW THE QUEST FOR  
THE ULTIMATE  
LEARNING MACHINE WILL  
REMAKE OUR WORLD

PEDRO DOMINGOS





机器学习主要有5个学派：  
符号学派

将学习看作逆向演绎，并从哲学、心理学、逻辑学中寻求洞见；

联结学派

对大脑进行逆向分析，灵感来源于神经科学和物理学；

进化学派

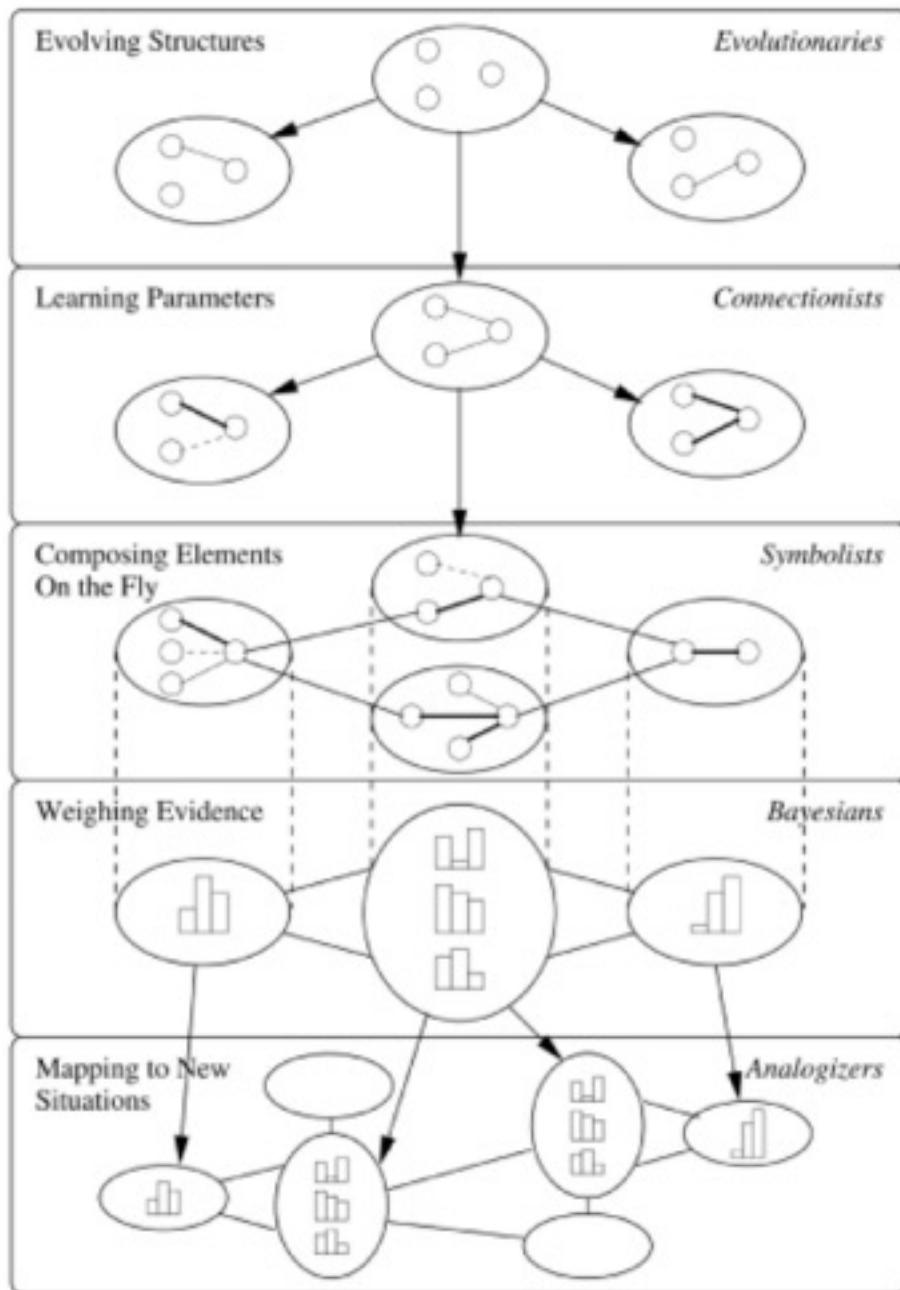
在计算机上模拟进化，并利用遗传学和进化生物学知识；

贝叶斯学派

认为学习是一种概率推理形式，理论根基在于统计学；

类推学派

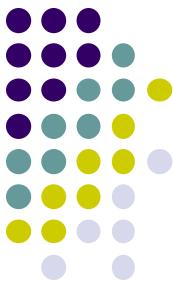
通过对相似性判断的外推来进行学习，并受心理学和数学最优化的影响。





# What is machine learning?

- 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

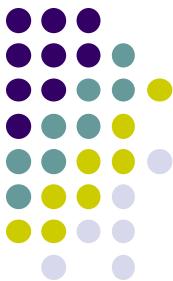
Different from standard CS: want to implement unknown function, only have access to sample input-output pairs (training examples)

# Main categories of learning problems



Learning scenarios differ according to the available information in training examples

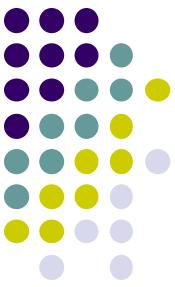
- **Supervised**: correct output available
  - **Classification**: 1-of-N output (speech recognition, object recognition, medical diagnosis)
  - **Regression**: real-valued output (predicting market prices, temperature)
- **Unsupervised**: no feedback, need to construct measure of good output
  - **Clustering** : 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?

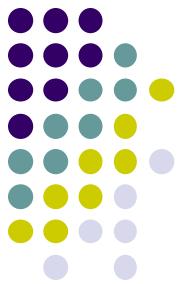
- **Develop enhanced computer systems**
  - automatically adapt to user, customize
  - often difficult to acquire necessary knowledge
  - discover patterns offline in large databases (*data mining*)
- **Improve understanding of human, biological learning**
  - computational analysis provides concrete 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

# Is it really useful for computer science and technology?

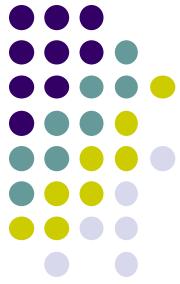


- Con: Everything is machine learning or everything is human tuning?
  - Sometimes, this may be true
- Pro: more understanding of learning, but yields much more powerful and effective algorithms.
  - Problem taxonomy
  - General-purpose models
  - Reasoning with probabilities
- ❖ I believe the mathematic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

# What will be a successful D-D algorithm?



- Computational efficiency
- Robustness
- Statistical stability



Old and New ...

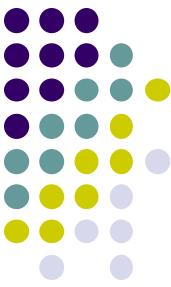
# APPLICATIONS OF DATA-DRIVEN



# The First Example: Google!



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



# The great example in China: TikTok/Douyin (ByteDance)



We regret to inform you that we have discontinued operating TikTok in Hong Kong.

## Make Your Day

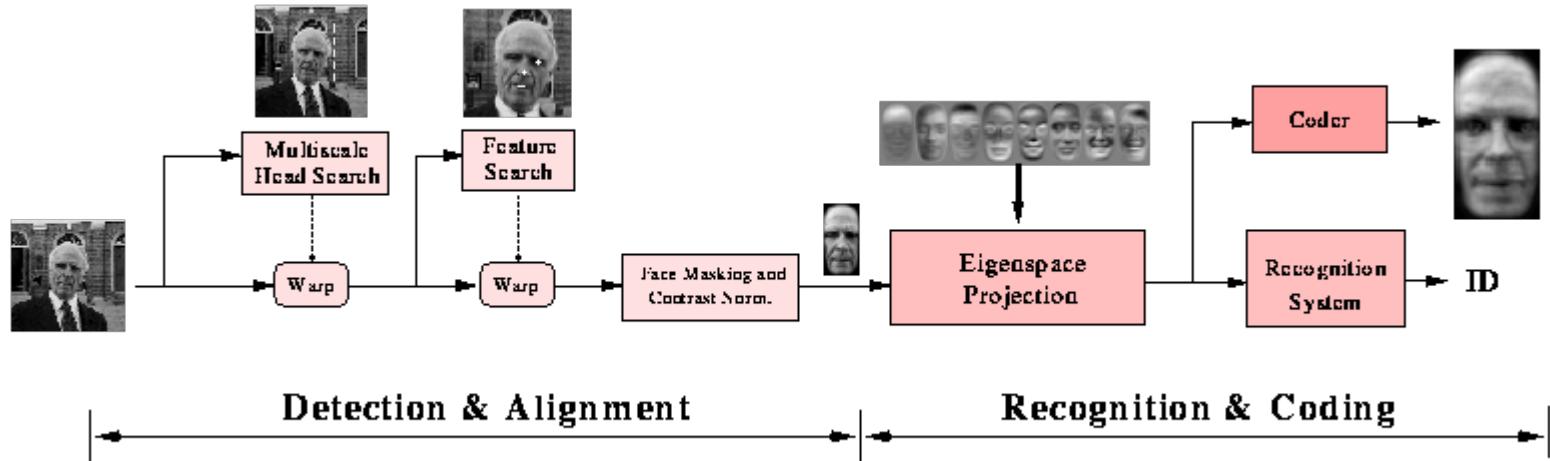
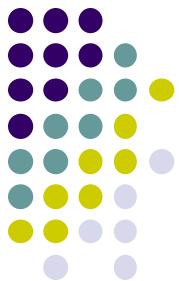
Real People. Real Videos.

- 月活用户>10亿
- 应用下载>20亿
- 用户每天>52分钟

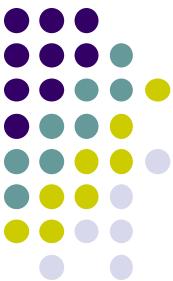


<https://www.affde.com/zh-CN/50-important-tiktok-statistics-for-2021.html>

# Object detection and recognition - the power of DD



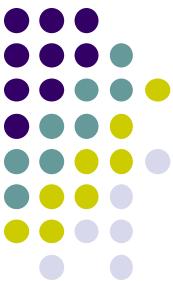
The image is copied from  
<http://vismod.media.mit.edu/vismod/demos/facerec/>



# 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]

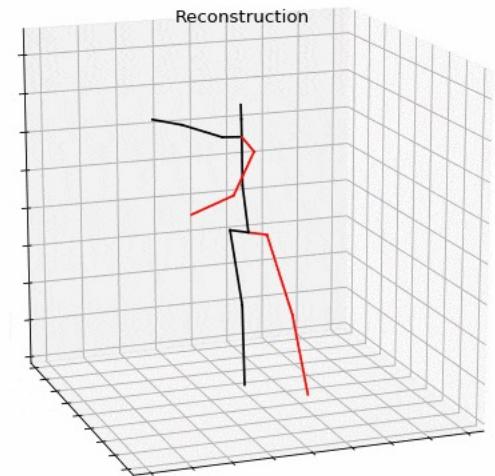


# In Olympic Games Beijing 2022



3D Modeling and Visualization

3D Skeleton Detection

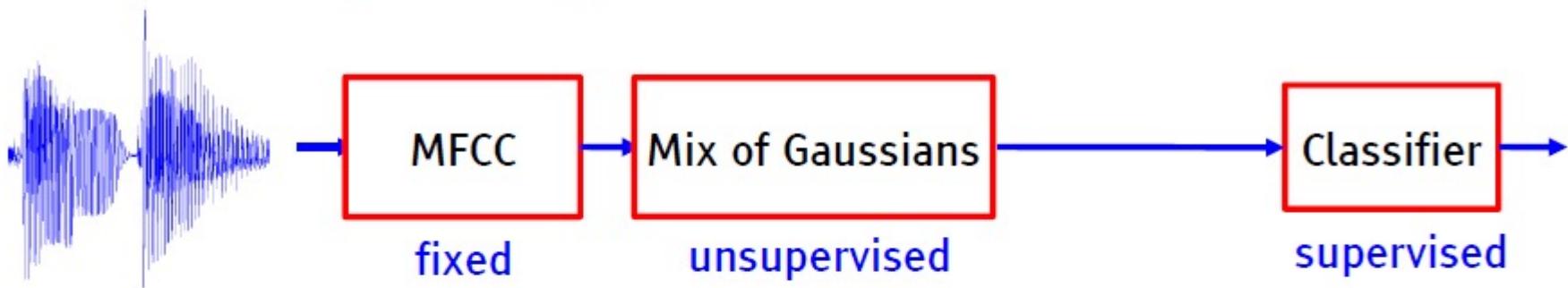




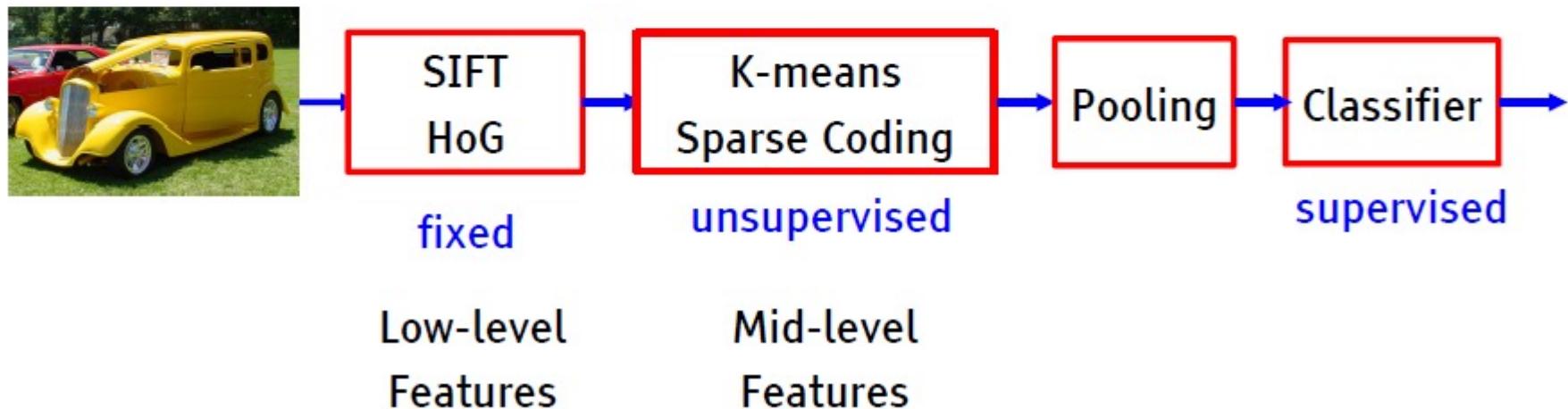
# Speech recognition

## ■ Modern architecture for pattern recognition

▶ Speech recognition: early 90's – 2011



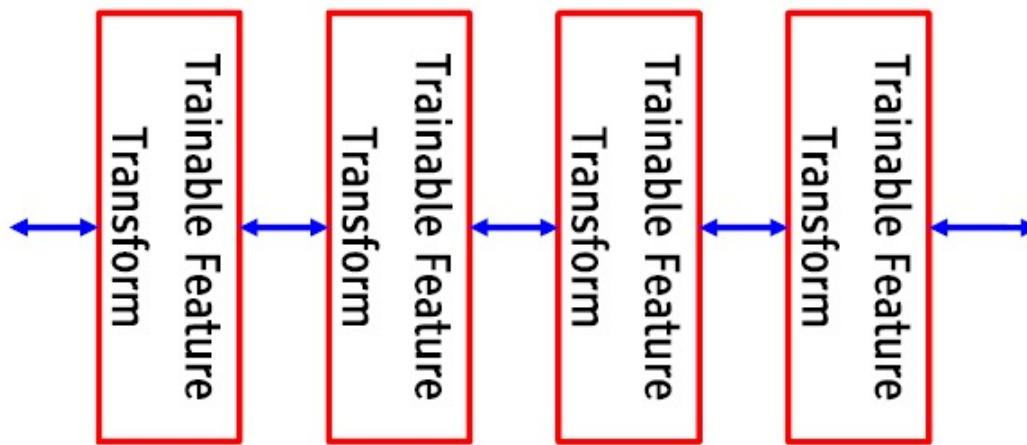
▶ Object Recognition: 2006 - 2012



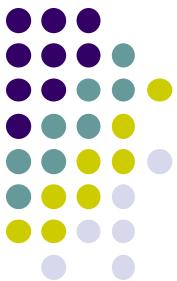


# Speech recognition

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



# Document processing – Bayesian classification

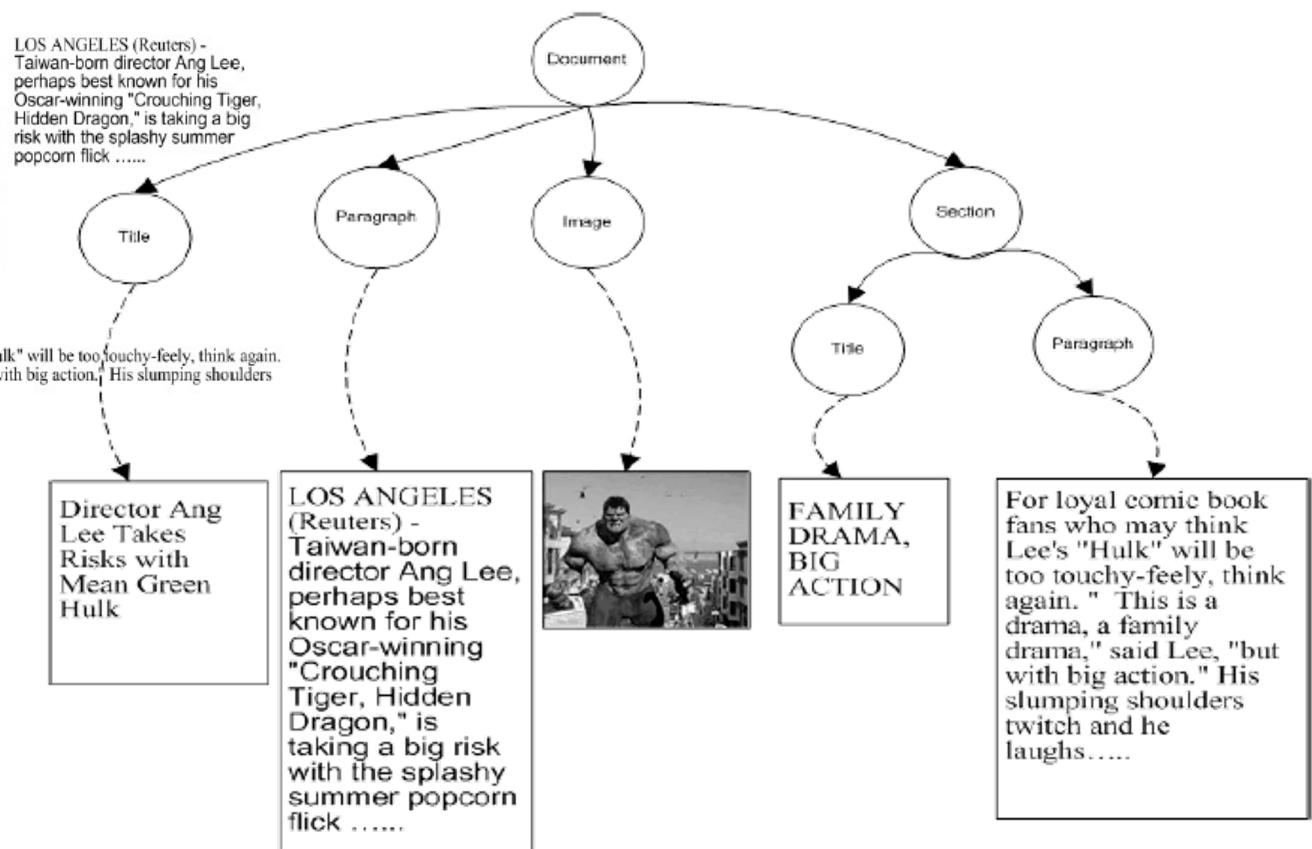


## Director Ang Lee Takes Risks with Mean Green 'Hulk'

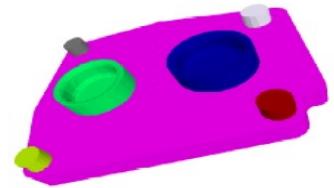


### FAMILY DRAMA, BIG ACTION

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.....



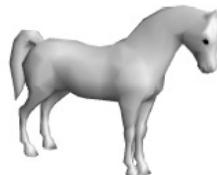
# Mesh Processing – Data clustering/segmentation



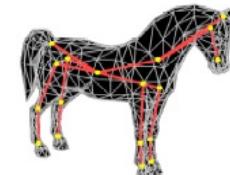
(c) mechanical part – 1270 faces  
7 patches



(d) heart – 1619 faces  
4 patches



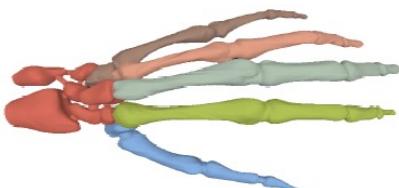
(a) object



(b) skeleton



(e) Venus – 67,170 faces  
3 patches



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



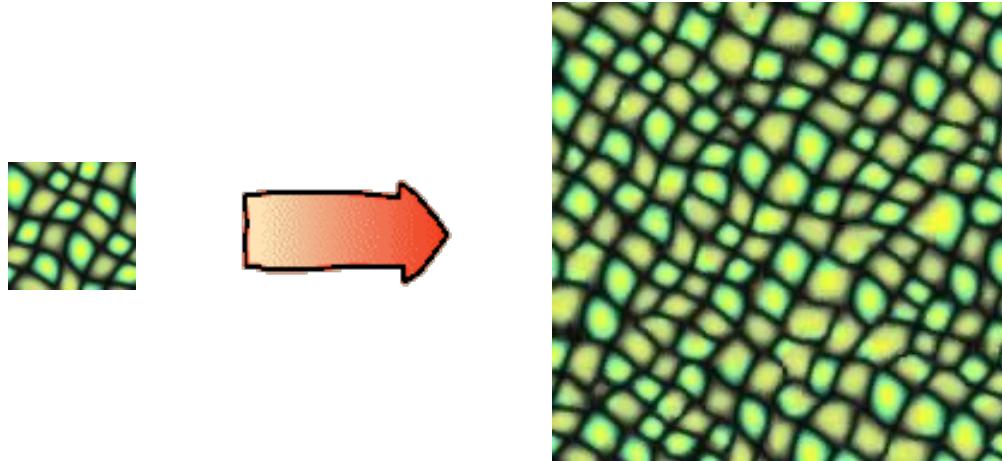
(c) deformed skeleton



(d) deformed object

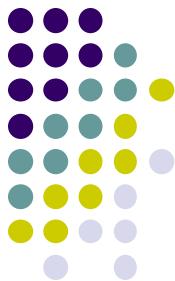
- *Hierarchical Mesh Decomposition using Fuzzy Clustering and Cuts.*  
By Sagi Katz and Ayellet Tal, SIGGRAPH 2003

# Texture synthesis and analysis – Hidden Markov Model



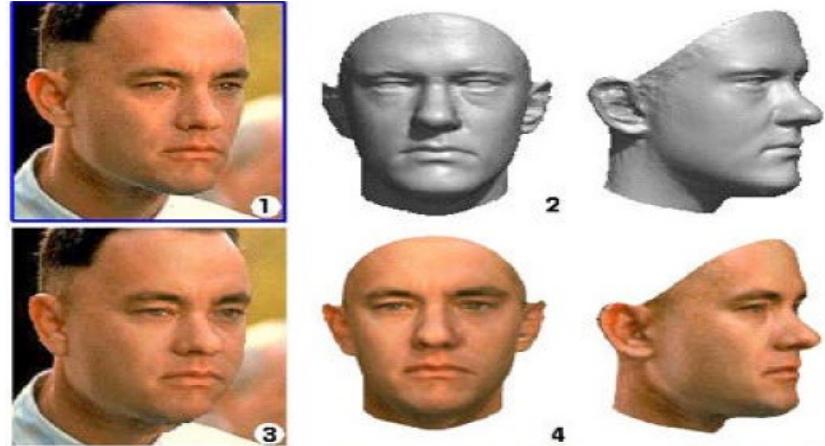
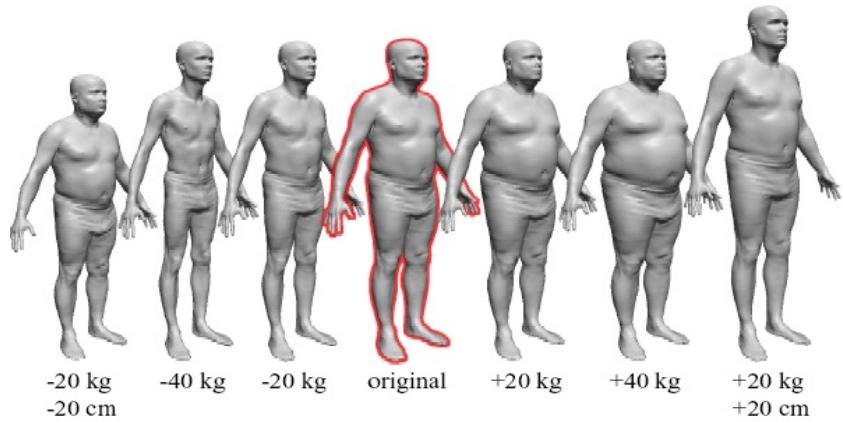
- *Texture Synthesis over Arbitrary Manifold Surfaces.* Li-Yi Wei and Marc Levoy. SIGGRAPH 2001.
- *Fast Texture Synthesis using Tree-structured Vector Quantization.* Li-Yi Wei and Marc Levoy. SIGGRAPH 2000.

# Reflectance texture synthesis – Dimension reduction



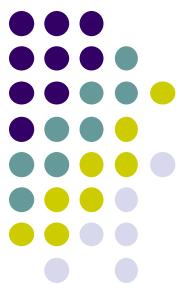
- *Synthesizing Bidirectional Texture Functions for Real-World Surfaces.* Xinguo Liu, Yizhou Yu and Heung-Yeung Shum. SIGGRAPH 2001.
- More recent papers...

# 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.

# Learning a Probabilistic Latent Space of Object Shapes – GANs (NIPS2016)



## Single Image 3D Reconstruction



Input  
image

Reconstructed  
3D shape

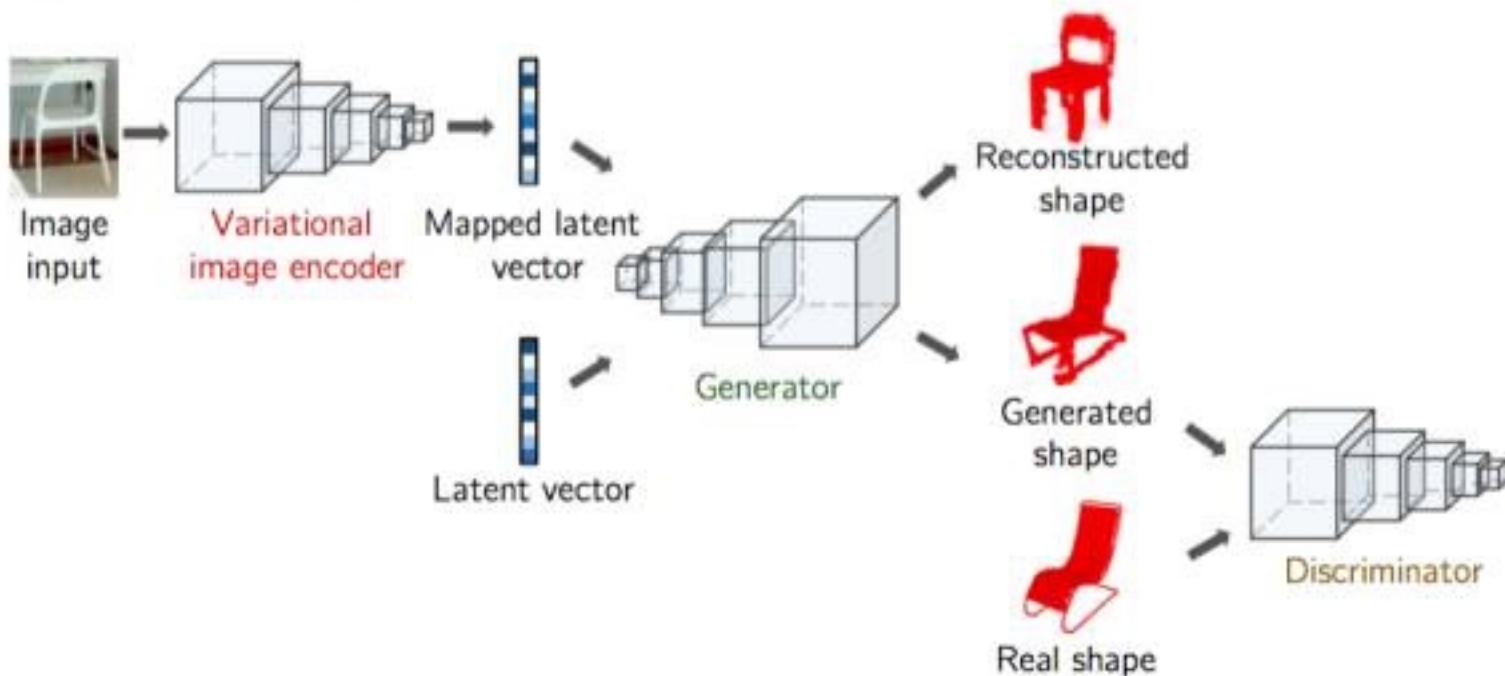
Input  
image

Reconstructed  
3D shape

# Learning a Probabilistic Latent Space of Object Shapes – GANs

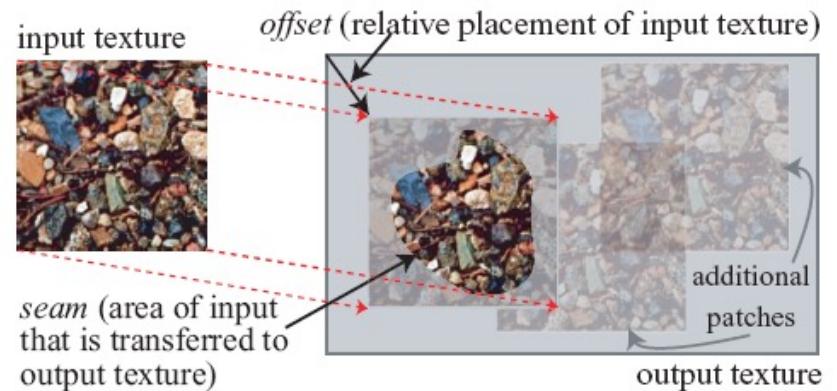
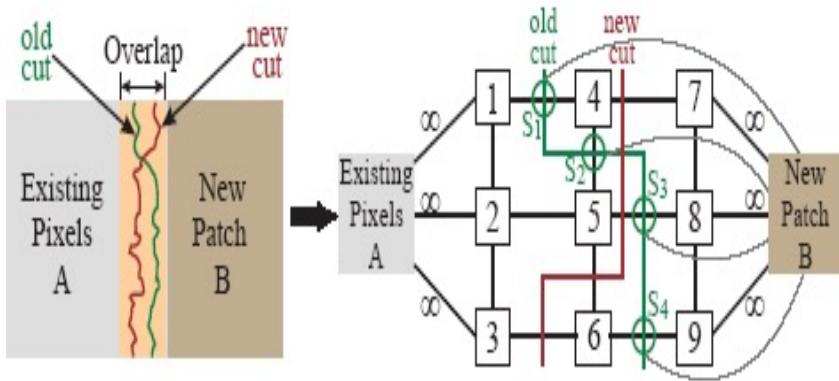
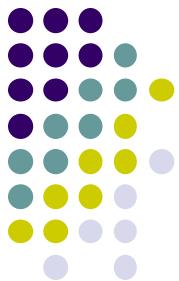


## Model: 3D-VAE-GAN



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

# 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.



# Data-driven based Rendering

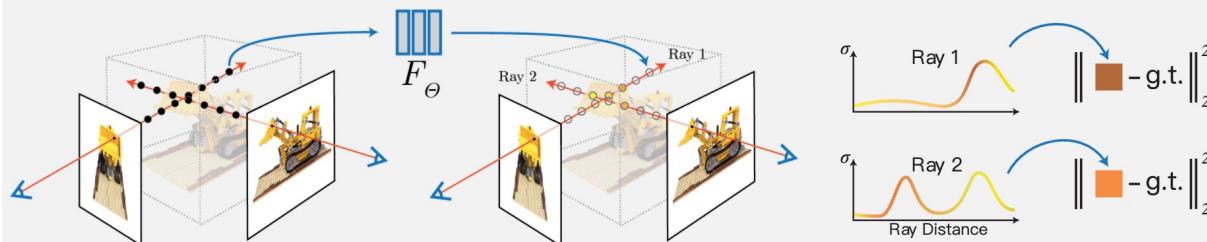
## • NeRF: Neural Radiance Fields (ECCV2020)

### Abstract & Method

We present a method that achieves state-of-the-art results for synthesizing novel views of complex scenes by optimizing an underlying continuous volumetric scene function using a sparse set of input views.

$$(x, y, z, \theta, \phi) \rightarrow \begin{array}{c} | \\ | \\ | \\ | \end{array} \rightarrow (RGB\sigma)$$
$$F_{\Theta}$$

Our algorithm represents a scene using a fully-connected (non-convolutional) deep network, whose input is a single continuous 5D coordinate (spatial location ( $x, y, z$ ) and viewing direction ( $\theta, \phi$ )) and whose output is the volume density and view-dependent emitted radiance at that spatial location.



We synthesize views by querying 5D coordinates along camera rays and use classic volume rendering techniques to project the output colors and densities into an image. Because volume rendering is naturally differentiable, the only input required to optimize our representation is a set of images with known camera poses. We describe how to effectively optimize neural radiance fields to render photorealistic novel views of scenes with complicated geometry and appearance, and demonstrate results that outperform prior work on neural rendering and view synthesis.

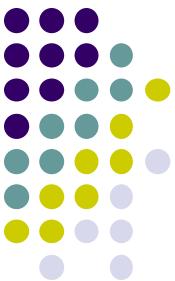


# Human Motion - Time series analysis



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.

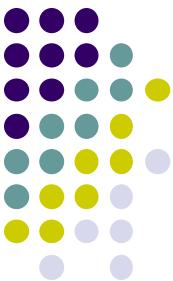
- *Style Machines*. M. Brand and A. Hertzmann. SIGGRAPH 2000.
- *A Data-Driven Approach to Quantifying Natural Human Motion*. L. Ren, A. Patrick, A. Efros, J. Hodgins, J. Rehg. SIGGRAPH 2005



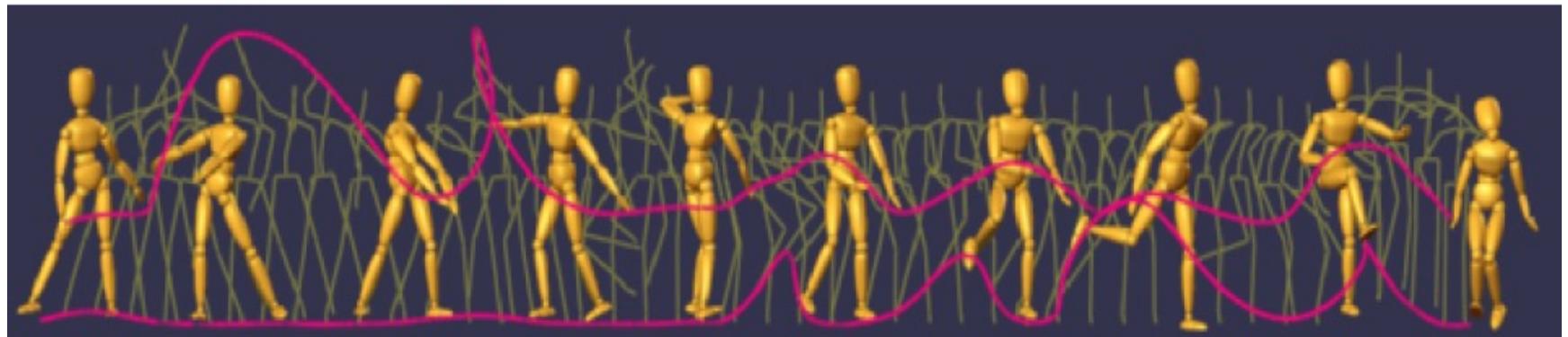
# Video Textures - Reinforcement Learning



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



# Motion texture - Linear dynamic system



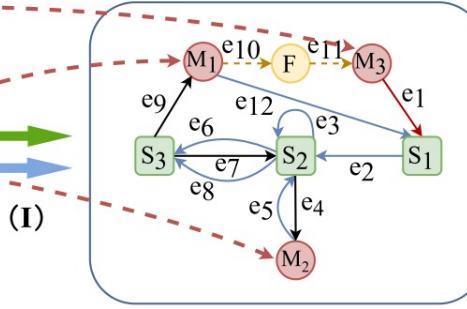
- *Motion Texture: A Two-Level Statistical Model for Character Motion Synthesis.* Yan Li, Tianshu Wang, and Heung-Yeung Shum. SIGGRAPH 2002.

```

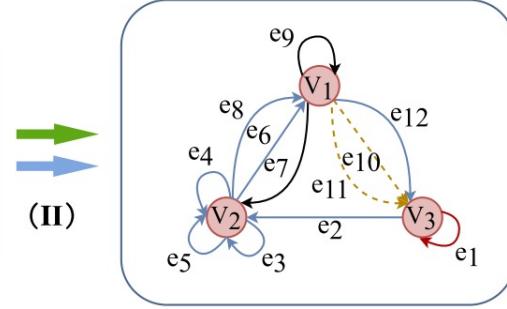
1 function getBonus(address recipient) {
2   require(!Bonus[recipient]);
3   Reward[recipient] += 100;
4   withdraw(recipient);
5   Bonus[recipient] = true;
6 }
7 function withdraw(address recipient) {
8   uint amount = Reward[recipient];
9   Reward[recipient] = 0;
10  recipient.call.value(amount)();
11 }

```

(a) Contract snippet



(b) Contract graph



(c) Normalized graph

Temporal Edges

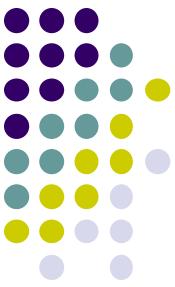
	V <sub>start</sub>	V <sub>end</sub>	Order	Type
e <sub>1</sub>	V <sub>3</sub>	V <sub>3</sub>	1	RG
e <sub>2</sub>	V <sub>3</sub>	V <sub>2</sub>	2	AC
e <sub>3</sub>	V <sub>2</sub>	V <sub>2</sub>	3	FW
...	...	...	...	...
e <sub>10</sub>	V <sub>1</sub>	V <sub>3</sub>	10	FB
e <sub>11</sub>	V <sub>1</sub>	V <sub>3</sub>	11	FB
e <sub>12</sub>	V <sub>1</sub>	V <sub>3</sub>	12	AG

Clustered Nodes

Clus	Maj	S <sub>in</sub>		S <sub>out</sub>
		Sec	S <sub>in</sub>	S <sub>out</sub>
V <sub>1</sub>	M <sub>1</sub>		S <sub>2</sub> , S <sub>3</sub>	F, S <sub>1</sub>
V <sub>2</sub>	M <sub>2</sub>		S <sub>1</sub> , S <sub>2</sub>	S <sub>2</sub>
V <sub>3</sub>	M <sub>3</sub>		F	S <sub>1</sub>

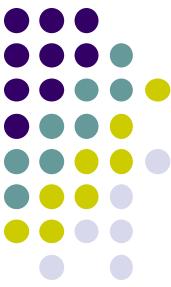
● Major Node   ■ Secondary Node   ○ Fallback Node   → Forward Edge   → Control-flow Edge   → Data-flow Edge   - -> Fallback Edge

# Smart Contract Vulnerability Detection ( IJCAI2020 )



# Summary

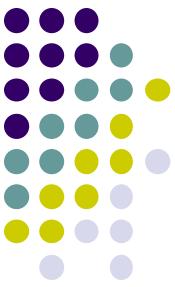
- Learning (from Data) is a nut-shell, :-D
  - Keywords
    - Noun: data, models, patterns, features;
    - Adj.: probabilistic, statistical;
    - Verb: fitting, reasoning, mining.



# Questions for review

- Try to find potential learning based (data driven) applications in your research area
- Is there any disadvantage / weakness?





# Reference

- Reinforcement learning: A survey

# The End

新浪微博: @浙大张宏鑫

微信公众号:

