

Modeling & Computation

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Learning from Data



Future Data



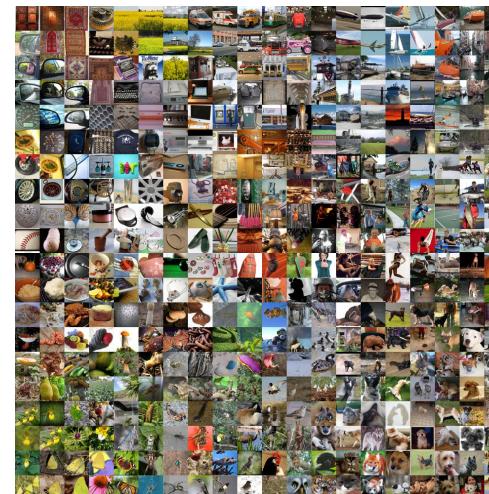
Model



Predict

“car”

Train



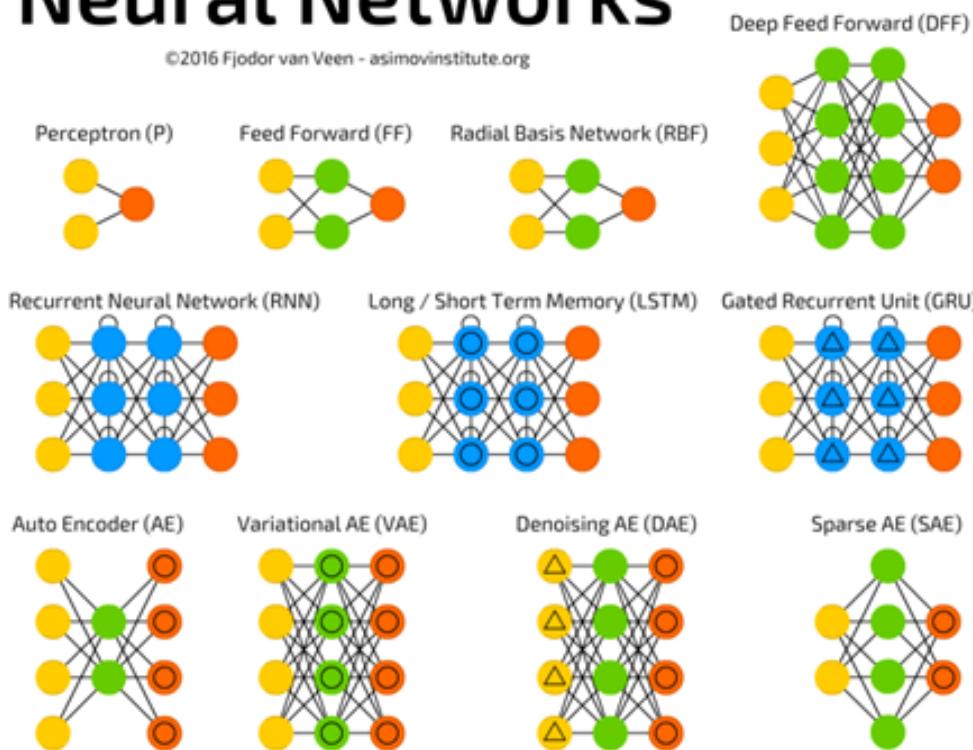
Past Data
(perhaps with **labels**)

Machine Learning in Practice

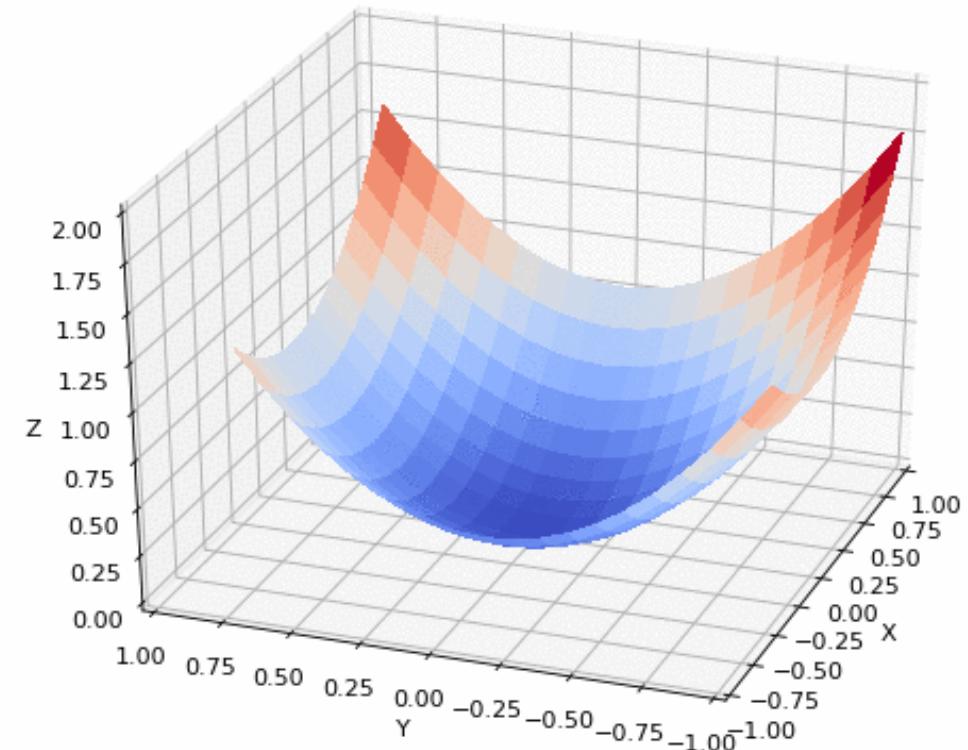
Modeling

Neural Networks

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Computation



Modeling

Linear Models

- Feature vector: $\mathbf{x} \in \mathbb{R}^d$ (e.g., features of a house).
- Prediction: $f(\mathbf{x}) = \mathbf{x}^T \mathbf{w}$ (e.g., housing price).

Linear Models

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- $f(\mathbf{x}) = w_1 x_1 + w_2 x_2 + \cdots + w_d x_d$
- w_1, w_2, \dots, w_d : weights
- x_1 : # of bedrooms
- x_2 : # of bathroom
- x_3 : square feet
- ...

Linear Models

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Question: How to find \mathbf{w} ?



Price = \$0.5M

Features of a House

$$\mathbf{x} \in \mathbb{R}^d$$

Prediction:

$$f(\mathbf{x}) = \mathbf{x}^T \mathbf{w}$$

Linear Models

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Question: How to find \mathbf{w} ?

- Training features: $\mathbf{x}_1, \dots, \mathbf{x}_n \in \mathbb{R}^d$.
- Training targets: $y_1, \dots, y_n \in \mathbb{R}$.



• • •



totally n houses

Linear Models

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- Loss function: $L(\mathbf{w}) = \frac{1}{n} \sum_{i=1}^n (\mathbf{x}_i^T \mathbf{w} - y_i)^2$.
- Least squares regression: $\mathbf{w}^\star = \min_{\mathbf{w}} L(\mathbf{w})$.

Linear Models Are Not Expressive

Example: Given a person's photo, predict her/his age.



Age = 36

Photo (features)

Linear Models Are Not Expressive

线性模型具有 不可解释性

Example: Given a person's photo, predict her/his age.



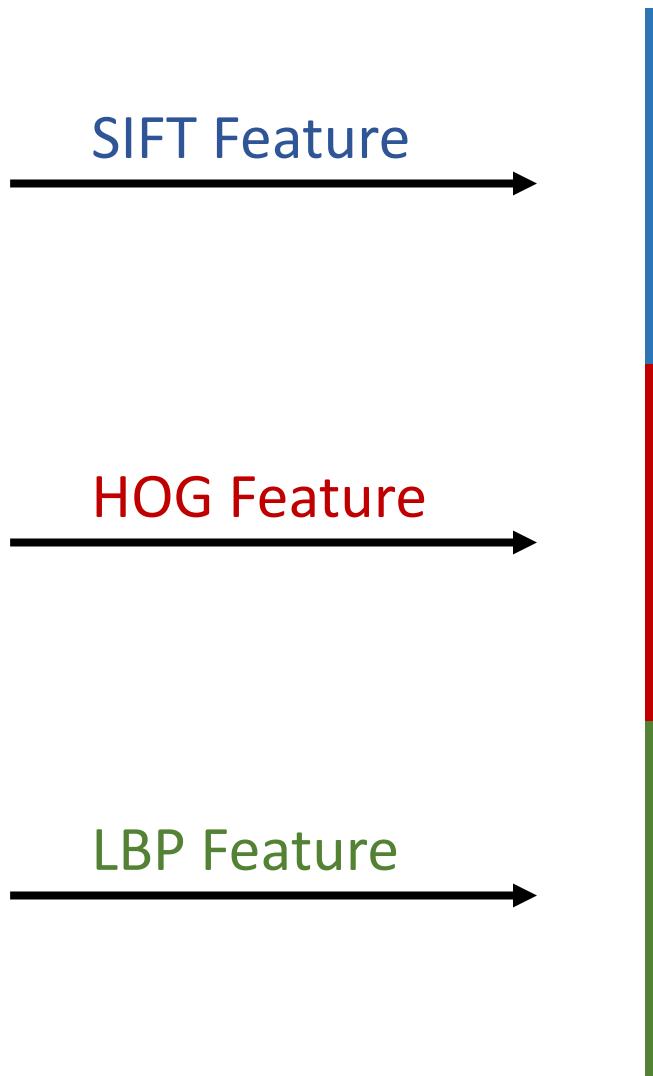
Age = 36

Photo (features)

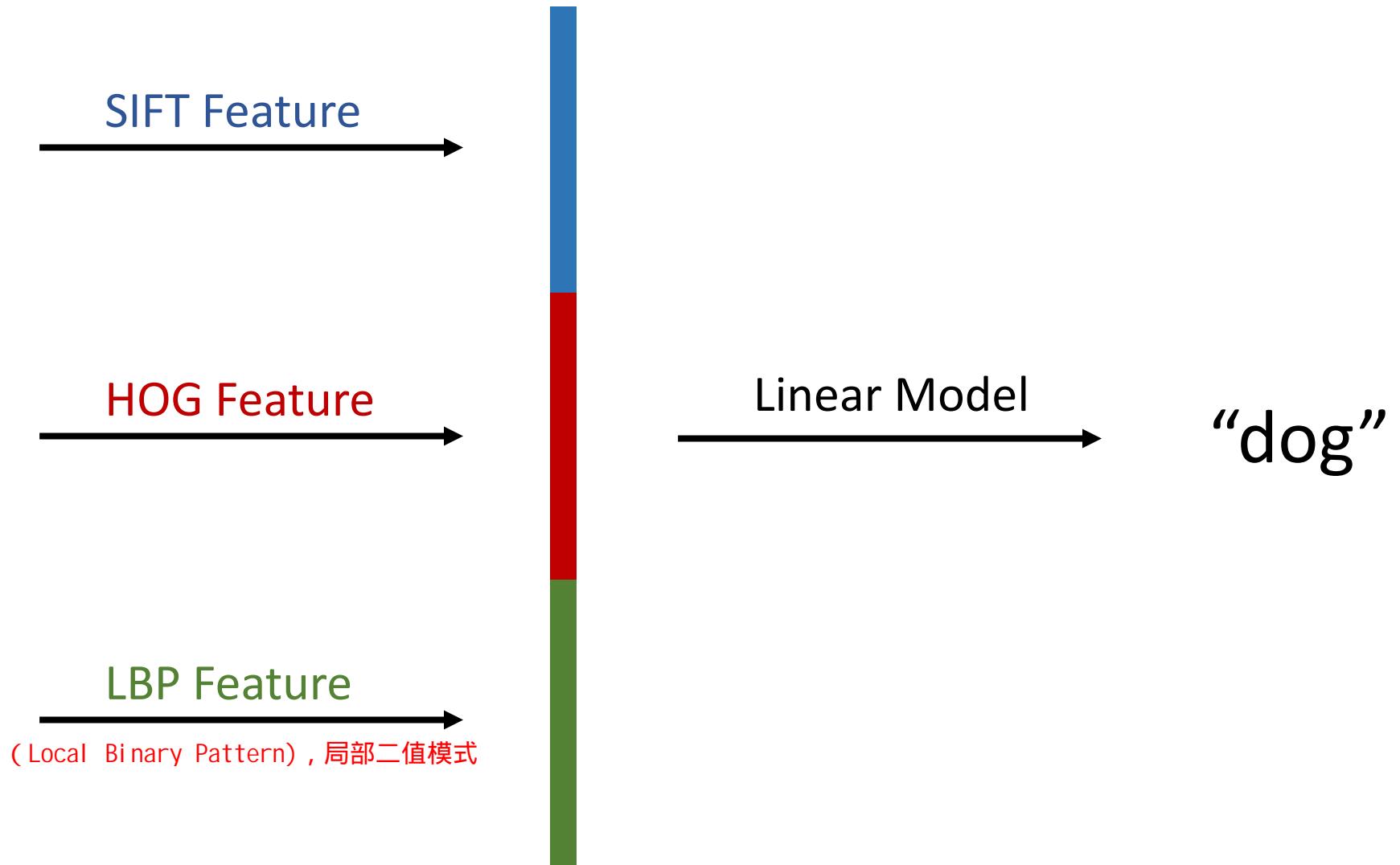
Question: Can we use linear regression?

- Linear regression assumes the **target** is a weighted average of every **pixel in the photo**.
- Linear regression works poorly for age prediction. 线性回归 年龄预测 效果不佳

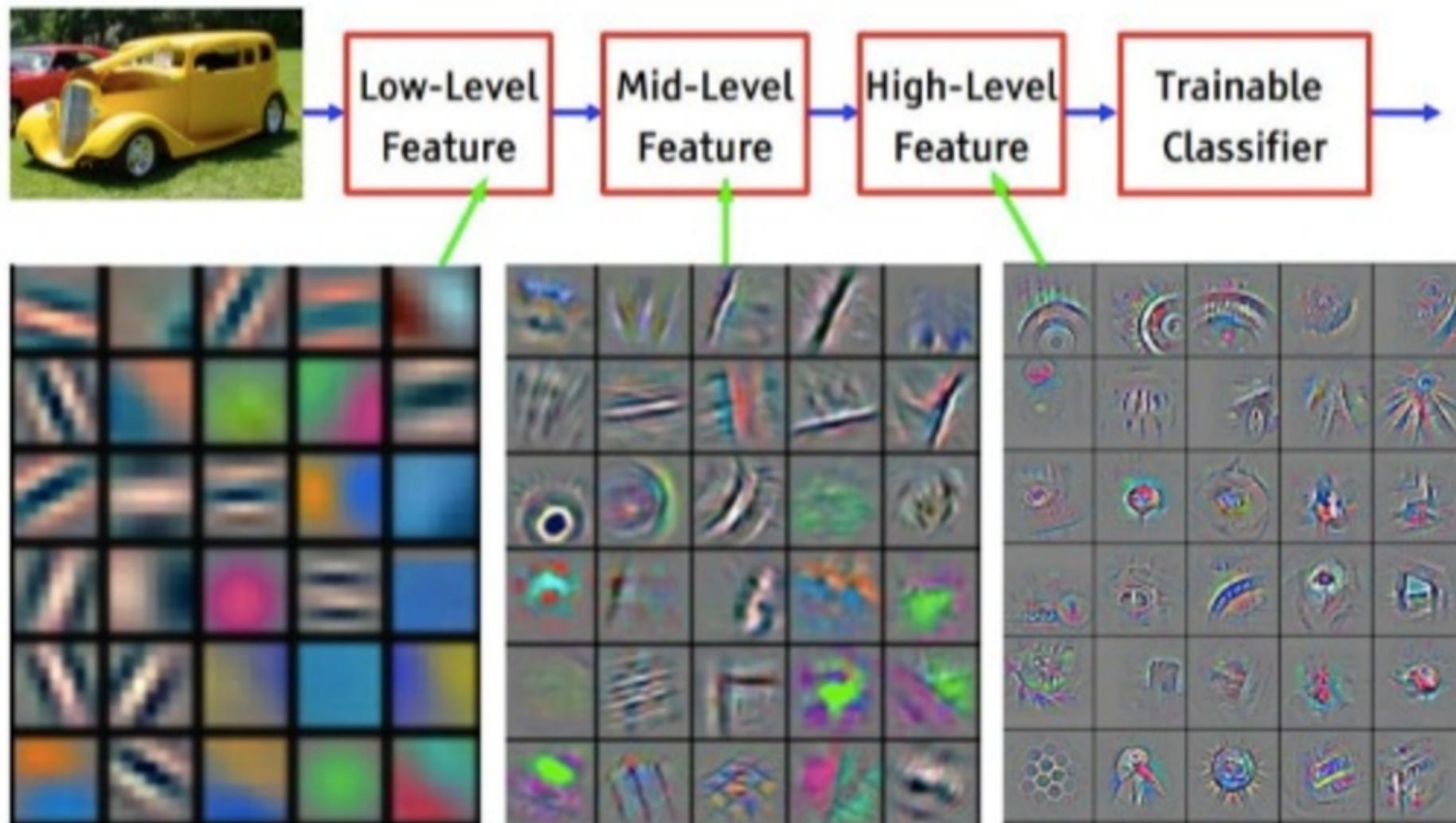
Traditional Approaches



Traditional Approaches



Convolutional Neural Networks (CNNs)

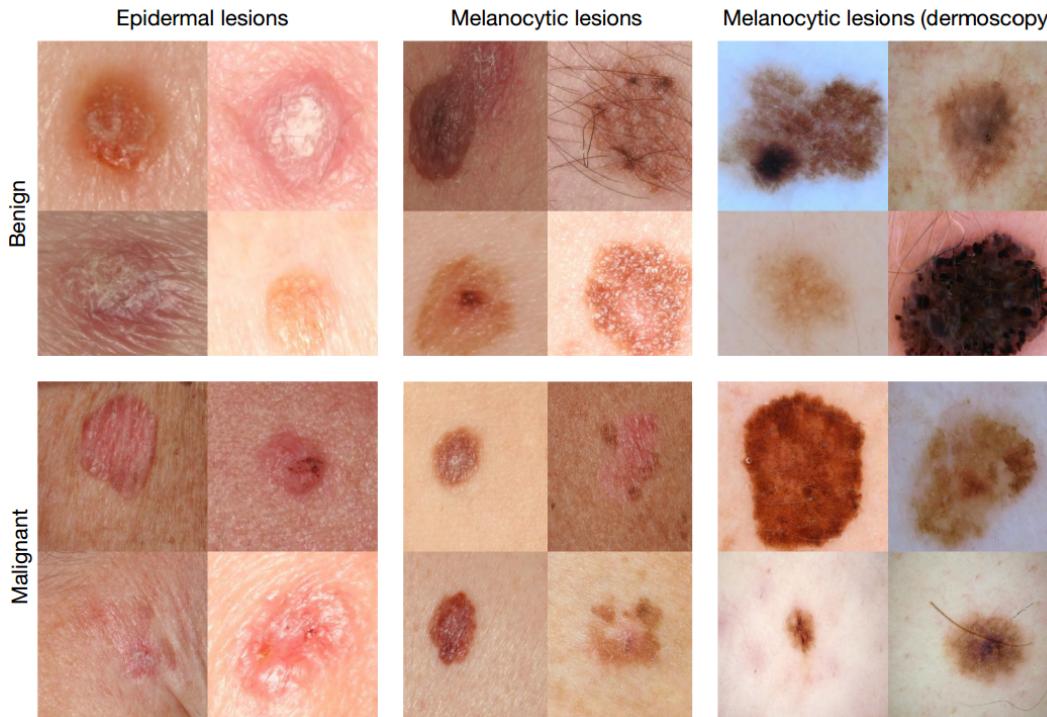


Applications of CNNs



- CNNs are suitable for image data.
- CNNs convert images to effective representations. (Feature extraction.)
- Applications:
 - Image/video recognition.
 - Face recognition.
 - Image generation.
 - ...

Applications of CNNs: Medical Diagnosis



皮肤癌 诊断

Example: Skin cancer diagnosis

- Input: an image.
- Outputs:
 - Is it skin cancer?
 - Benign or malignant?
- The same accuracy as human experts.

Reference

[1] Esteva et al. [Dermatologist-level classification of skin cancer with deep neural networks](#). *Nature*, 2017.

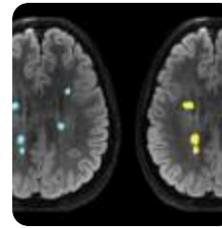
Applications of CNNs: Medical Diagnosis

 Healthcare IT News

Australia's SNAC develops AI tools to improve brain scan analysis

... algorithms using the NVIDIA Clara suite of medical imaging tools, as well ... SNAC is building its deep learning algorithms using the PyTorch ...

10 hours ago



 Imaging Technology News (press release) (blog)

Improved Imaging Technique Could Increase Chances of Prostate Cancer Survival

"Imaging has become an essential component of modern medicine," said ... Using a deep learning approach, Yan plans to simplify the process.

20 hours ago



 Data Center Frontier (blog)

Startups Target AI Opportunities to Disrupt Medical Imaging

AI, or more specifically machine learning and deep learning algorithms, are learning how to analyze the images produced by radiology scans.

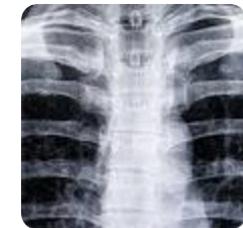
2 weeks ago



 Xtelligent Healthcare Media

Deep Learning Tool Detects Cancer in Radiology Reports

Human reviewers analyzed the imaging text reports and noted whether cancer ... The team then used these reports to train a deep learning algorithm to ... Amazon's Comprehend Medical, an advanced machine



3 weeks ago

 Medical Xpress

Deep learning model detects diabetic eye diseases accurately

The deep learning model identified referable diabetic retinopathy ... Currently, retinal imaging is the most widely used method for screening and detecting retinopathy, and medical experts evaluate the severity and the

2 days ago



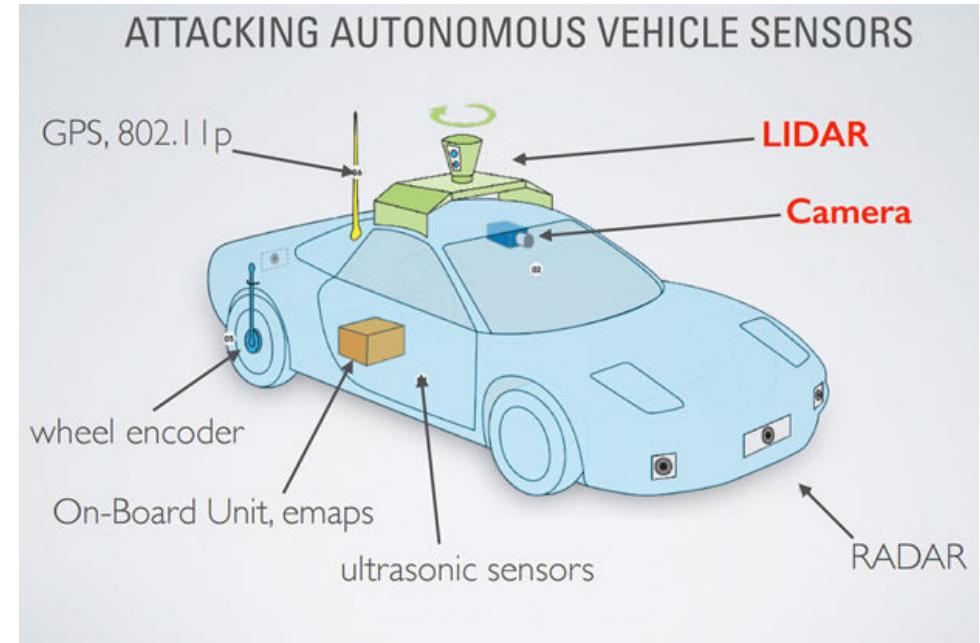
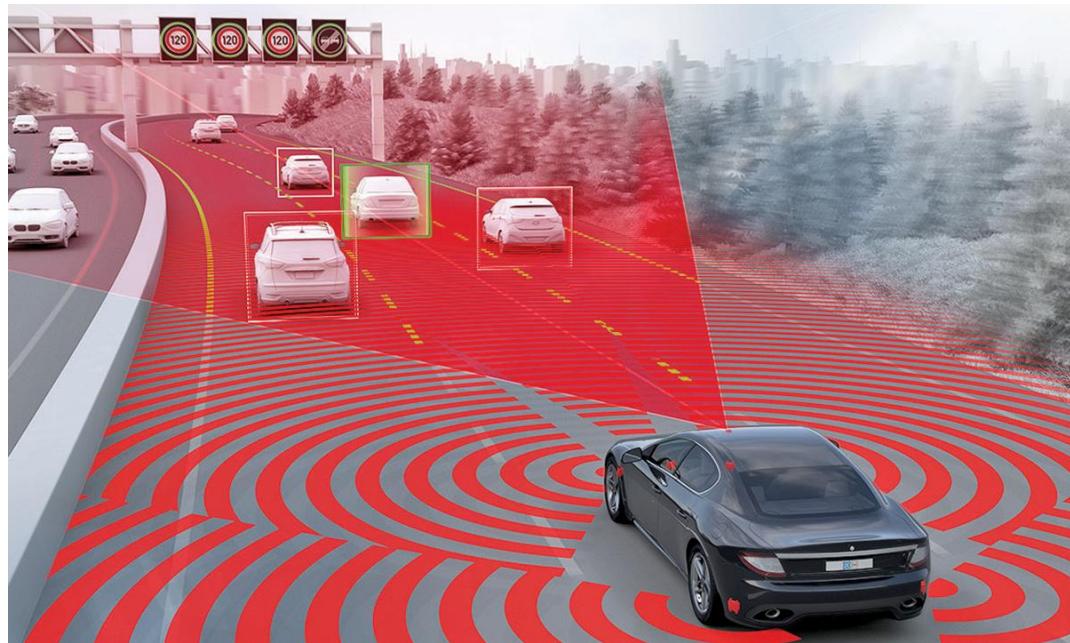
 Yahoo Finance

Progenics Pharmaceuticals Announces Collaboration with Veterans Affairs on the AI Research Program for Medical Image Analysis

The collaboration with VA Greater Los Angeles Healthcare System is nation's first to validate deep learning algorithms in medical imaging of ...

4 weeks ago

Applications of CNNs: Self Driving Cars

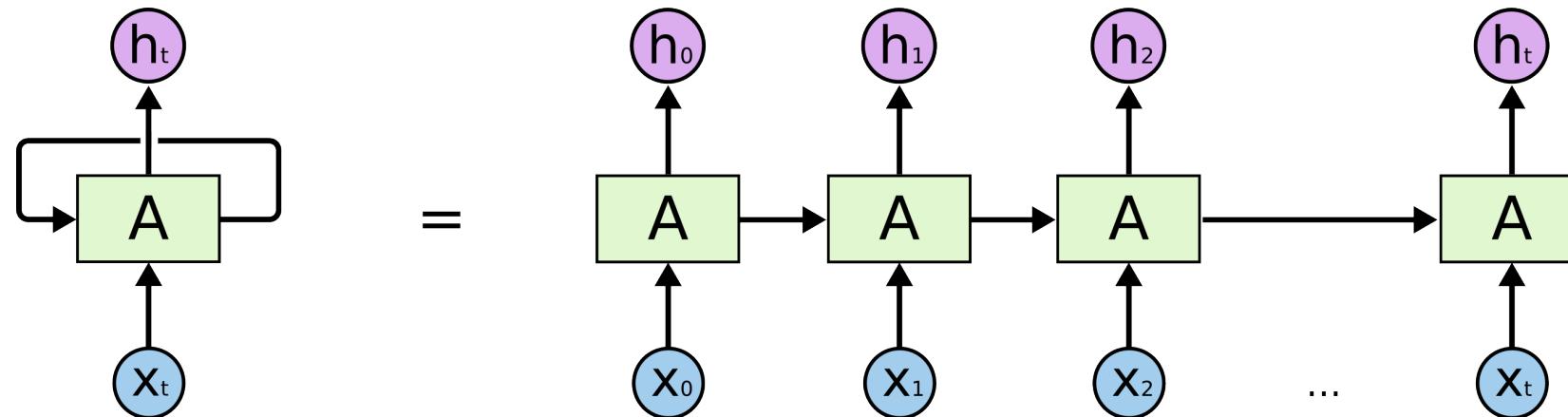


- CNNs play an import role in self driving cars.
 - Understand images taken by the cameras. 图像理解 (摄像机拍摄的图片)
 - Recognize signs, cars, pedestrians, and obstacles. 识别标志、汽车、行人和障碍物
- CNN is not everything; self-driving car is a sophisticated system.

自动驾驶汽车是一个复杂的系统

Recurrent Neural Networks (RNNs)

- RNNs naturally fit sequence data, e.g.,
 - time series data,
 - text data,
 - speech data...



Applications of RNNs: Machine Translation

机器翻译

The image shows a machine translation application window. At the top, there are language selection dropdowns for "Chinese - detected" and "English", with a double-headed arrow icon between them. The Chinese text input field contains the sentence "机器翻译让沟通交流变得更容易" with a red cursor at the end of "容易". Below the input, the pinyin transcription "Jīqì fānyì ràng gōutōng jiāoliú biàn dé gèng róngyì" is displayed. The English output field contains the sentence "Machine translation makes communication easier". At the bottom of the window, there are two small icons: a speaker icon and a microphone icon on the left, and a speaker icon and a square icon on the right.

Chinese - detected ▾

↔

English ▾

机器翻译让沟通交流变得更
容易 |

Jīqì fānyì ràng gōutōng jiāoliú
biàn dé gèng róngyì

Machine
translation
makes
communication
easier

Speaker Microphone

Speaker Square

Applications of RNNs: Machine Translation



Applications of RNNs: Speech Recognition

语音识别



amazon alexa



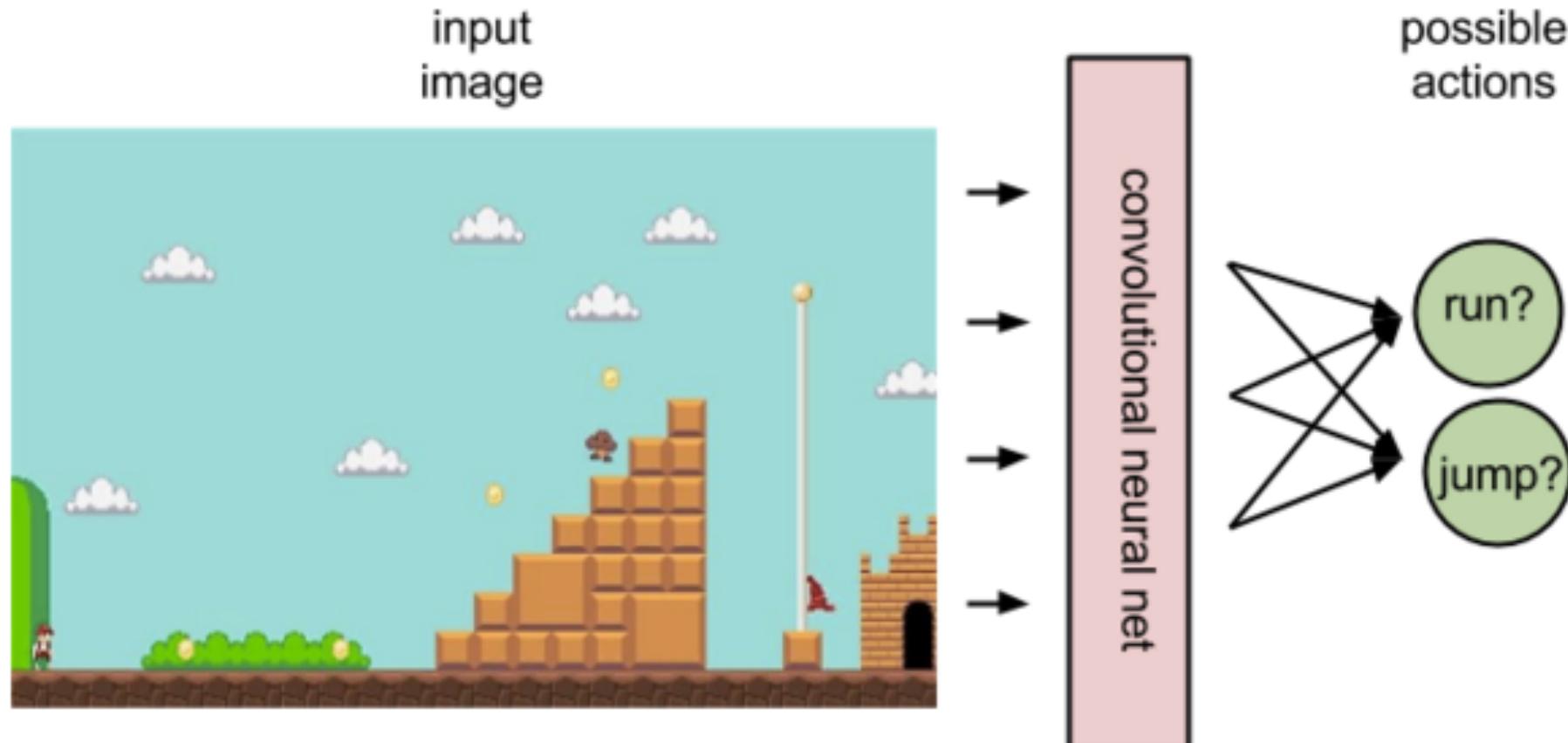
Hey Siri

Deep Reinforcement Learning (DFL)

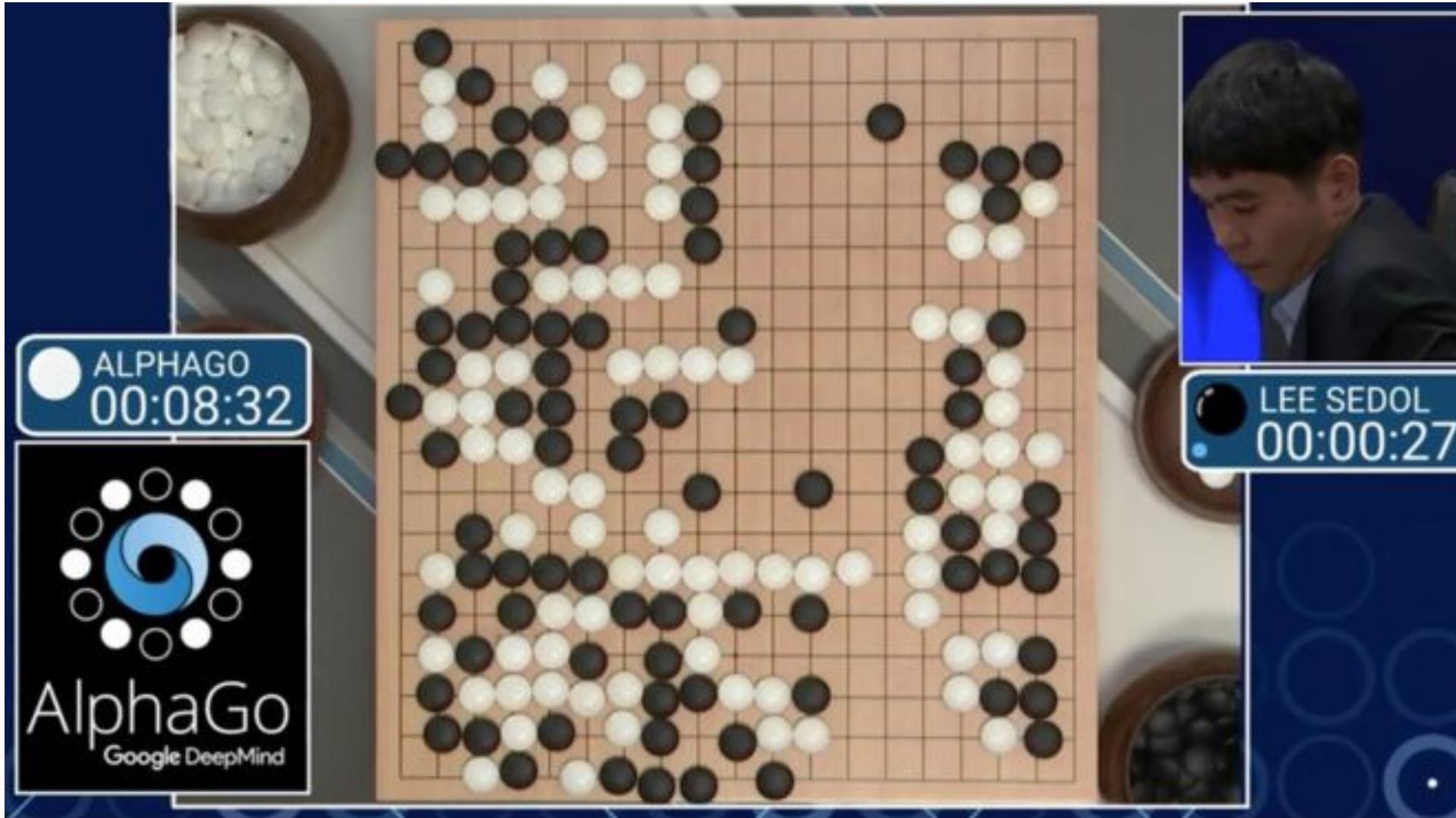
深度强化学习

- DFL has applications in robotics, video game, and finance.

深度 强化学习 在机器人技术、电子游戏和金融等方面都有应用。



Applications of DRL: Games



Applications of DRL: Robotics

Control Theory v.s. DRL



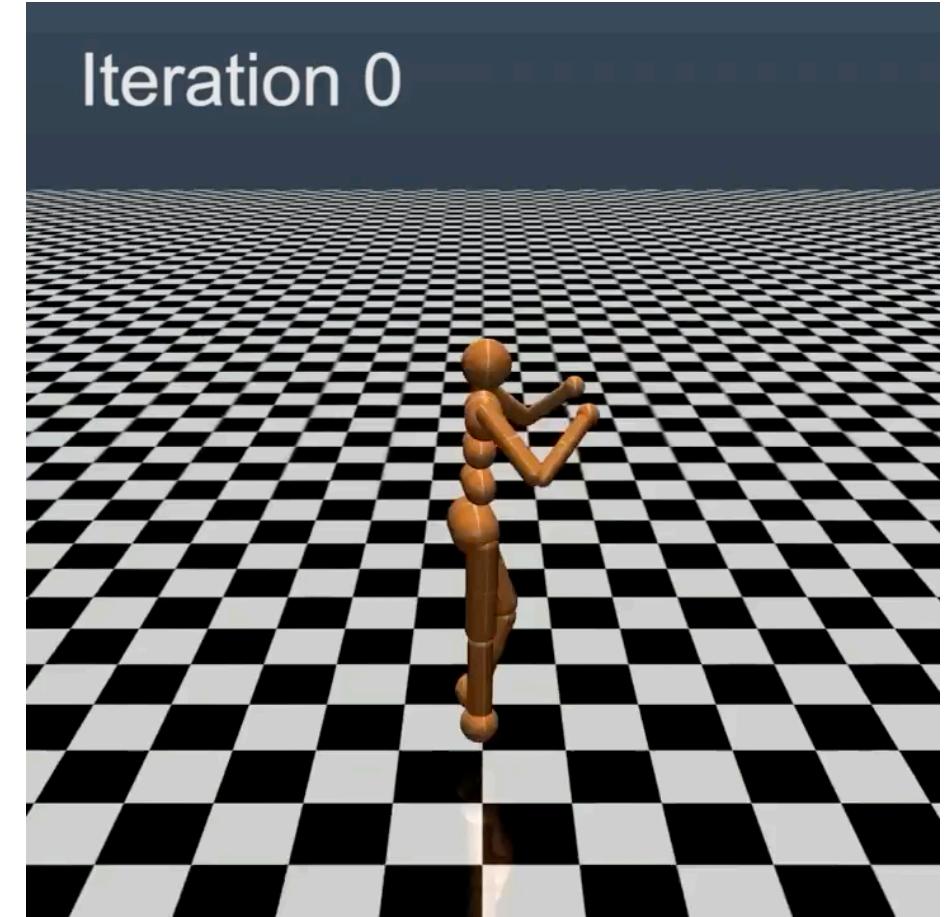
Boston Dynamics' Atlas

Applications of DRL: Robotics

Control Theory v.s. **DRL**



Iteration 0



What we have learned so far...

- **ML tasks**: regression, classification, ...
- **ML models**: linear models, CNNs, RNNs, ...

Example: least squares regression model

- Loss function: $L(\mathbf{w}) = \frac{1}{n} \sum_{i=1}^n (\mathbf{x}_i^T \mathbf{w} - y_i)^2$.
- Optimization: $\mathbf{w}^* = \min_{\mathbf{w}} L(\mathbf{w})$.

- How to solve the model?

Computations

Computational Methods

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- **Computations:** solve the model using numerical algorithms, e.g., gradient descent (GD) or stochastic descent (SGD).

Gradient Descent

Example: least squares regression model

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Gradient: $\frac{\partial L}{\partial \mathbf{w}}$

- \mathbf{w} is a d -dimensional vector.
- $L(\mathbf{w})$ is a scalar.
- Thus $\frac{\partial L}{\partial \mathbf{w}}$ is a d -dimensional vector.

Gradient Descent

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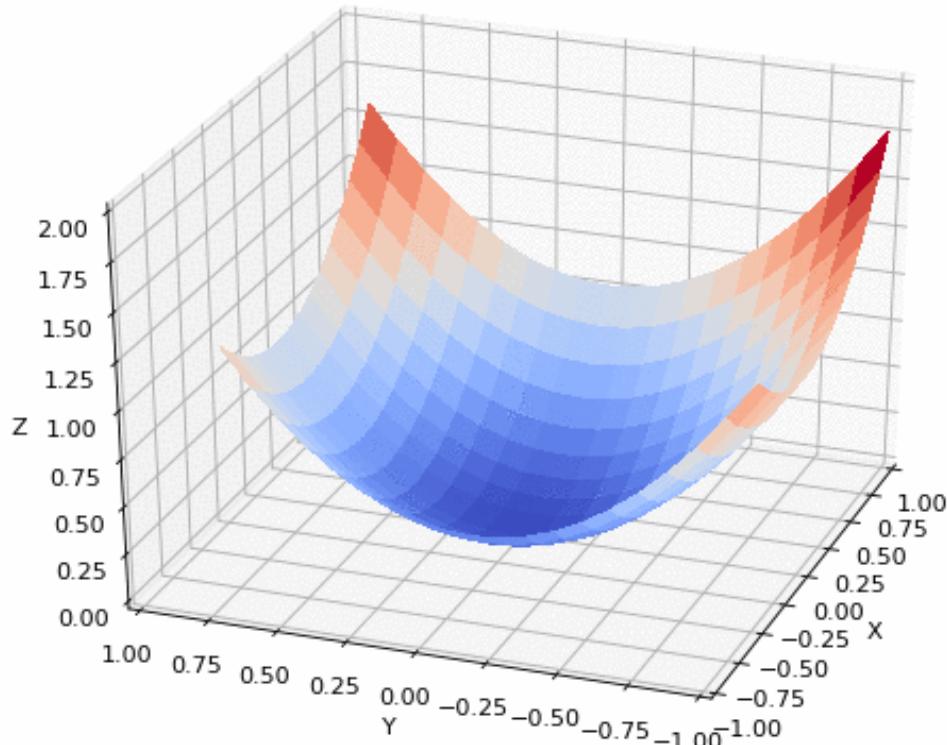
Gradient descent algorithm

- Randomly initialize \mathbf{w}_0 .
- For $t = 0$ to T :
 - Gradient at \mathbf{w}_t : $\mathbf{g}_t = \frac{\partial L}{\partial \mathbf{w}} \Big|_{\mathbf{w}_t};$
 - $\mathbf{w}_{t+1} = \mathbf{w}_t - \alpha \mathbf{g}_t.$

Gradient Descent

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Variants of Gradient Descent

- Stochastic gradient descent (SGD).
- SGD with momentum.
- RMSProp.
- ADAM...

Computational Challenges

- **Big data**: too many training samples.
 - ImageNet: **14 million** 256×256 images.
- **Big model**: too many model parameters.
 - ResNet-50 (a very popular CNN architecture) has **25 million parameters**.
2500万个参数

Computational Challenges

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- **Big data + big model** bring computational challenges.
- Training ResNet-50 on ImageNet using a **single GPU** takes around **14 days**.

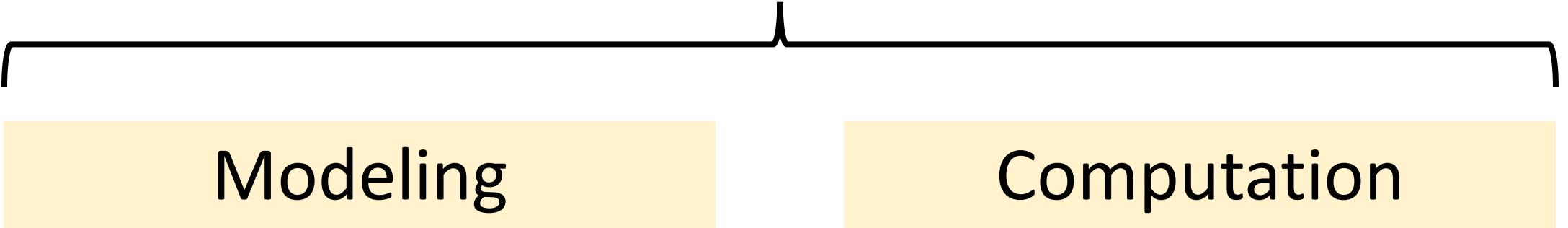
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Efficient algorithms and software systems are necessary.

Machine Learning in Practice



A horizontal line with a bracket underneath it spans the width of the slide. Inside the bracket, there are two yellow rectangular boxes. The left box contains the word 'Modeling' and the right box contains the word 'Computation'.
Modeling

Computation

Machine Learning in Practice

Modeling

- The model that fits the data and problem.
- Decide the network structures, activation functions, loss functions, etc.
- Improve the prediction accuracy.
 - Experience in ML models.
 - Understanding of the problem and data.

Computation

适合数据和问题的模型。

决定网络结构、激活功能、损失功能等

提高预测精度。

在ML模型方面的经验。

了解问题和数据

Machine Learning in Practice

Modeling

Computation

- Design or apply efficient algorithms. 设计或应用有效的算法
- Implement the algorithm using systems like TensorFlow, Hadoop, etc.

Machine Learning in Practice

Modeling

Computation

- Design or apply efficient algorithms.
- Implement the algorithm using systems like TensorFlow, Hadoop, etc.
- Optimize your code.
 - Experience in algorithms.
 - Experience in systems.