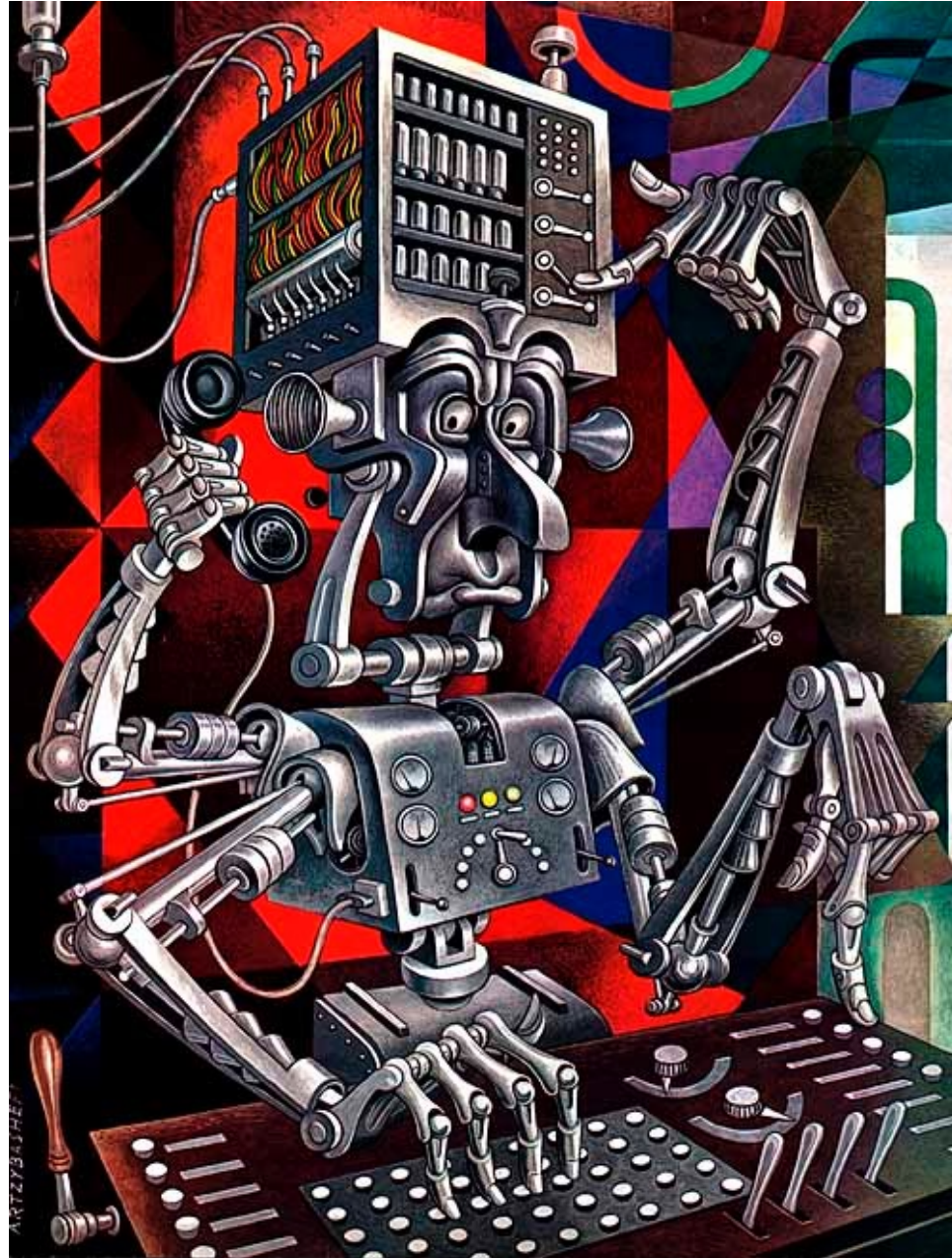


# Machine learning: An overview of the landscape



# Lecture overview

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- The statistical learning viewpoint
- Supervised classification
- Beyond supervised classification: A taxonomy of prediction problems and types of supervision

# How can we build an agent to...

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Play chess?



Translate between languages?



Recognize object categories?

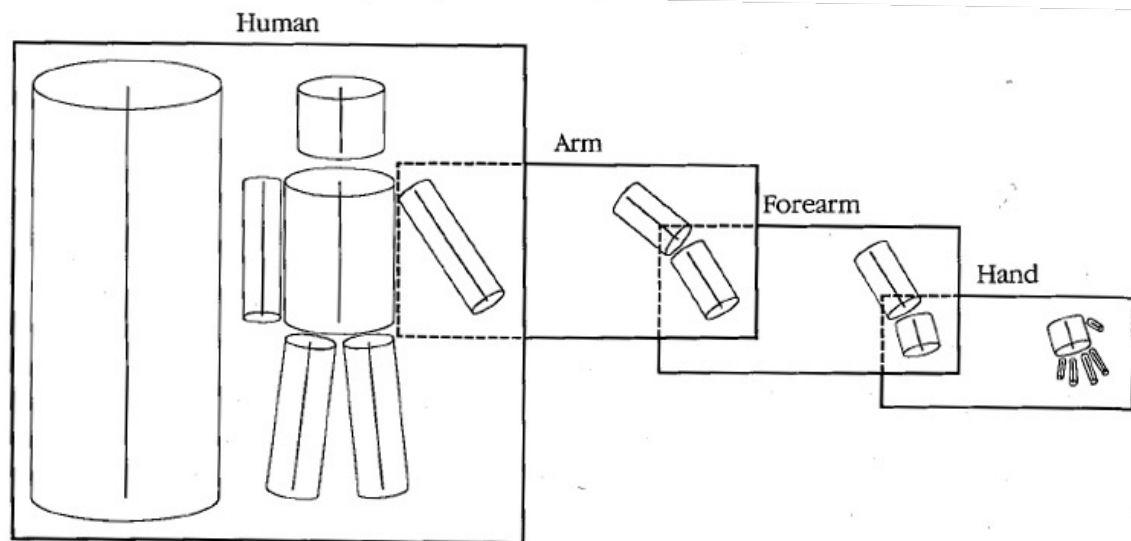


Fly a drone?

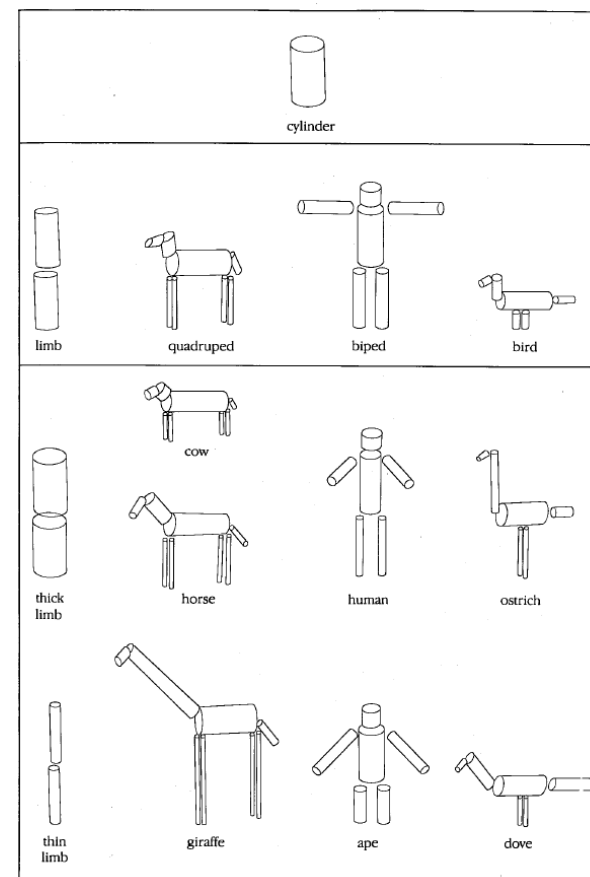


# How can we build an agent to achieve expertise?

- Good old-fashioned AI (GOF AI) answer:  
Program expertise into the agent
  - Never worked (in general)...



Figures from Marr's Vision (1982)



# How can we build an agent to achieve expertise?

---

- Good old-fashioned AI (GOF AI) answer:  
Program expertise into the agent
  - Never worked (in general)...
- Modern answer: Program into the agent the *ability to improve performance based on experience*
  - Experience should come from *training data* or *demonstrations*
  - Learning is optimizing performance of the agent on the training data, with the hope that it will *generalize* to unseen inputs
  - This is the *statistical learning* viewpoint

# Example: Image classification

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Input

Desired output



apple

pear

tomato

cow

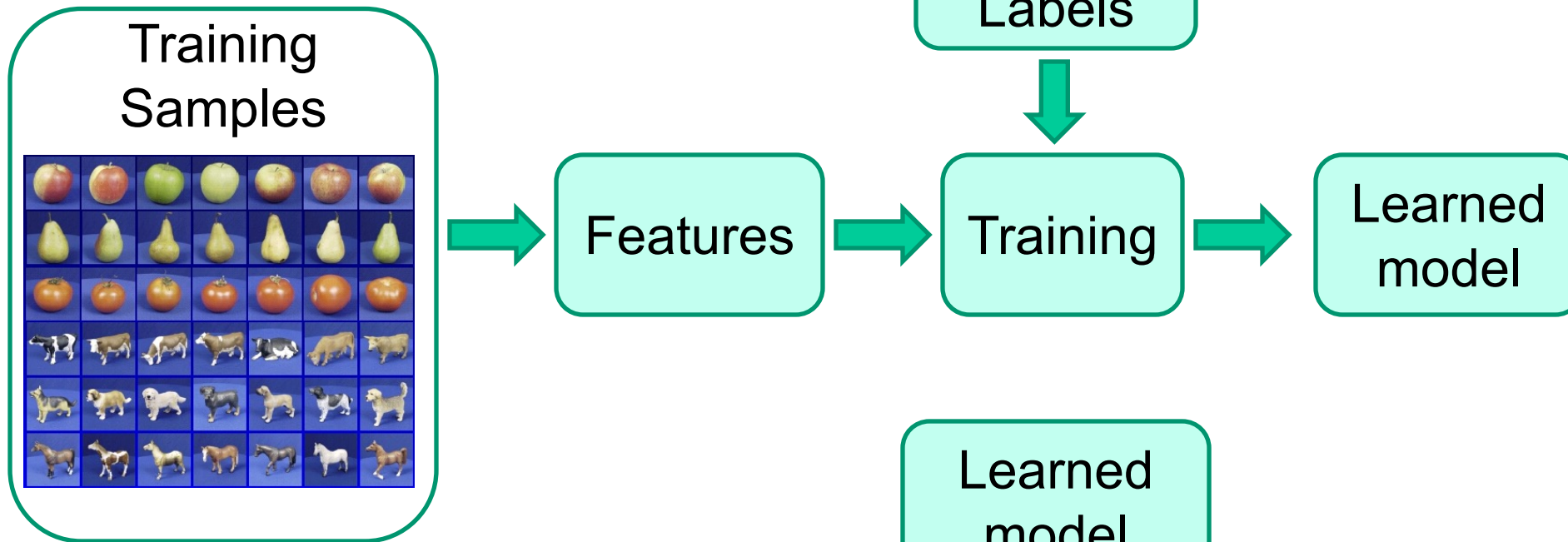
dog

horse

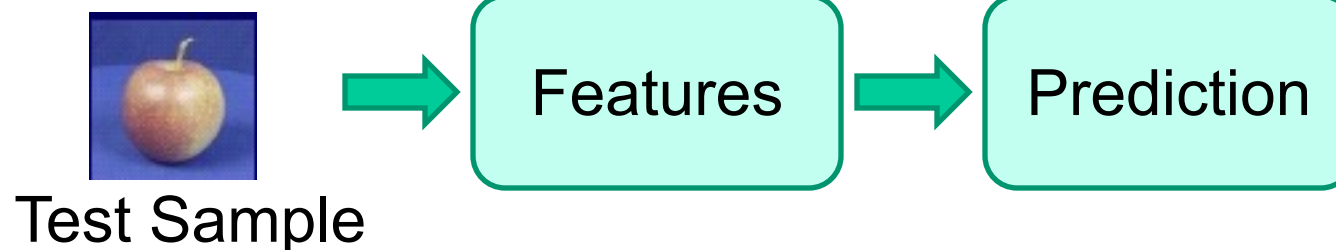
# Example: Image classification

---

## Training time



## Testing time



# The basic *supervised learning* framework

---

$$y = f(x)$$

output      prediction function      input

- **Training** (or **learning**): given a *training set* of labeled examples  $\{(x_1, y_1), \dots, (x_N, y_N)\}$ , instantiate a predictor  $f$
- **Testing** (or **inference**): apply  $f$  to a new *test example*  $x$  and output the predicted value  $y = f(x)$



# More supervised learning examples: Text classification

---

## Spam classification



Ok, I know this is blatantly OT but I'm beginning to go insane. Had an old Dell Dimension XPS sitting in the corner and decided to put it to use, I know it was working pre being stuck in the corner, but when I plugged it in, hit the power nothing happened.



Dear Sir.

First, I must solicit your confidence in this transaction, this is by virtue of its nature as being utterly confidential and top secret. ...

## Sentiment classification

"I love this movie.  
I've seen it many times  
and it's still awesome."

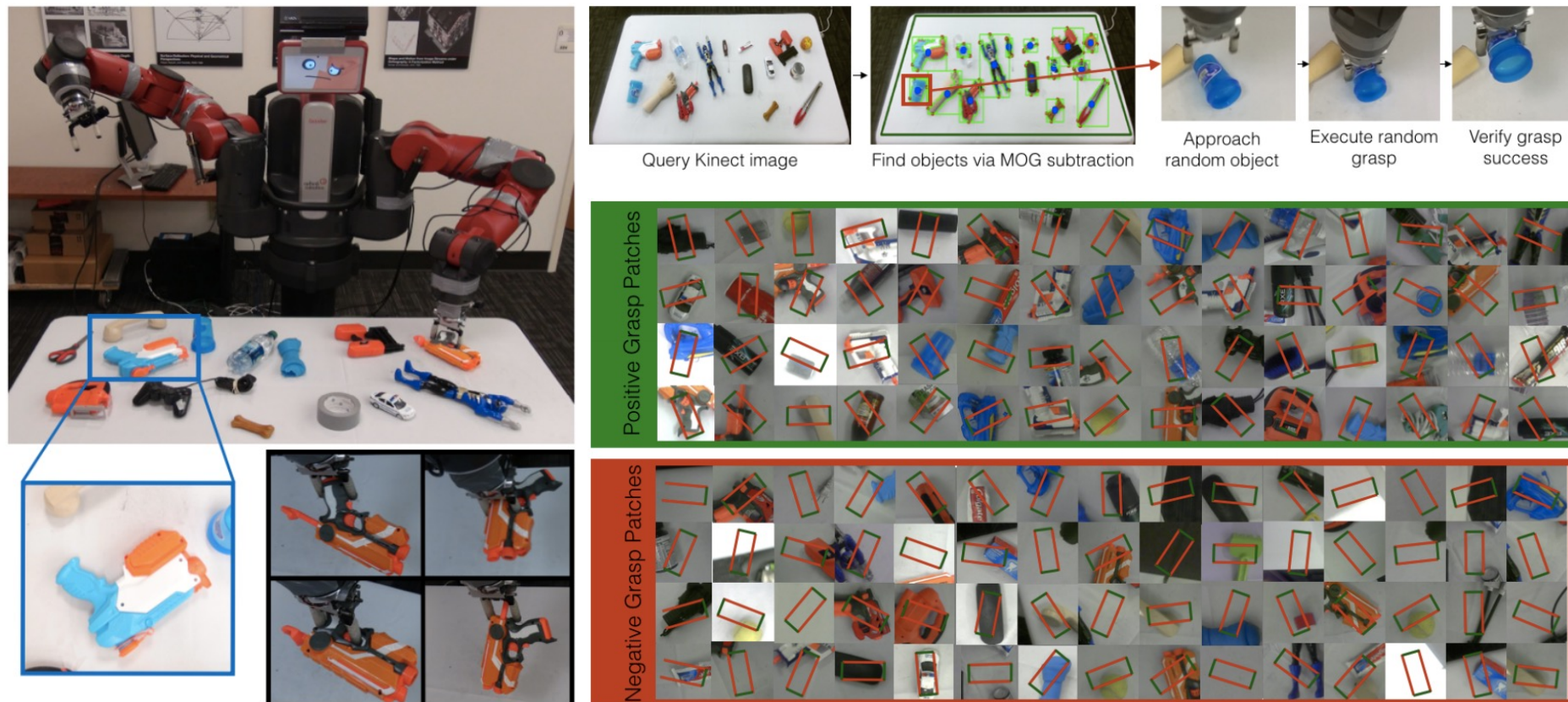


"This movie is bad.  
I don't like it at all.  
It's terrible."



[Image source](#)

# Another example: Grasp classification



L. Pinto and A. Gupta. [Supersizing self-supervision: Learning to grasp from 50K tries and 700 robot hours](#). ICRA 2016

[YouTube video](#)

# Lecture overview

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- The statistical learning viewpoint
- Supervised classification
- **Beyond supervised classification: A brief taxonomy**

# Beyond classification: Regression

Date prediction



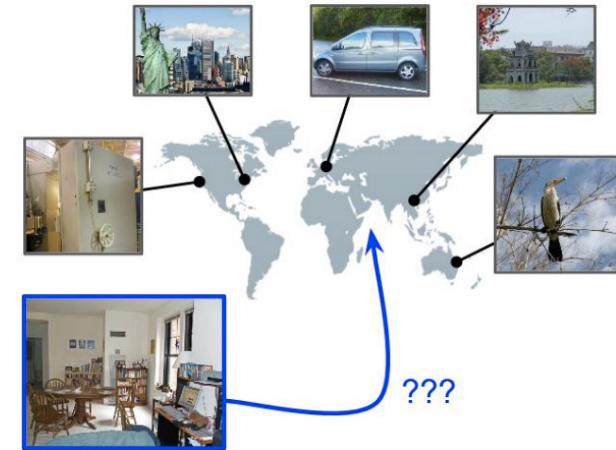
[Vittayakorn et al. \(2017\)](#)

Image colorization



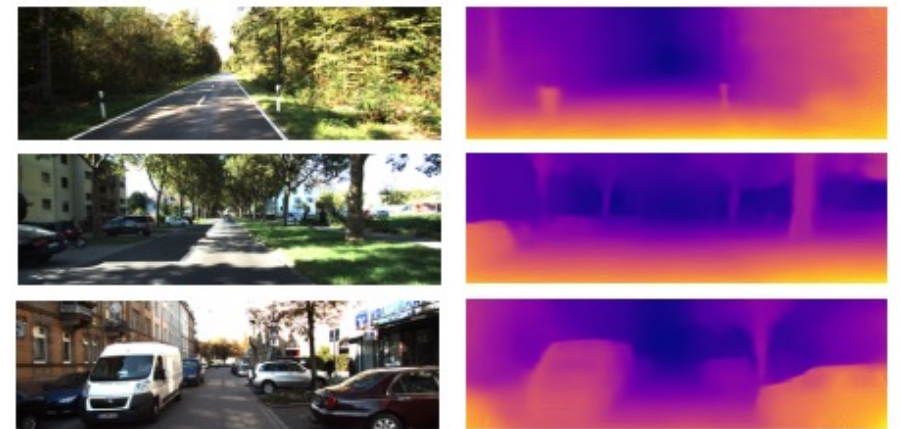
[Zhang et al. \(2016\)](#)

Location prediction



[Vo et al. \(2017\)](#)

Depth prediction



[Wang et al. \(2017\)](#)

# Beyond classification: Structured prediction



Image

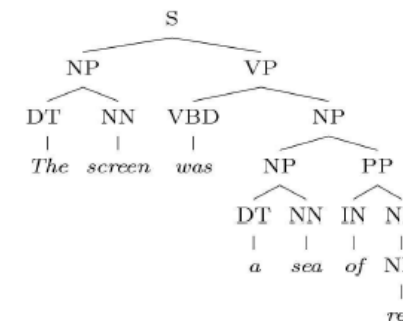


**brace**

Word

The screen was  
a sea of red

Sentence



Parse tree

RSCCPCYWGGCPW  
GQNCYPEGCSGPKV

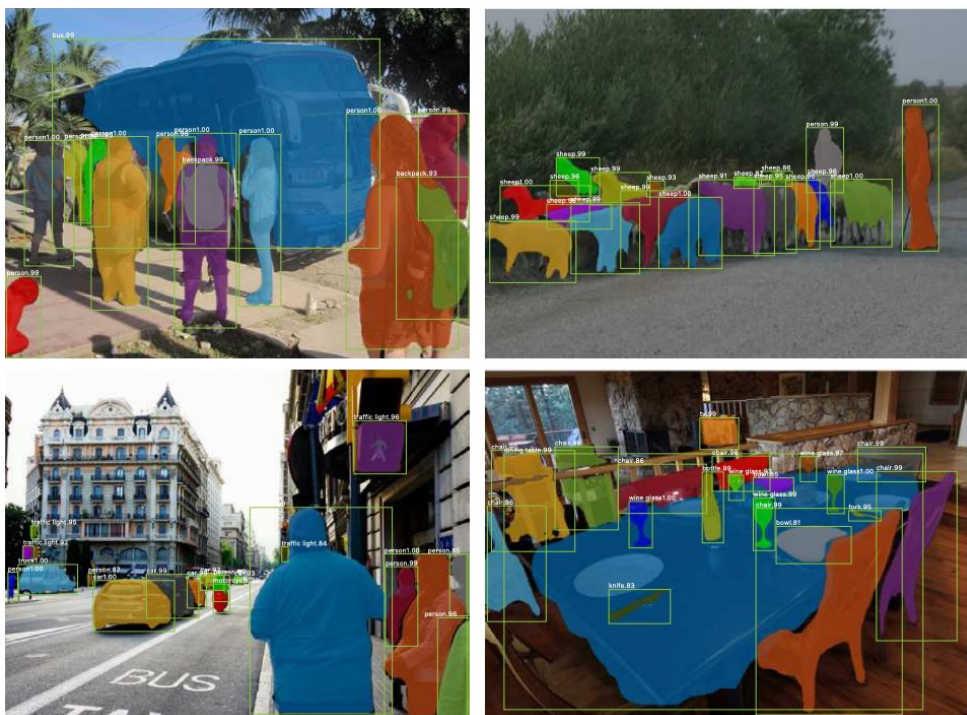
Amino-acid sequence



Bond structure

# Structured and dense prediction for scene understanding

Bounding box prediction,  
dense prediction



Keypoint prediction



# Structured and dense prediction for scene understanding

## Image captioning



"man in black shirt is playing guitar."



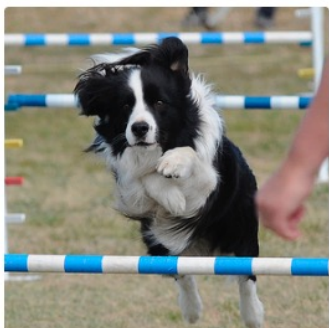
"construction worker in orange safety vest is working on road."



"two young girls are playing with lego toy."



"girl in pink dress is jumping in air."

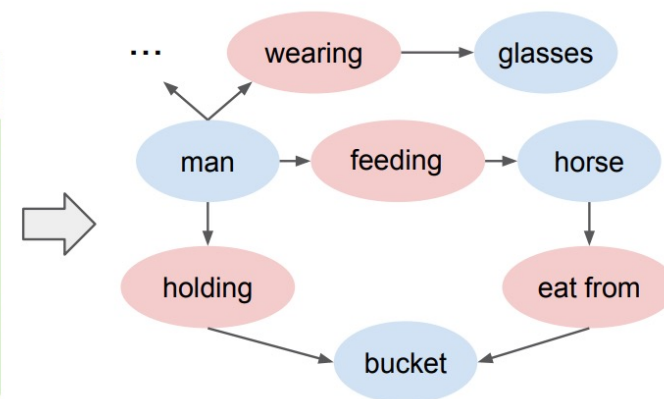


"black and white dog jumps over bar."



"young girl in pink shirt is swinging on swing."

## Scene graph generation



A. Karpathy, L. Fei-Fei. [Deep Visual-Semantic Alignments for Generating Image Descriptions](#). CVPR 2015

D. Xu, Y. Zhu, C. Choy, and L. Fei-Fei. [Scene Graph Generation by Iterative Message Passing](#). CVPR 2017

# Beyond classification and supervised learning

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- Other prediction scenarios (output types)
  - Regression
  - Structured prediction
  - Dense prediction
- Other supervision scenarios
  - Unsupervised learning
  - Self-supervised or predictive learning
  - Reinforcement learning
  - Active learning
  - Lifelong learning



# Unsupervised learning

- **Clustering**

- Discover groups of “similar” data points

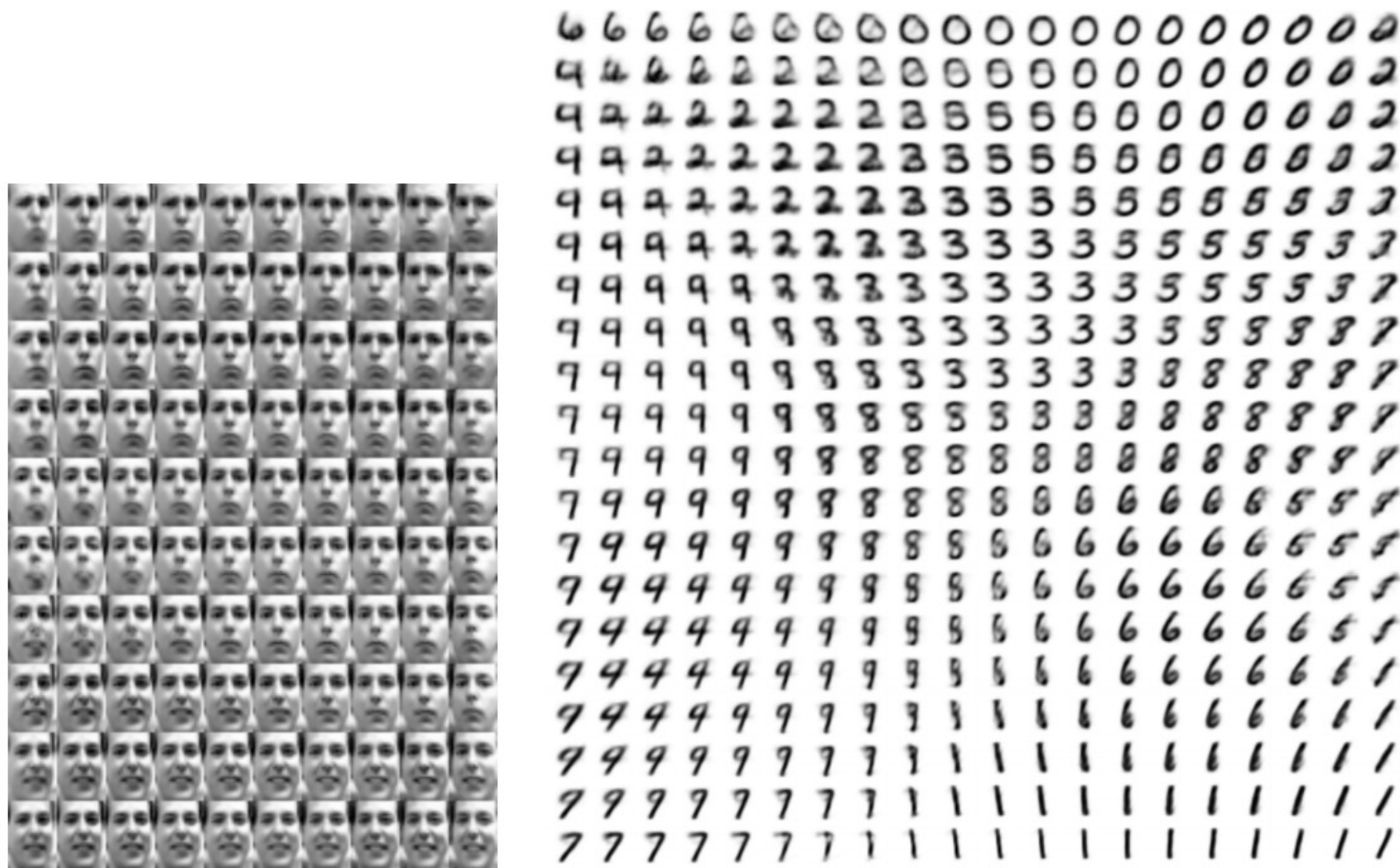


Y. Gong, Q. Ke, M. Isard, and S. Lazebnik. [A Multi-View Embedding Space for Modeling Internet Images, Tags, and Their Semantics](#). IJCV 2014

# Unsupervised learning

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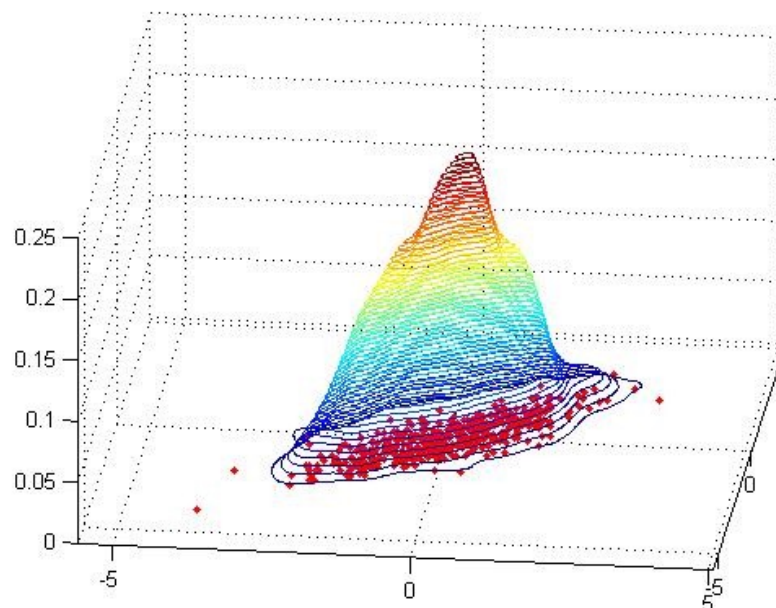
- **Dimensionality reduction, manifold learning**
  - Discover a lower-dimensional surface on which the data lives



# Unsupervised learning

---

- Learning the data distribution
  - **Density estimation:** Find a function that approximates the probability density of the data (i.e., value of the function is high for “typical” points and low for “atypical” points)
  - An extremely hard problem for high-dimensional data...

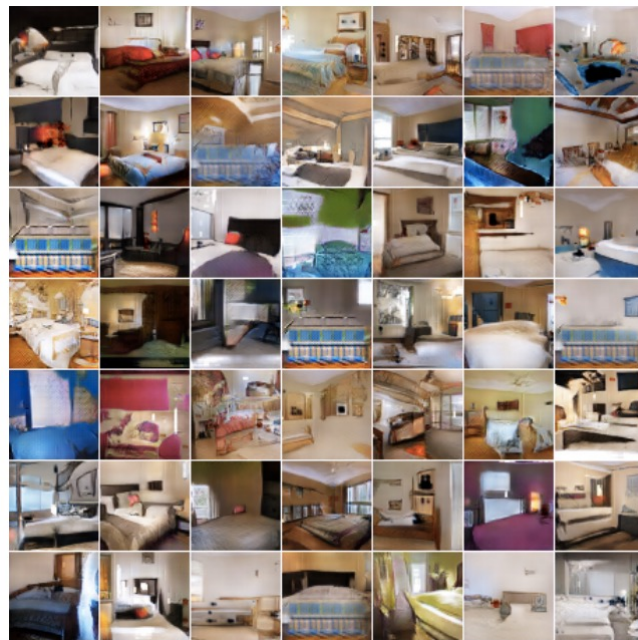


# Unsupervised learning

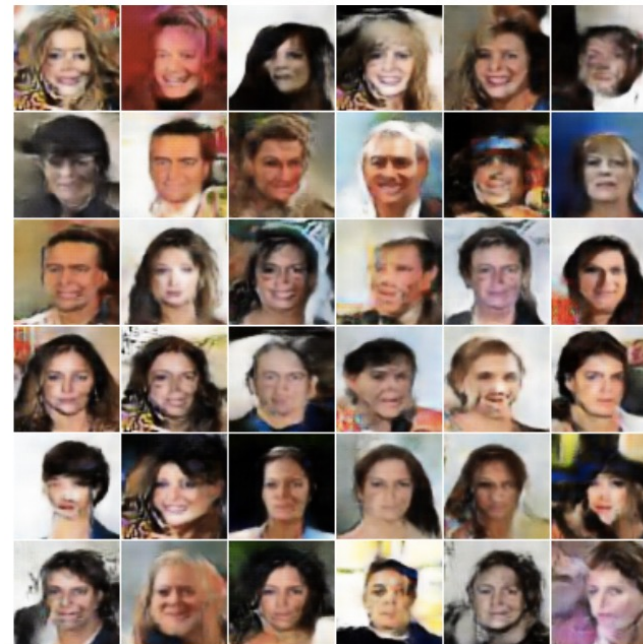
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- Learning the data distribution
  - **Learning to sample:** Produce samples from a data distribution that mimics the training set

## Generative adversarial networks



“Bedroom”  
(circa 2015)



“Face”  
(circa 2015)

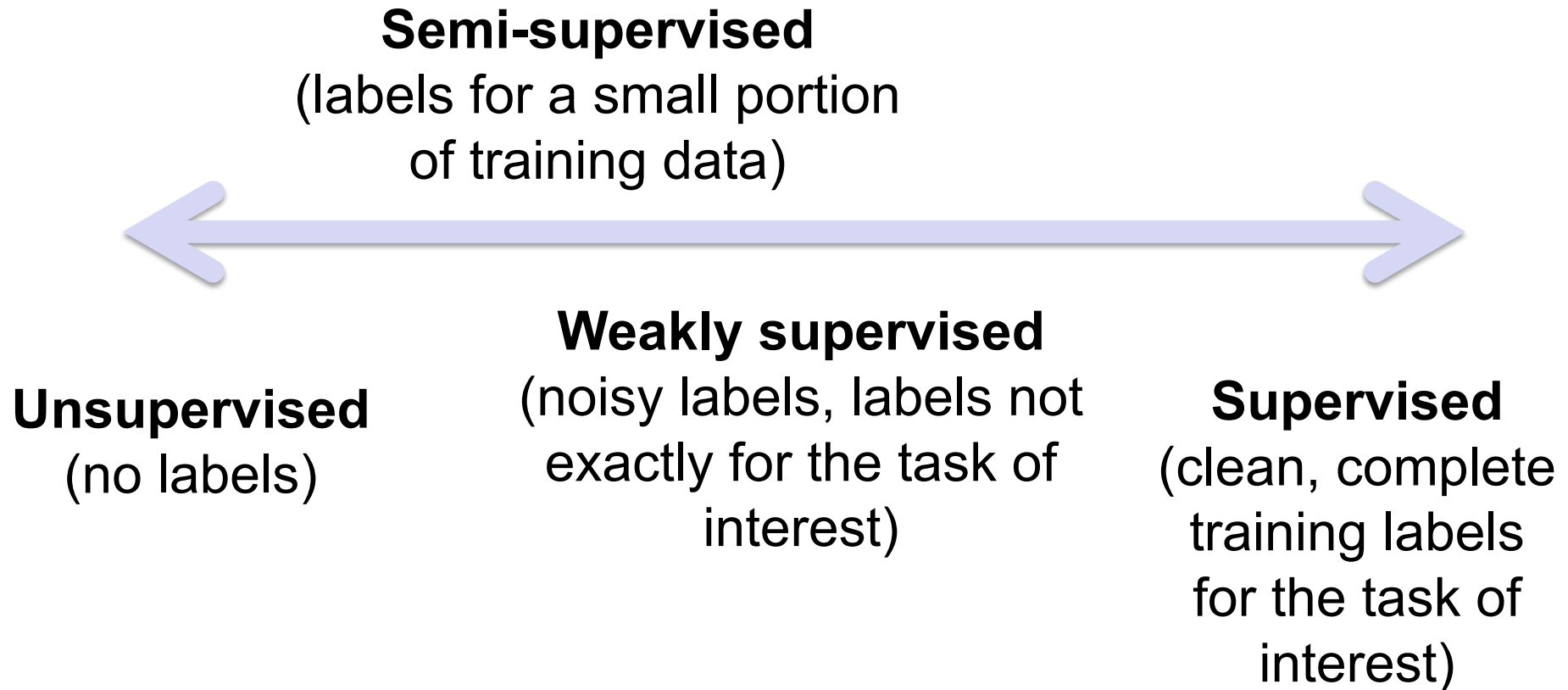
# Beyond classification and supervised learning

---

- Other prediction scenarios (output types)
  - Regression
  - Structured prediction
  - Dense prediction
- Other supervision scenarios
  - Unsupervised learning
    - Clustering and quantization
    - Dimensionality reduction, manifold learning
    - Density estimation
    - Learning to sample

# Between “unsupervised” and “fully supervised”

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# Beyond classification and supervised learning

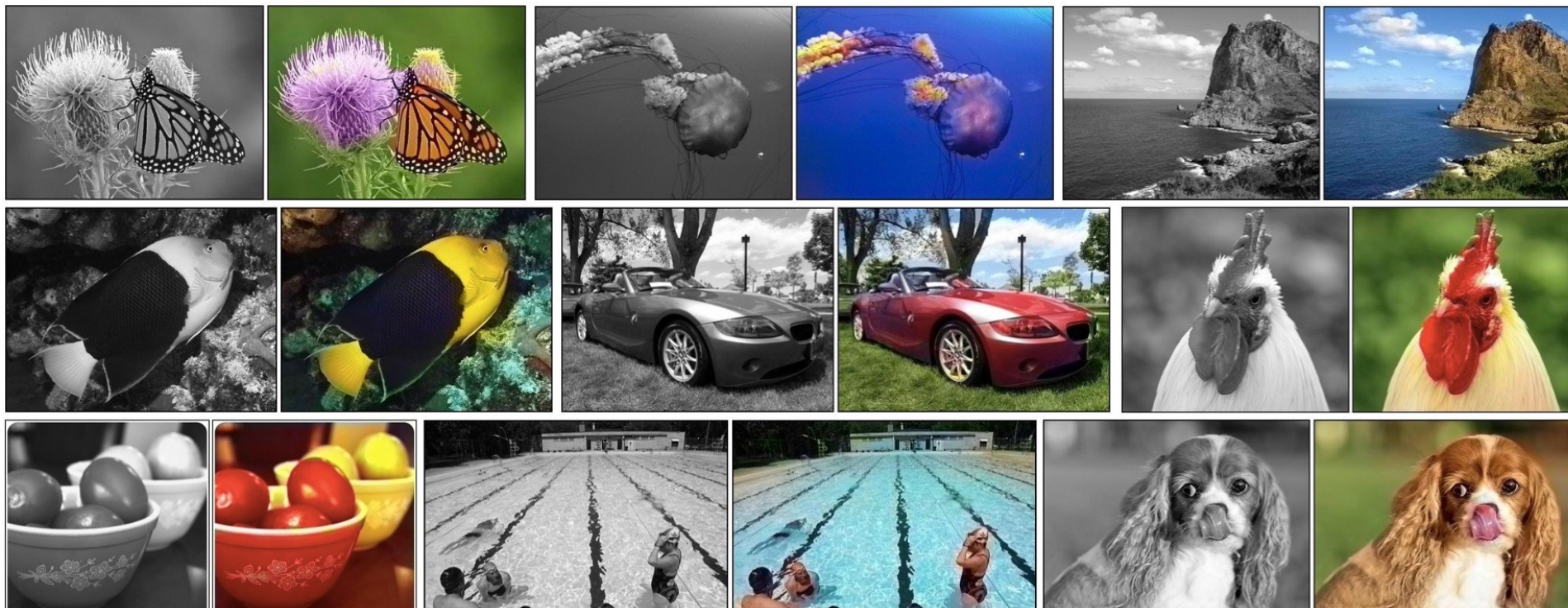
---

- Other prediction scenarios (output types)
  - Regression
  - Structured prediction
  - Dense prediction
- Other supervision scenarios
  - Unsupervised learning
  - Self-supervised or predictive learning
  - Reinforcement learning
  - Active learning
  - Lifelong learning

# Self-supervised or predictive learning

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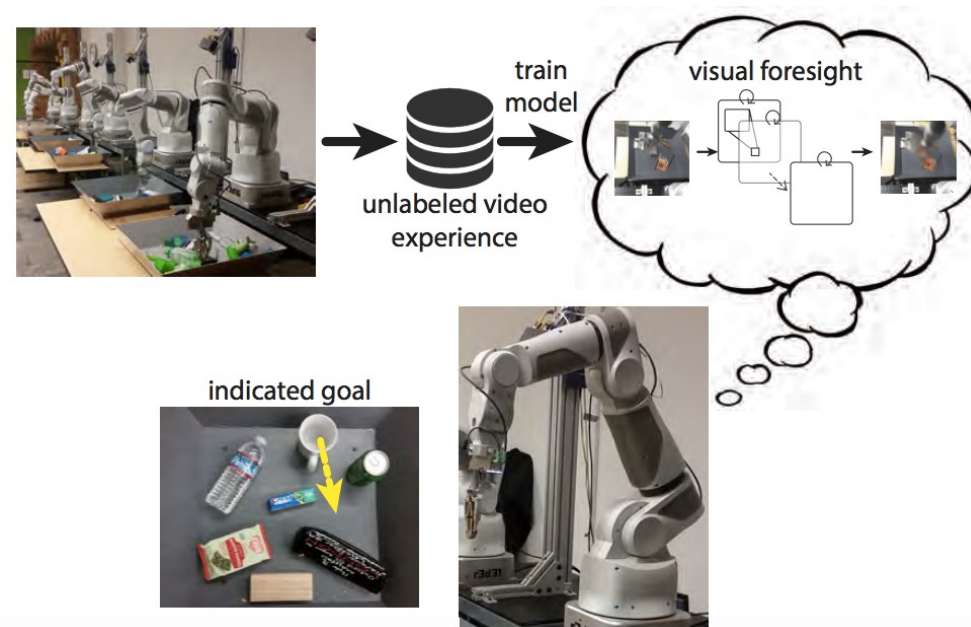
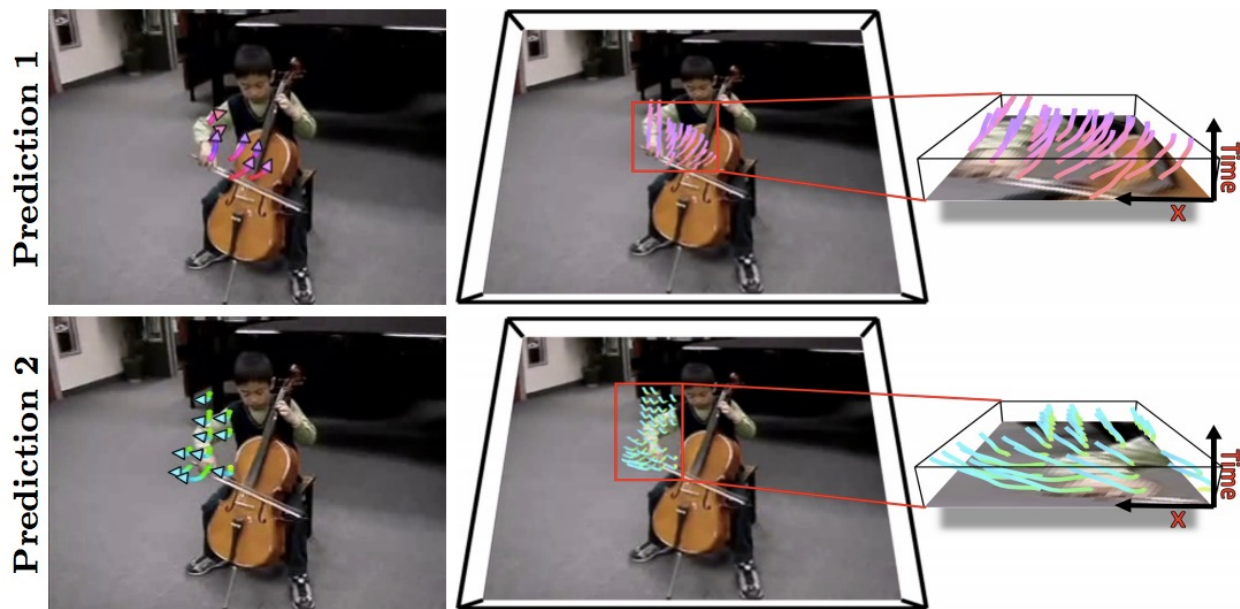
- Use part of the data to predict other parts of the data
  - Example: Image colorization





# Self-supervised or predictive learning

- Use part of the data to predict other parts of the data
  - Example: Future prediction



J. Walker et al. [An Uncertain Future: Forecasting from Static Images Using Variational Autoencoders](#). ECCV 2016

C. Finn and S. Levine. [Deep Visual Foresight for Planning Robot Motion](#). ICRA 2017. [YouTube video](#)

# Reinforcement learning

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- Learn from rewards in a *sequential* environment



Initial gait

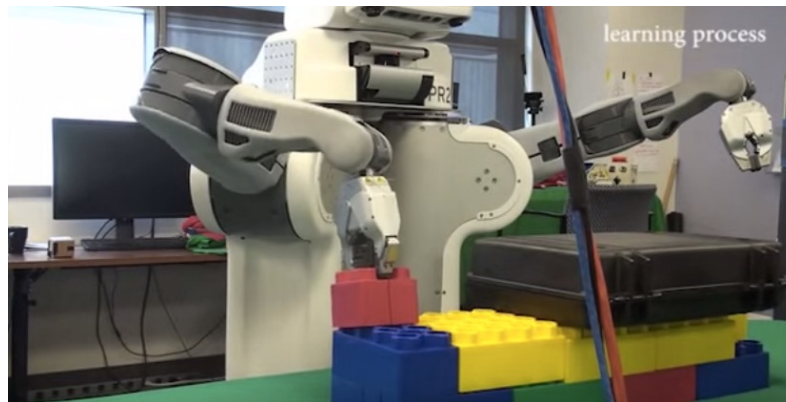


Learned gait

# Reinforcement learning

Learn from rewards in a *sequential* environment

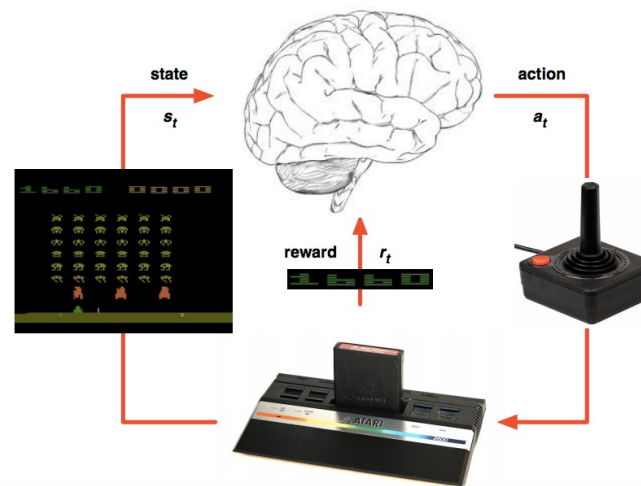
## Sensorimotor learning



## DeepMind's AlphaGo

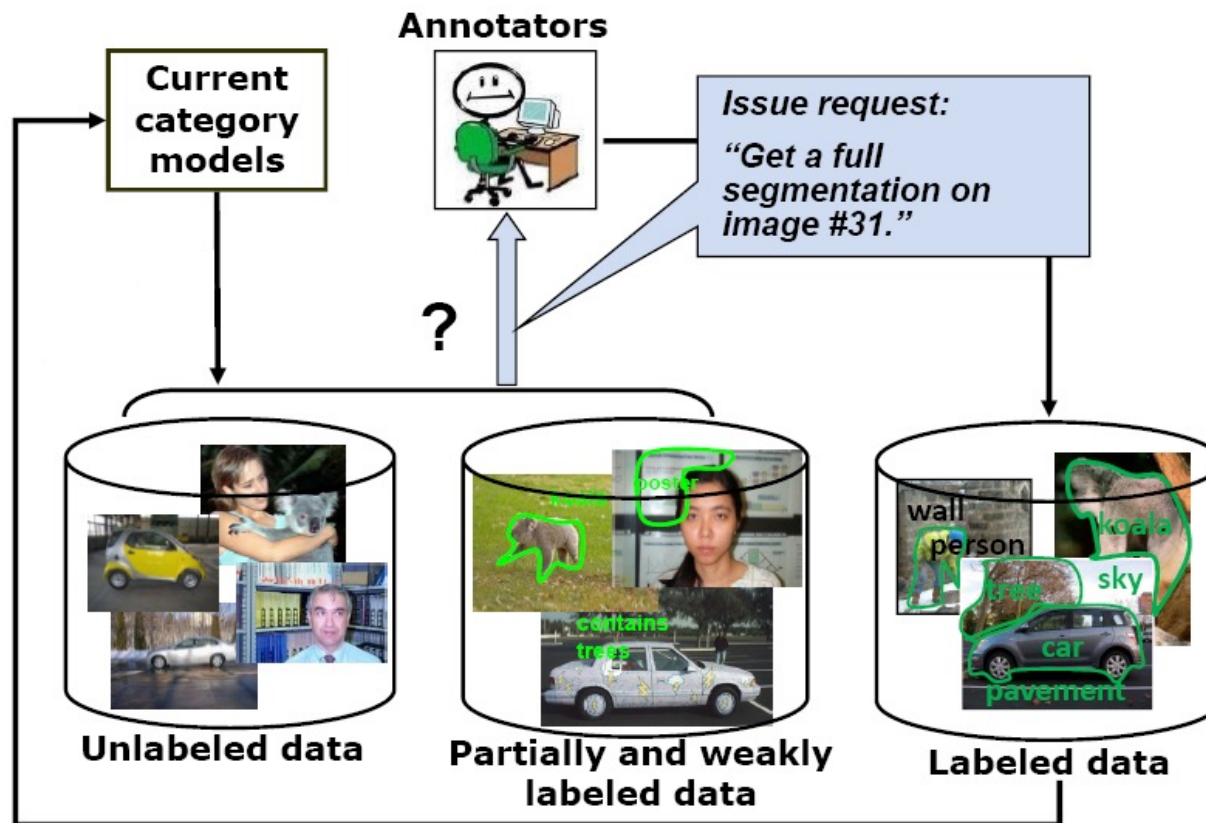


## DeepMind's Atari system



# Active learning

- The learning algorithm can choose its own training examples, or ask a “teacher” for an answer on selected inputs



# Lifelong or continual learning

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Figure 1: **Wanderlust**: Imagine an embodied agent is walking on the street. It may observe new classes and old classes simultaneously. The agent needs to learn fast given only a few samples (red) and recognize the subsequent instances of the class once a label has been provided (green). In this work, we introduce a new online continual object detection benchmark through the eyes of a graduate student to continuously learn emerging tasks in changing environments.

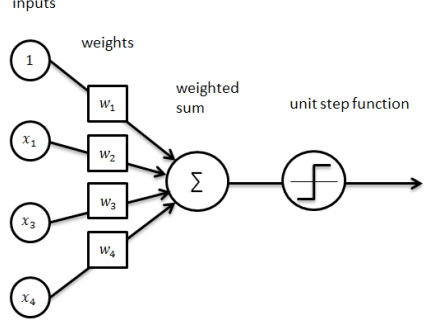
# Beyond classification and supervised learning

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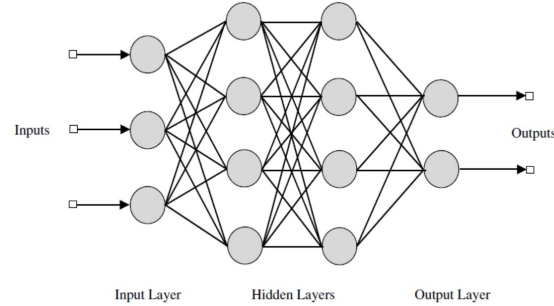
- Other prediction scenarios (output types)
  - Regression
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  - Dense prediction
- Other supervision scenarios
  - Unsupervised learning
  - Self-supervised or predictive learning
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  - Active learning
  - Lifelong learning

# In this class

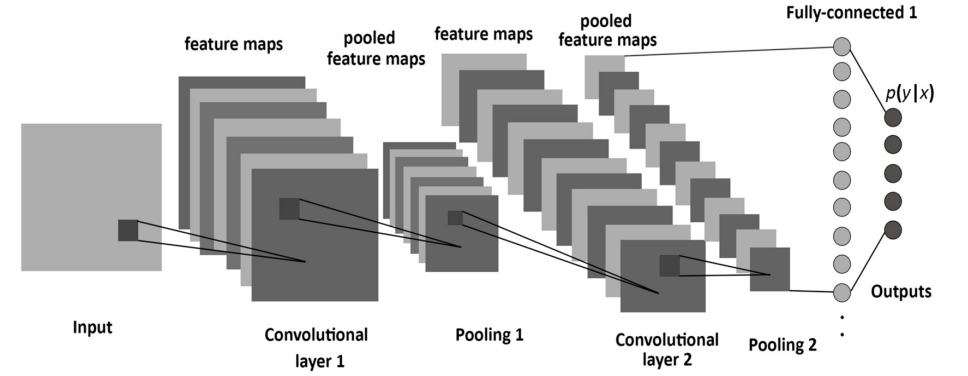
## ML basics, linear classifiers



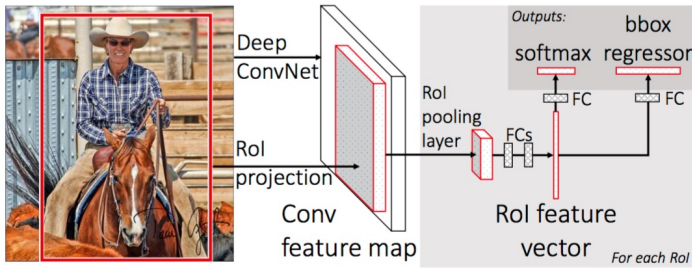
## Multilayer neural networks, backpropagation



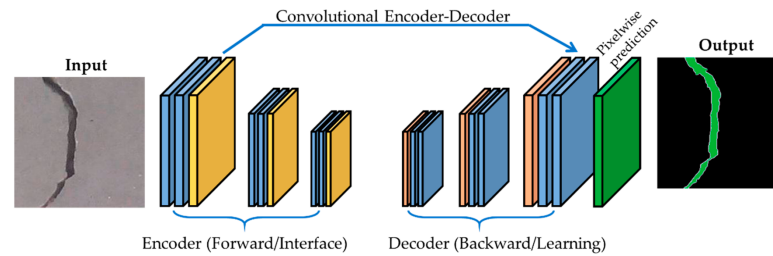
## Convolutional networks for classification



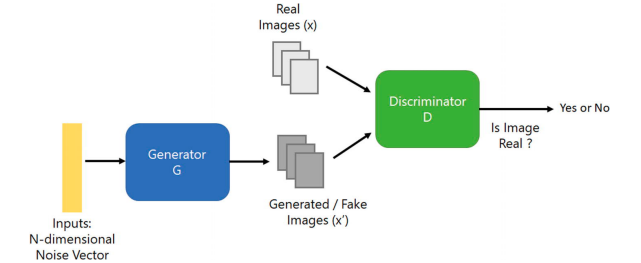
## Networks for detection



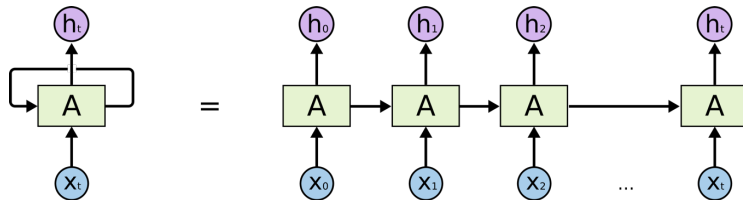
## Networks for dense prediction



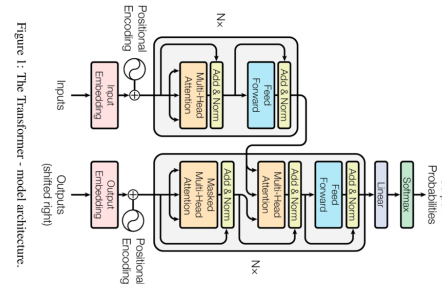
## Generative models (GANs, VAEs)



## Recurrent models



## Transformers



## Deep reinforcement learning

