

Machine Learning Basics

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- Learning Paradigms
- Learning Pipeline

Learning Paradigms

- Supervised learning
- Unsupervised learning
- Reinforcement learning

- Uses **labelled data**, hence “supervised”

Labelled data



Dog



Cat

Labelled data



18 lbs

14 lbs



12 lbs

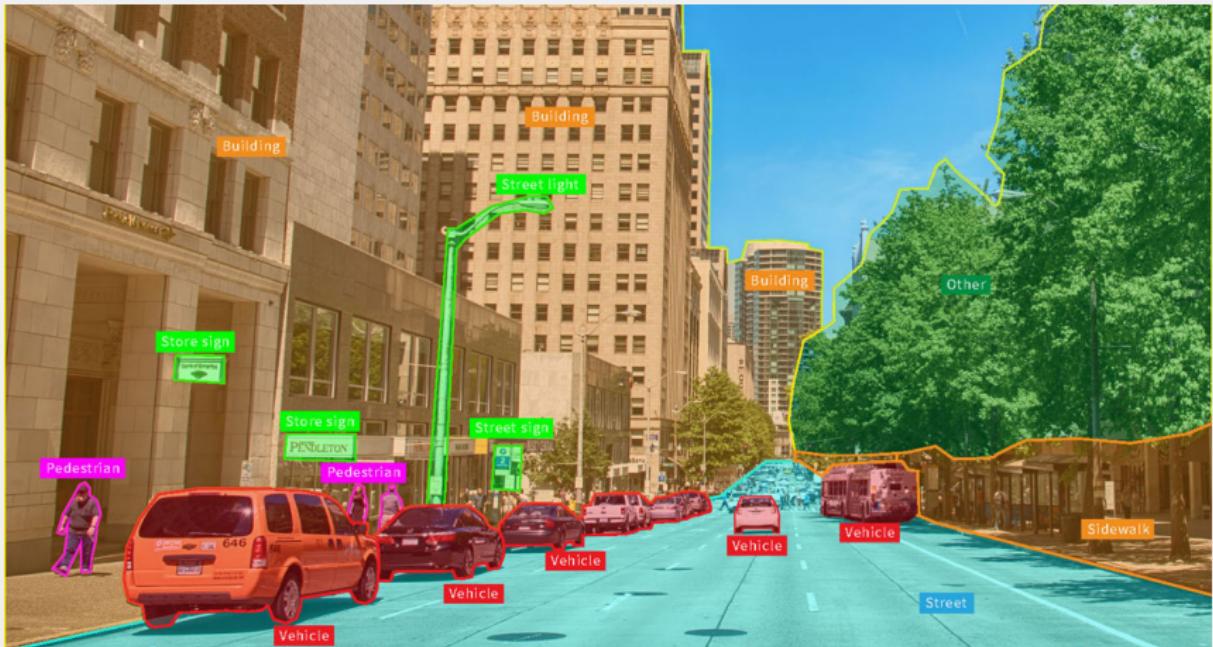
9 lbs

Unlabelled data



Supervised Learning: Labelled Data

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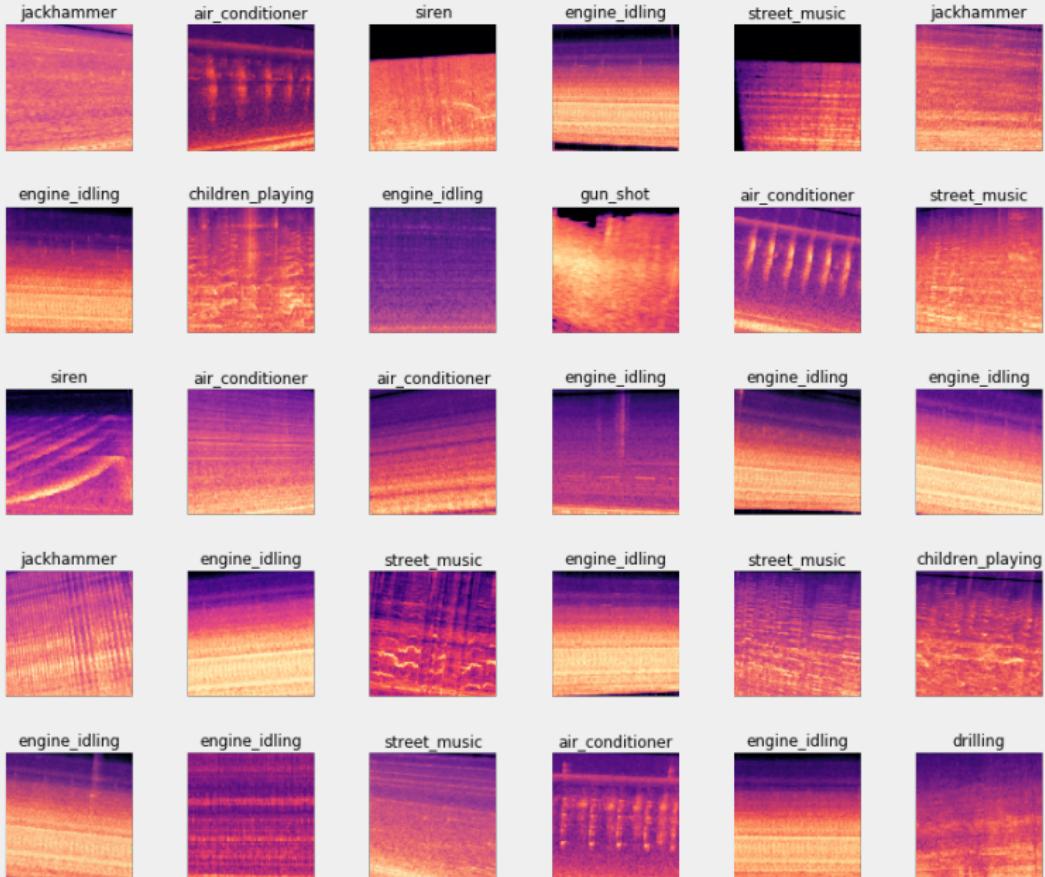
Economic growth has slowed down in recent years .

Das Wirtschaftswachstum hat sich in den letzten Jahren verlangsamt .

Economic growth has slowed down in recent years .

La croissance économique s' est ralentie ces dernières années .

Supervised Learning: Labelled Data



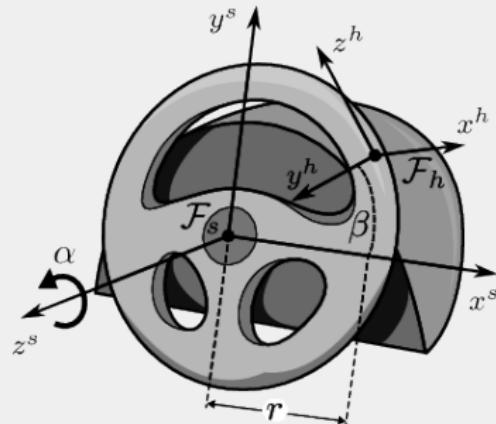
- **Classification:** predict **categorical** labels (discrete values)
- E.g., ImageNet: 1,000 object classes (categories), 1M+ images, labels are crowd-sourced



mite	container ship	motor scooter	leopard
mite	container ship	motor scooter	leopard
black widow	lifeboat	go-kart	jaguar
cockroach	amphibian	moped	cheetah
tick	fireboat	bumper car	snow leopard
starfish	drilling platform	golfcart	Egyptian cat

- **Regression:** predict **numerical** labels (continuous values)
- E.g., stock price, temperature, continuous control commands, etc.

23. 90	+12. 3%	▲	543. 23	120, 000
5. 89	+5. 34%	▲	254. 23	320, 000
6. 34	-7. 89%	▼	321. 56	430, 000
7. 34	+5. 97%	▲	100. 08	120, 000
8. 89	+2. 13%	▲	564. 23	900, 000
45	+6. 43%	▲	765. 90	600, 000
67	-11. 6%	▼	120. 34	380, 000
34	+23. 1%	▲	893. 23	120, 000
9	+5. 56%	▲	128. 98	320, 000
3	-8. 67%	▼	432. 12	75, 000
	+11. 3%	▲	765. 23	15, 000
42	+4. 2%	▲	400. 04	1, 000



- Training set: train a model
- Validation set: tune a model through its hyperparameters
- Test set: test a model
- Example ratio: 6:2:2

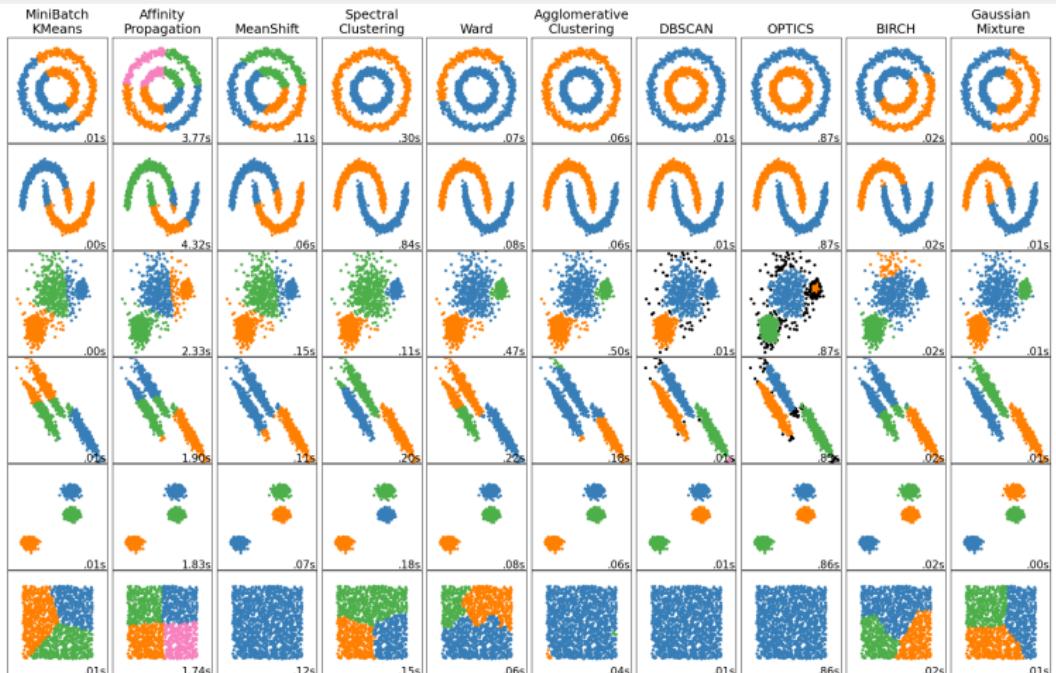
- Assumption 1: data are from the same distribution (usually unknown)
- Assumption 2: training data are i.i.d. (independent and identically distributed)
- Note: non-i.i.d. data may or may not break supervised learning

- Uses **unlabelled data**, hence “unsupervised”
- Tasks: clustering, generative models, ...

Unsupervised Learning: Clustering

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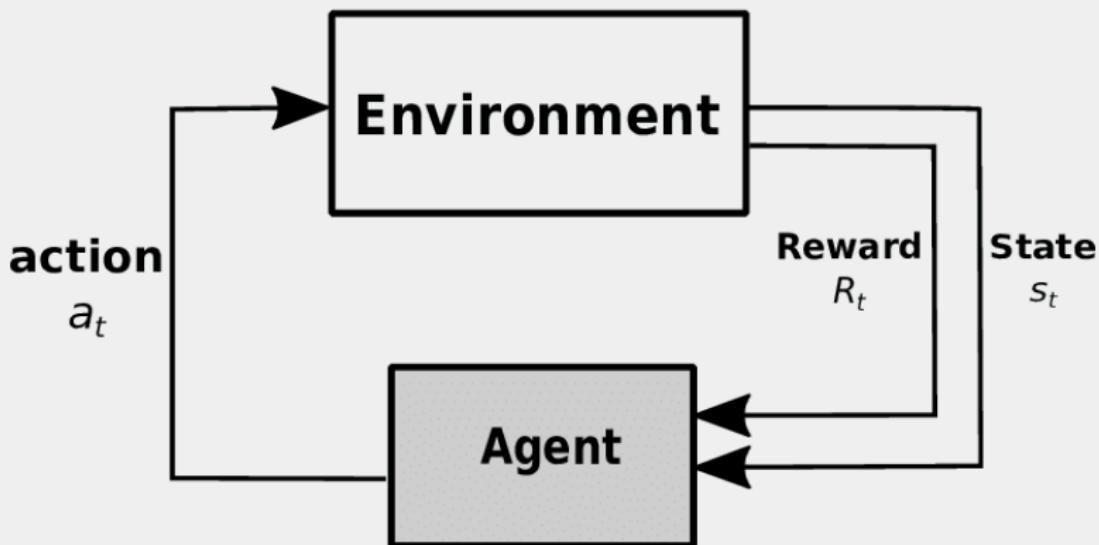
- Goal: discover hidden structures



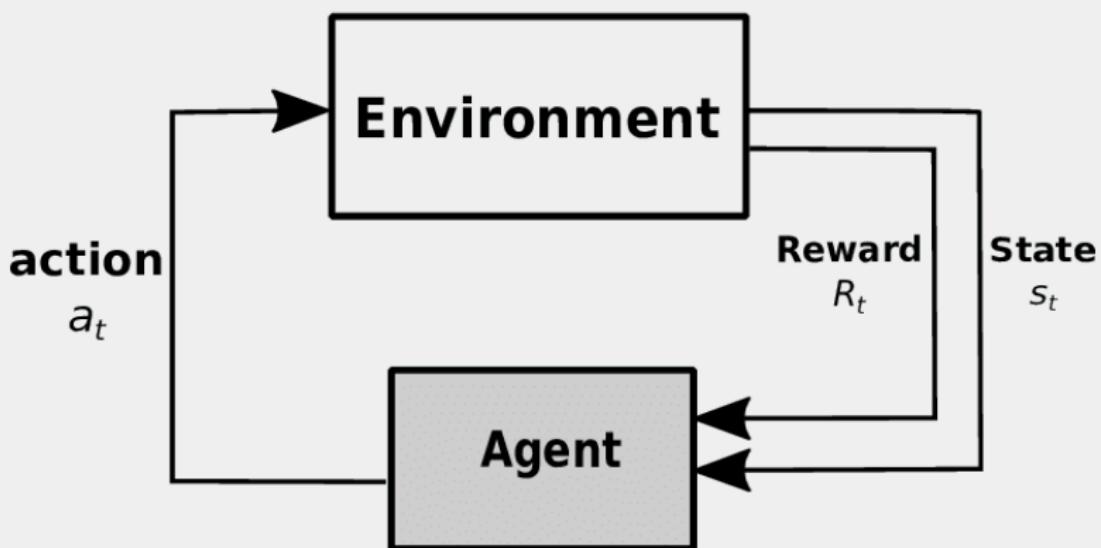
- Goal: generate synthetic data (image, sound, etc)



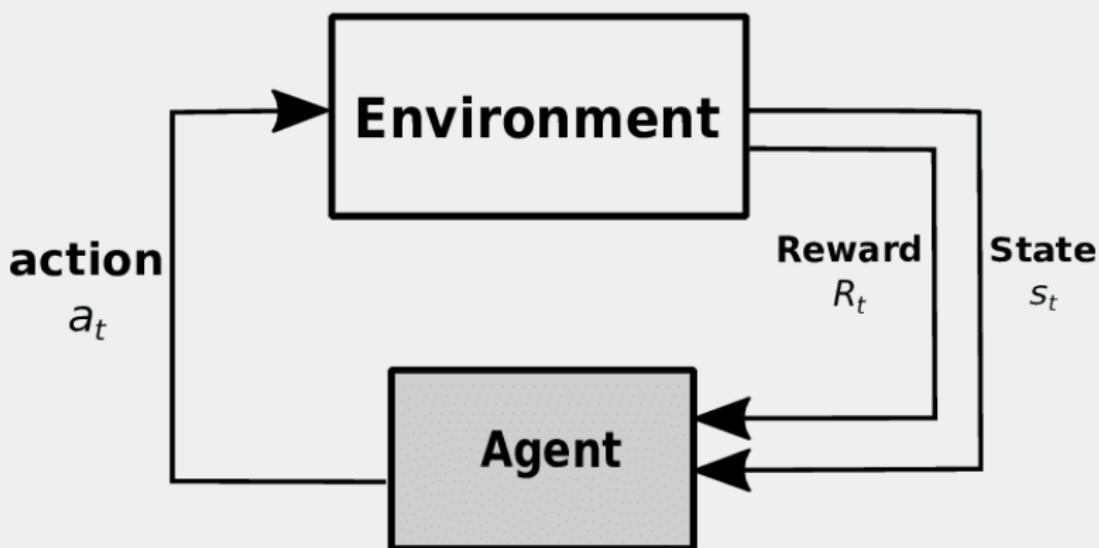
- Sequential decision making: interactions between **Agent** and **Environment**



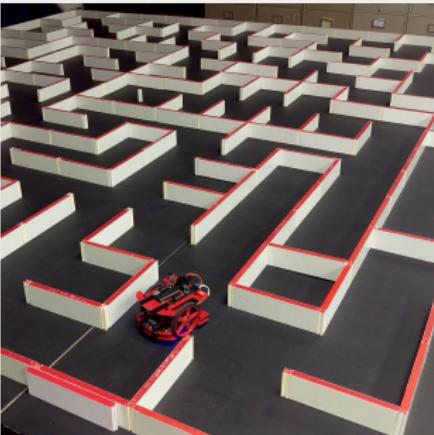
- Agent chooses an **action** to execute in the current **state** of Environment



- Environment provides **reward** to Agent and transits Agent to the next **state**



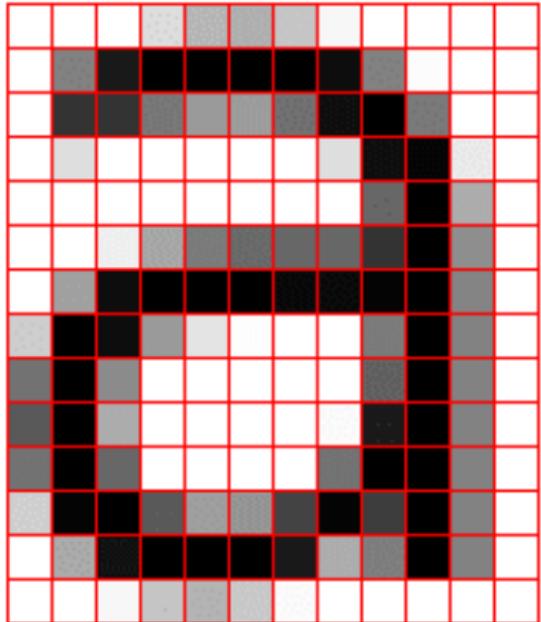
- Agent performs a sequence of actions, observes their “rewards,” and learns an optimal policy
- E.g., Go, robot and maze, Atari game, etc.



- Supervised learning
 - ▶ Solve the **recognition** problem
 - ▶ Assume i.i.d. data
 - ▶ Cannot pass human performance (i.e., ground truth)
 - Although for ImageNet: human 5% error, best model 3.57% error
- Reinforcement learning
 - ▶ Solve the **decision** problem
 - ▶ Learn via trial and error, no explicit supervisor only indirect, delayed feedback
 - ▶ Action order (time) matters → non-i.i.d. data
 - E.g., autonomous driving via a front-facing camera (image at t affects action at $t + 1$ then affects image at $t + 1$)
 - ▶ **Superhuman potential**

Learning Pipeline

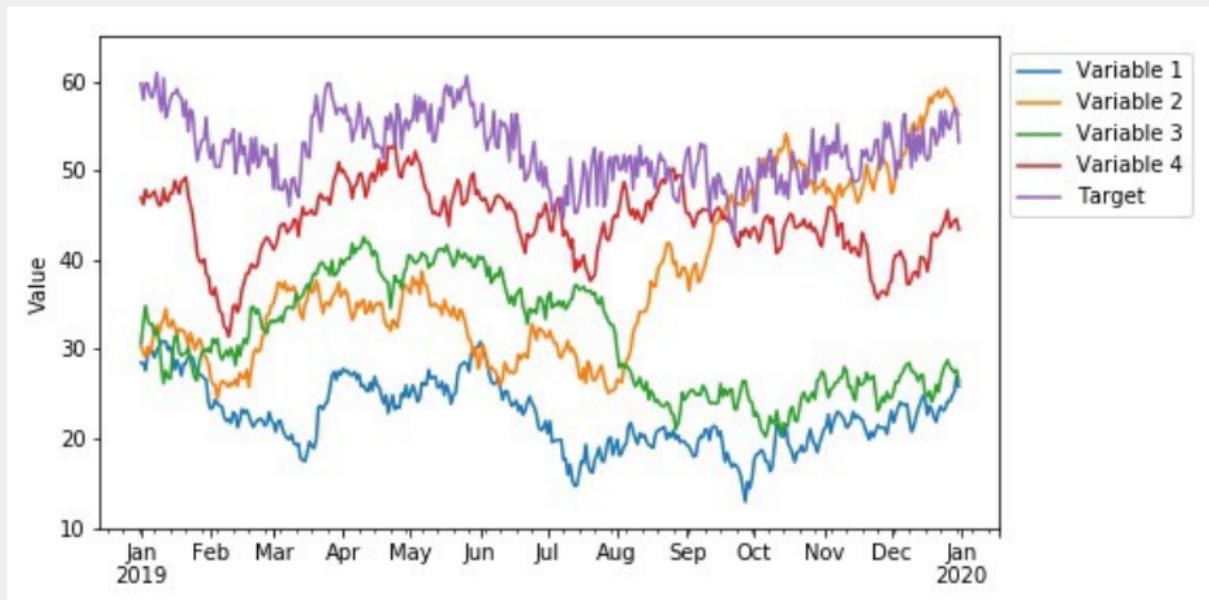
a



1.0	1.0	1.0	0.9	0.6	0.6	0.6	1.0	1.0	1.0	1.0
1.0	0.5	0.0	0.0	0.0	0.0	0.0	0.0	0.5	1.0	1.0
1.0	0.2	0.2	0.5	0.6	0.6	0.5	0.0	0.0	0.5	1.0
1.0	0.9	1.0	1.0	1.0	1.0	1.0	0.9	0.0	0.0	0.9
1.0	1.0	1.0	1.0	1.0	1.0	1.0	1.0	0.5	0.0	0.5
1.0	1.0	1.0	0.5	0.5	0.5	0.5	0.4	0.0	0.5	1.0
1.0	0.4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.5
0.9	0.0	0.0	0.6	1.0	1.0	1.0	0.5	0.0	0.5	1.0
0.5	0.0	0.6	1.0	1.0	1.0	1.0	0.5	0.0	0.5	1.0
0.5	0.0	0.7	1.0	1.0	1.0	1.0	0.0	0.0	0.5	1.0
0.6	0.0	0.6	1.0	1.0	1.0	1.0	0.5	0.0	0.0	0.5
0.9	0.1	0.0	0.6	0.7	0.7	0.7	0.5	0.0	0.5	0.0
1.0	0.7	0.1	0.0	0.0	0.0	0.1	0.9	0.8	0.0	0.5
1.0	1.0	1.0	0.8	0.8	0.9	1.0	1.0	1.0	1.0	1.0

- raw data → feature engineering via human experts → model training and tuning → outcome





- raw data → model training and tuning → outcome (a.k.a. “end-to-end learning”)

- Pros: easy to set-up, powerful (if works)

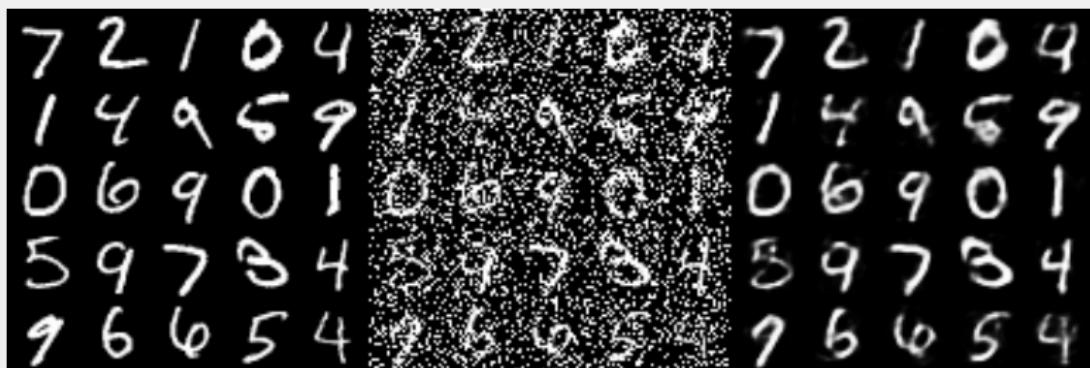
- Cons:

- ▶ blackbox (no human-interpretable features)
- ▶ difficult to defend adversarial attacks
- ▶ data hungry
- ▶ computation intensive
- ▶ hyperparameter search is a pain

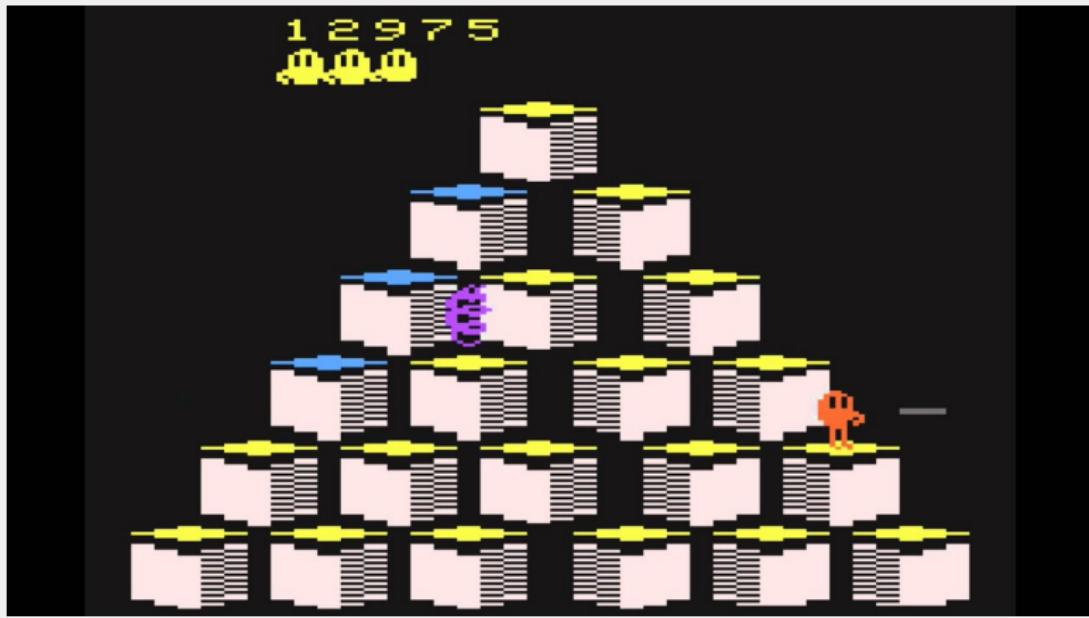
- Steering angle prediction = 90k images + 3080 (GPU) + 16 hours



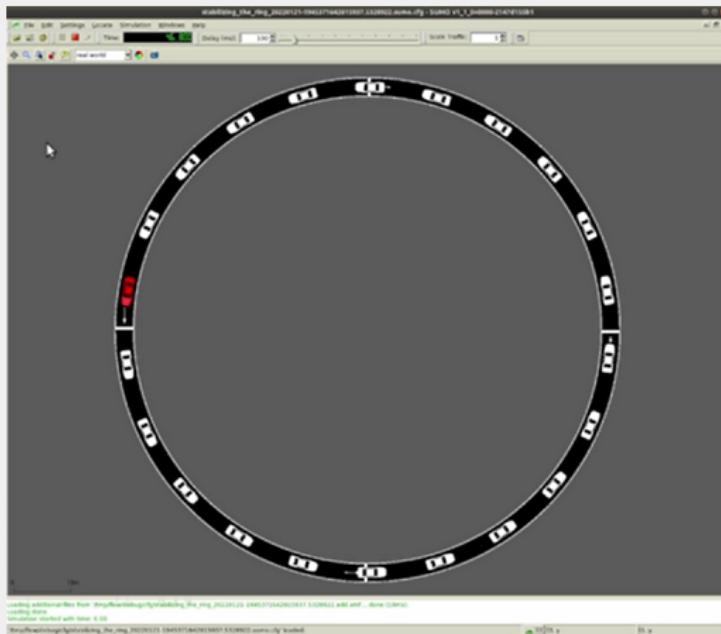
- De-noising Autoencoder: 75k images + 3080 (GPU) + 4 hours



- Q-bert: DQN (algorithm) + 3070 TI (GPU) + 2-3 hours



- Traffic on a ring: PPO (algorithm) + Intel core-i9 11900k (CPU) + 6 hours



- Pattern recognition (reporting in **accuracy**) → supervised learning
 - ▶ How to obtain (at least 20k) labelled data?
- Pattern discovery or generation (**show the found or generated patterns**) → unsupervised learning
 - ▶ How to obtain (at least 20k) pattern-specific data?
- Sequential decision making (**show agent-environment interactions**) → reinforcement learning
 - ▶ How to build the environment?
- computing resources required vs. computing resources possessed