

Recent NLP Highlights

AM11/AM12

Week 2

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A Glossary of Machine Learning Terms and Concepts

Overview

- Learning & Machine Learning
- Everything's a Function
- Experience
- Types of Machine Learning
- Supervised Learning: X , Y , h , and θ
- Loss and Cost (as a function of θ), and Gradient Descent
- Representations
- The Perceptron and Its Learning Rule; What it Can't Do
- Weights & Activation Functions
- Backpropagation
- The Computation Graph
- Fully Connected Neural Networks
- Other Types of NN Architectures

Learning

- Using experience to get better at something.
- Herb Simon: “Learning is any process by which a system improves performance from experience”
- Tom Mitchell: “A computer program is said to learn from experience E with respect to some class of tasks T and performance measure P , if its performance at tasks in T , as measured by P , improves with experience E .”
- Note: “Normal” programs don’t improve through experience. But ML models are still programs. ML is a form of *programming*. (Through spec of model + provision of data.)

Everything's a function

- Reminder: A function maps an object (from its domain) to another object (from its range). $f(x) = y$ $f: X \rightarrow Y$

$$f(x) = y \qquad f: X \rightarrow Y$$
$$f: X \rightarrow Y$$

- Understanding

$$\omega_i^n \in V^*$$

LFV

- Parsing

$$v_i \in V^*$$
$$t \in T$$

- Translation

- Dialogue

Dialogue State + w_n^u

W₁

- Labelling Images

- Seeing Similarities

- Writing Poems

- Acting in the world

- ■ ■ ■

Experience

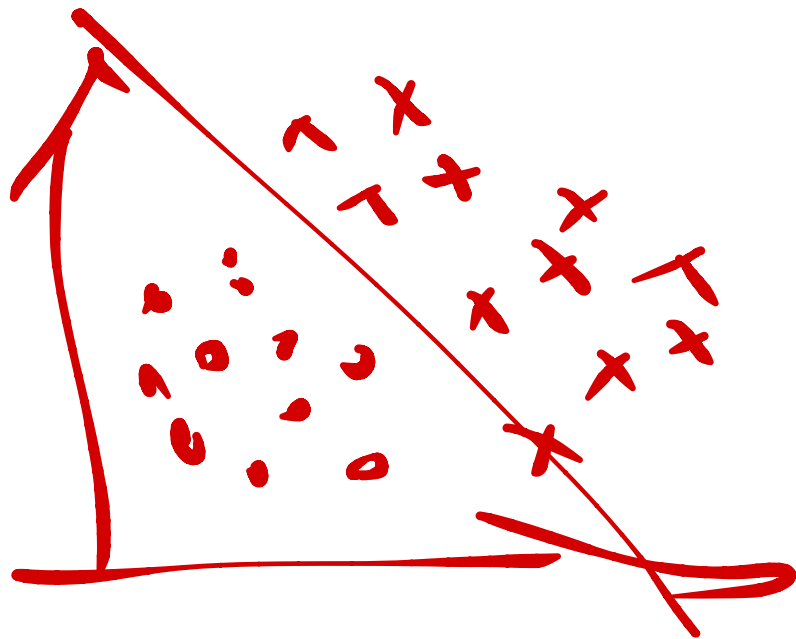
- = Data

$$X = \{x^{(1)}, x^{(2)}, \dots, x^{(m)}\} \quad x^{(i)} \in \mathcal{X}$$

~~X~~

Types of Machine Learning

- supervised learning: data + intended response
 - “supervision signal”
- unsupervised learning: just data, figure out structure



Supervised Learning

- supervised learning: data + intended response

- “supervision signal”

$$X = \{x^{(1)}, \dots, x^{(n)}\} \quad Y = \{y^{(1)}, \dots, y^{(n)}\}$$

- Hypothesis, Loss, Cost, Objective Function

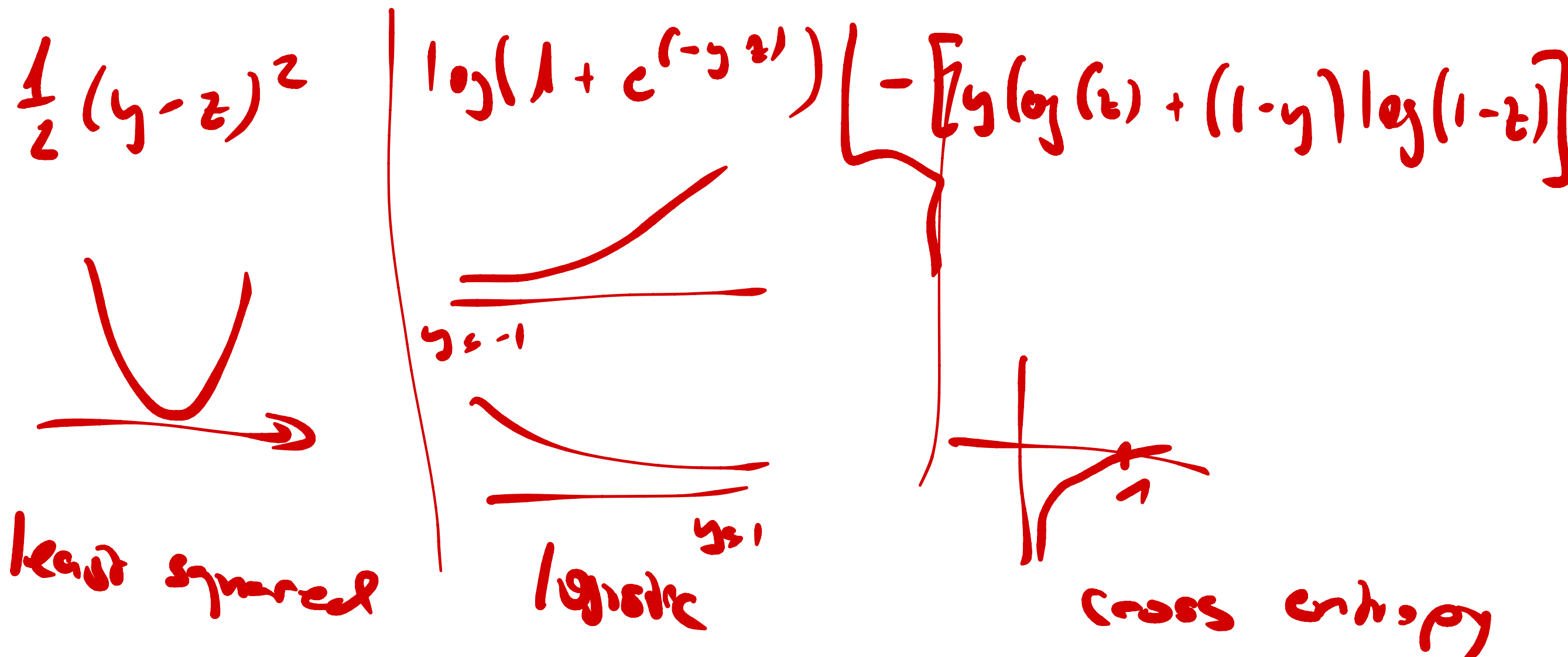
$$h_{\theta}(x^{(i)}) = z^{(i)}$$

$$L: (z, y) \in \mathbb{R} \times Y \mapsto L(z, y) \in \mathbb{R}$$

$$J(\theta) = \sum_{i=1}^n L(h_{\theta}(x^{(i)}), y^{(i)})$$

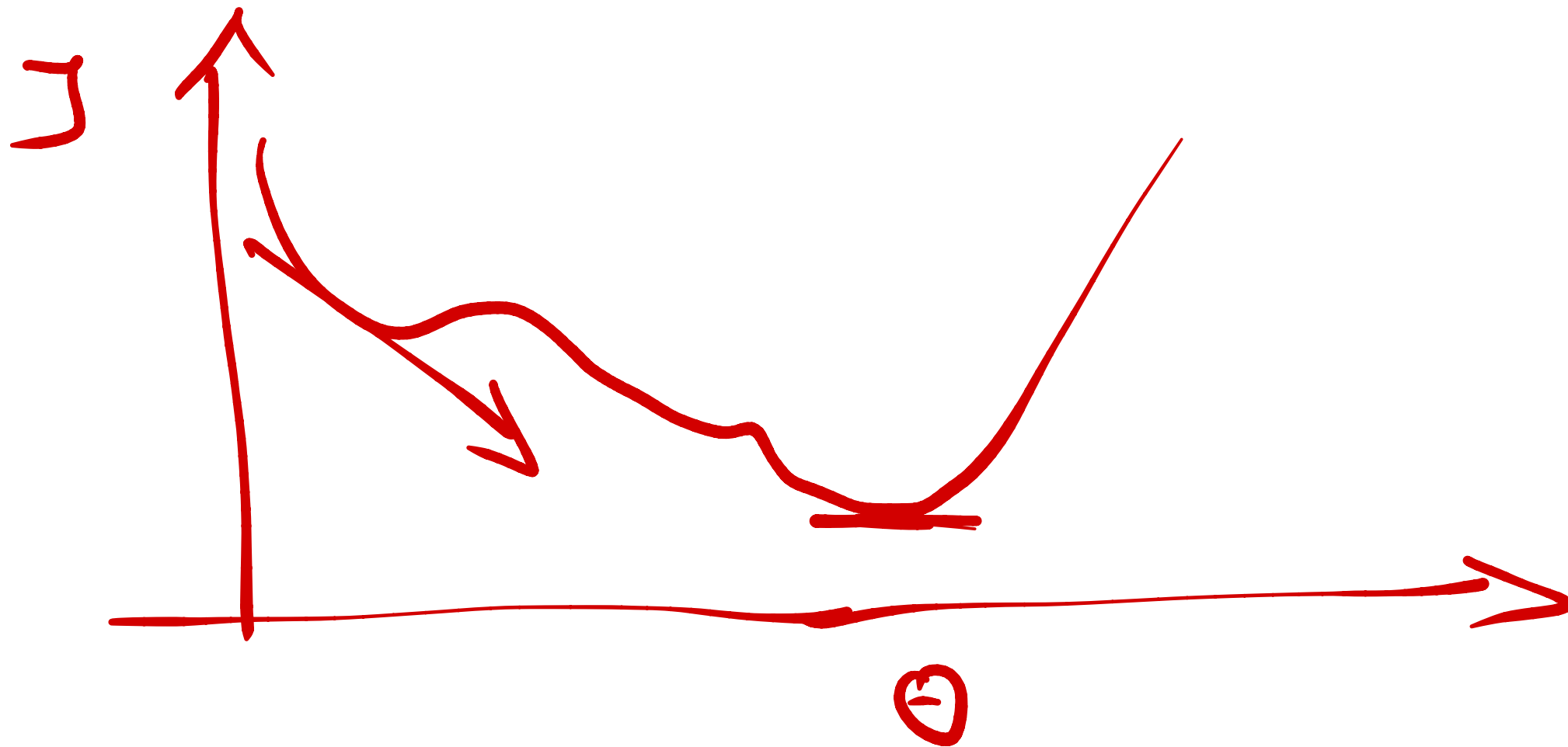
Loss & Cost

- main concept: Loss and cost as function of parameters; data is kept constant.
- loss function direct the learning. different kinds exist.



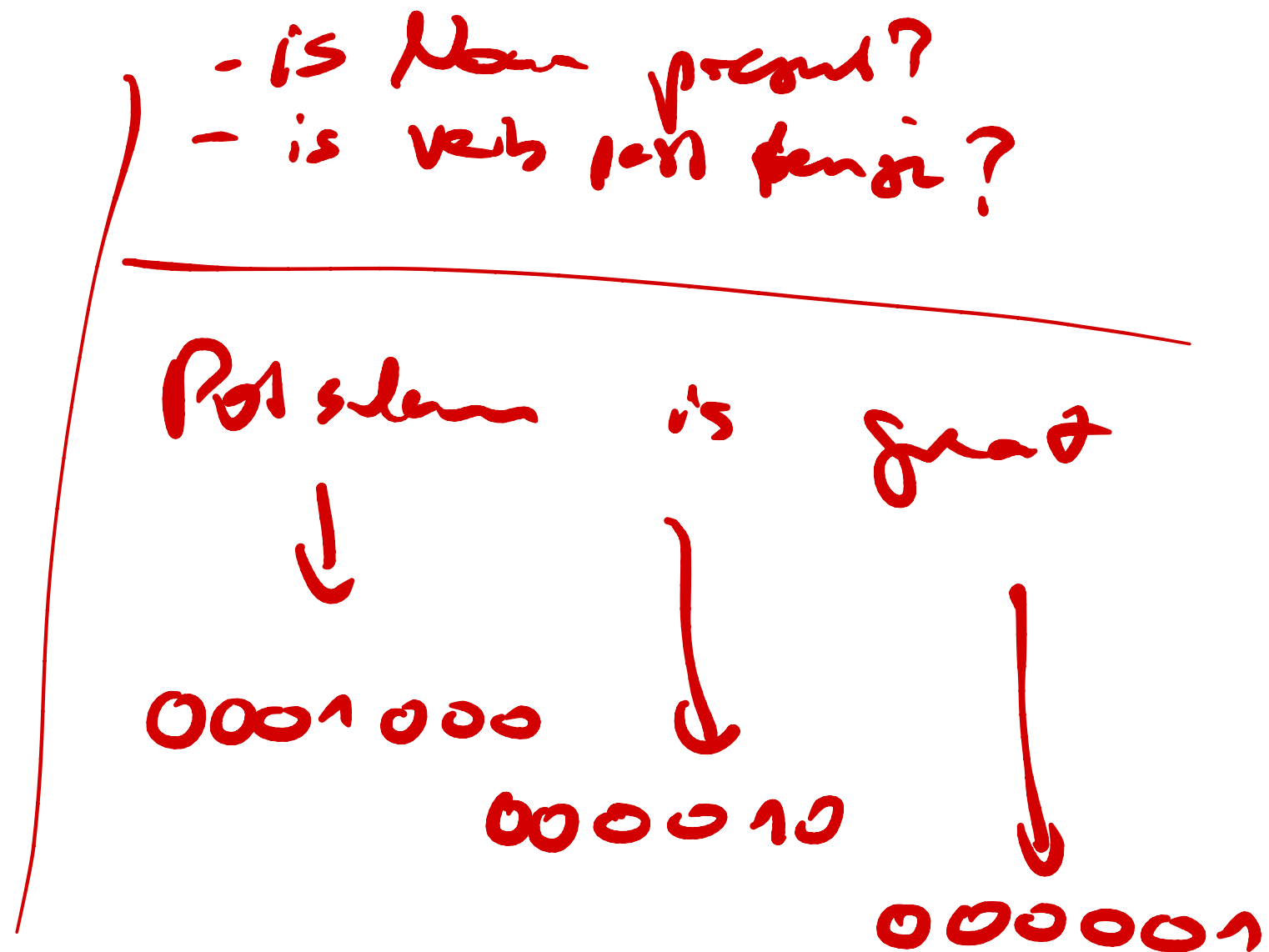
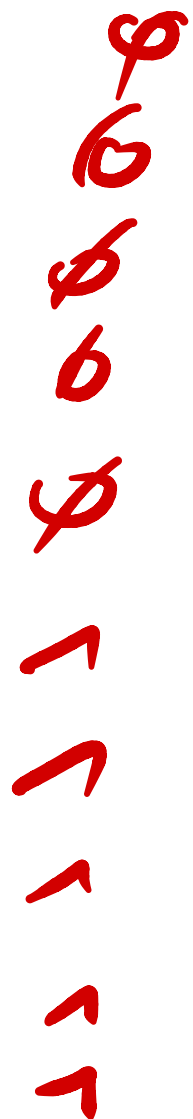
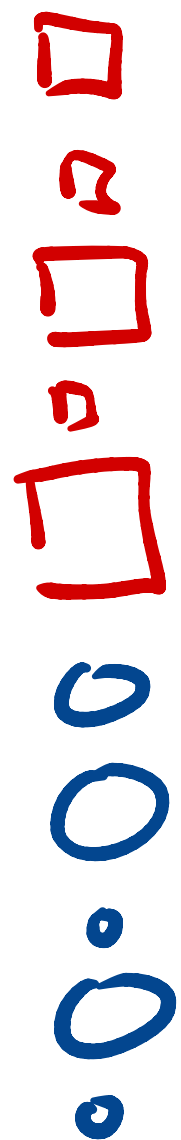
Minimise Cost

- differential calculus, finding minima

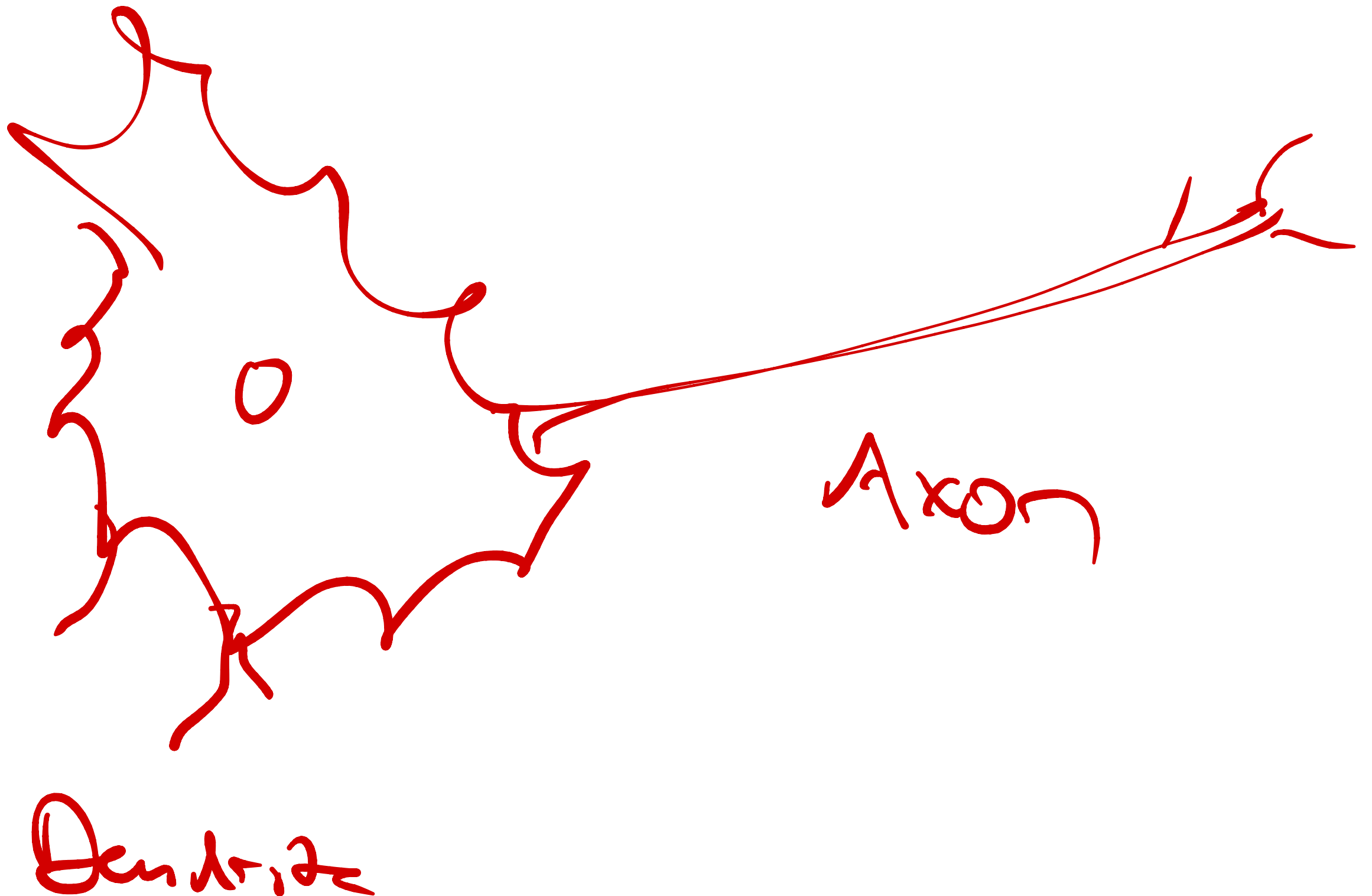


Representations

- Description of object, presenting the important (for the task) aspects

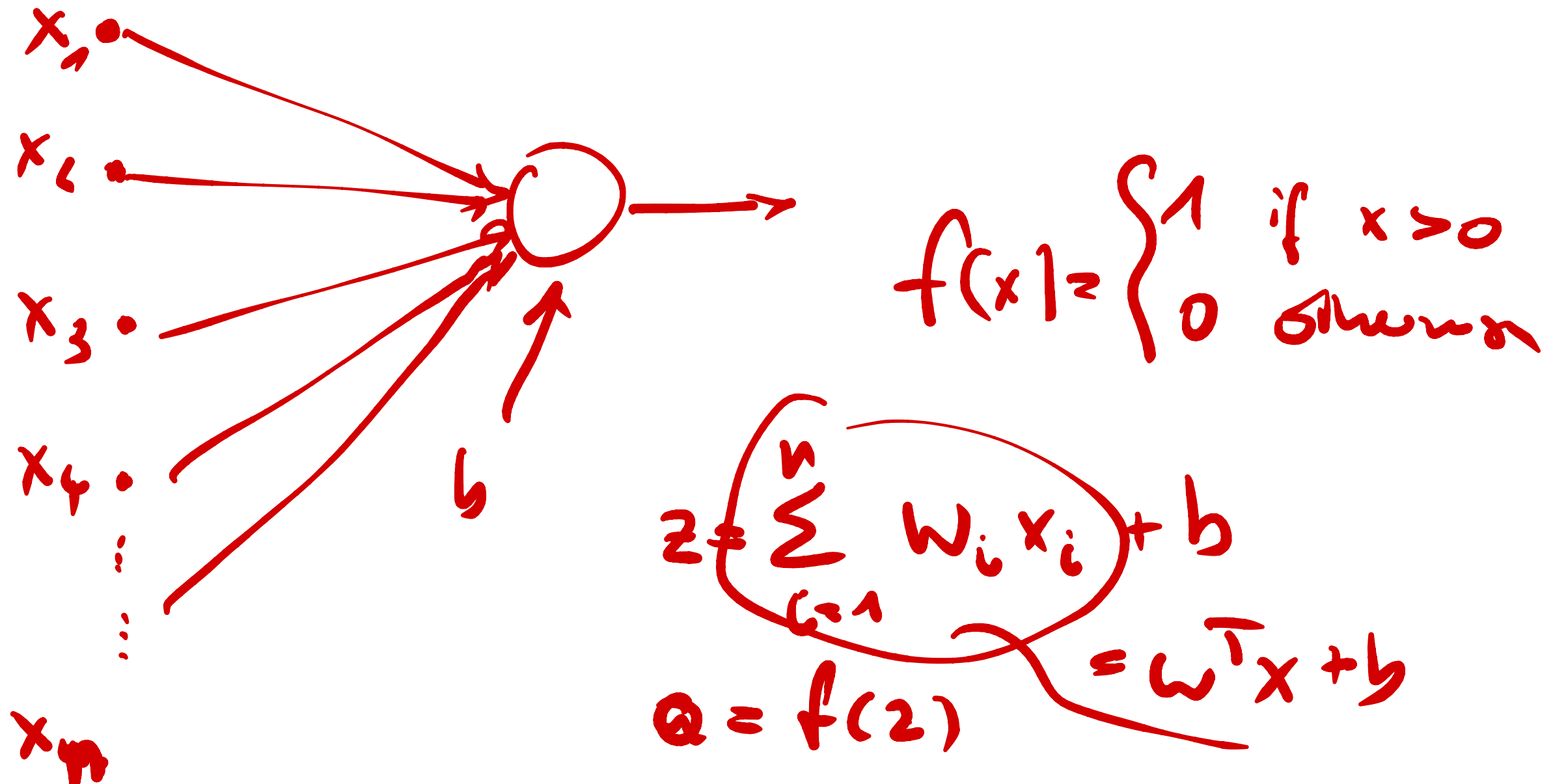


A Neuron



The Perceptron

- Rosenblatt 1957

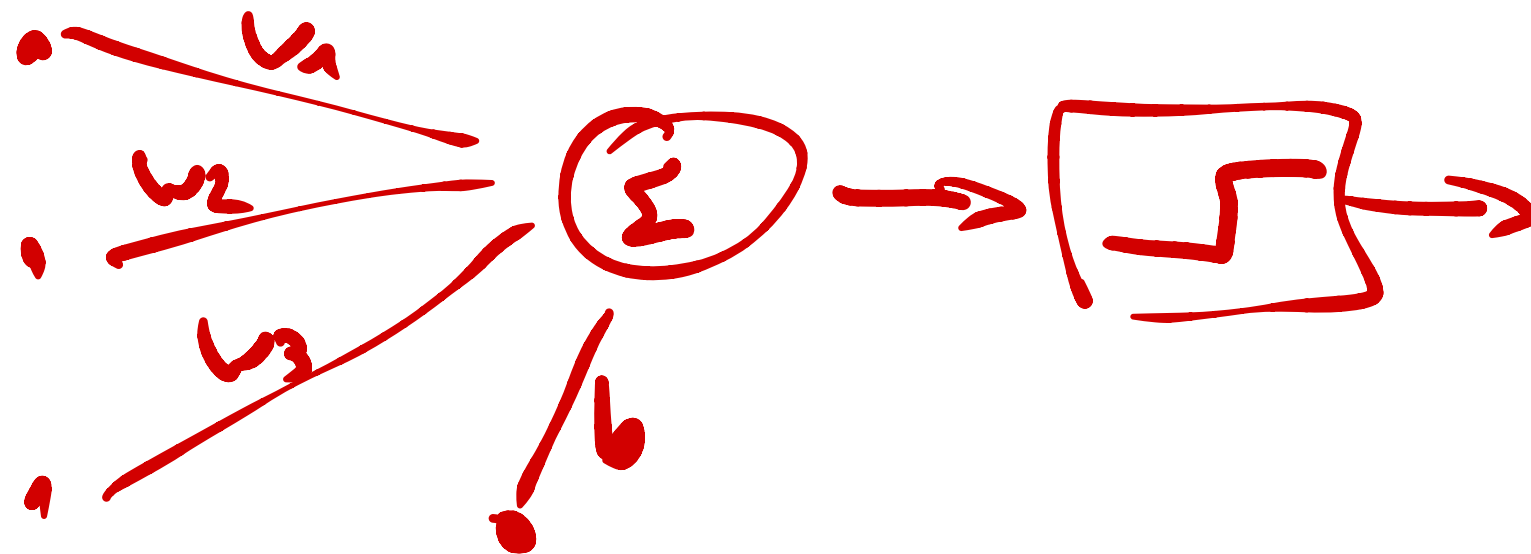


The Perceptron

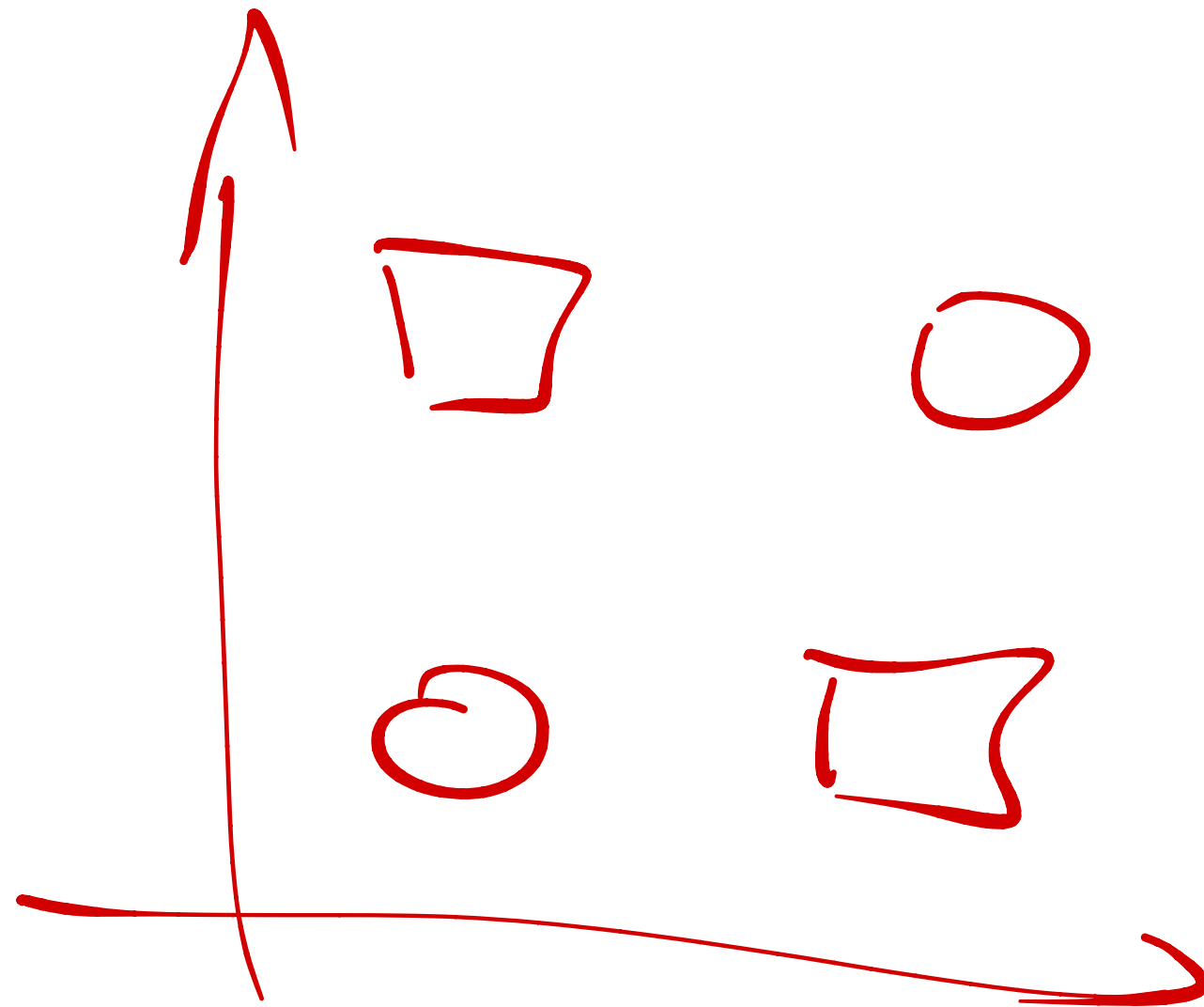
$$z^{(i)} = f(w_t x^{(i)})$$

$$= f(w_{1t} \cdot x_1 + \dots + w_{nt} \cdot x_n + b)$$

$$w_{i(t+1)} = w_{it} + \eta \cdot (y^{(i)} - z^{(i)}) x_i^{(i)}$$



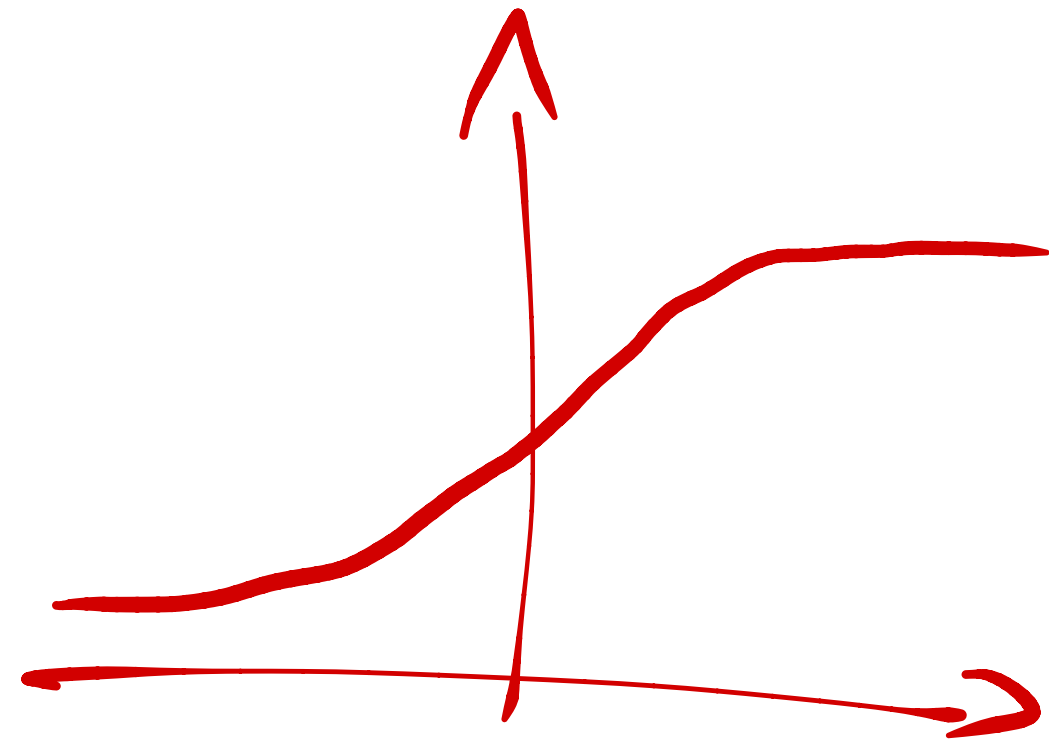
What the Perceptron Can't Learn



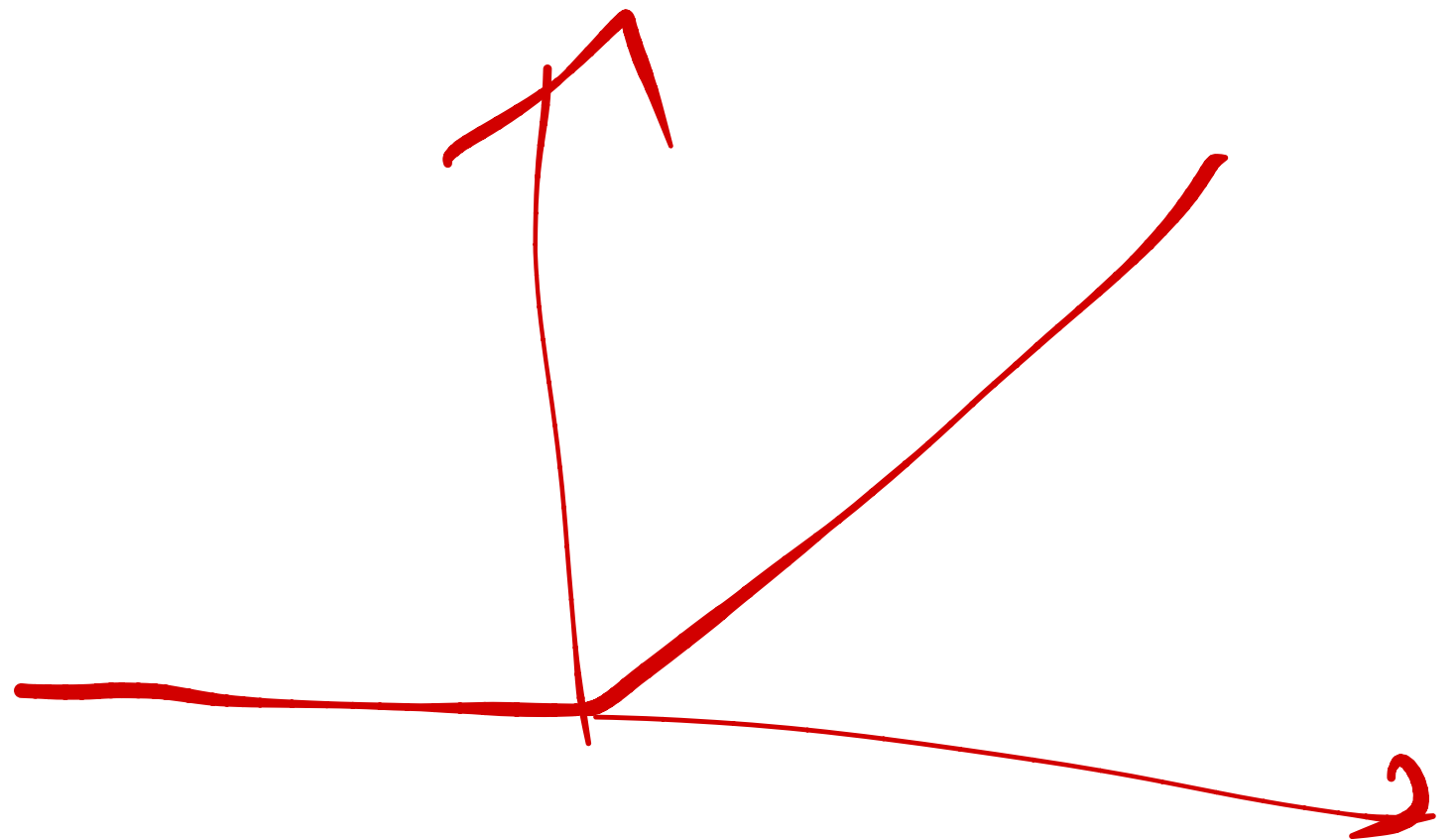
Other Activation Functions

Sigmoid

$$S(z) = \frac{1}{1 + e^{-z}}$$

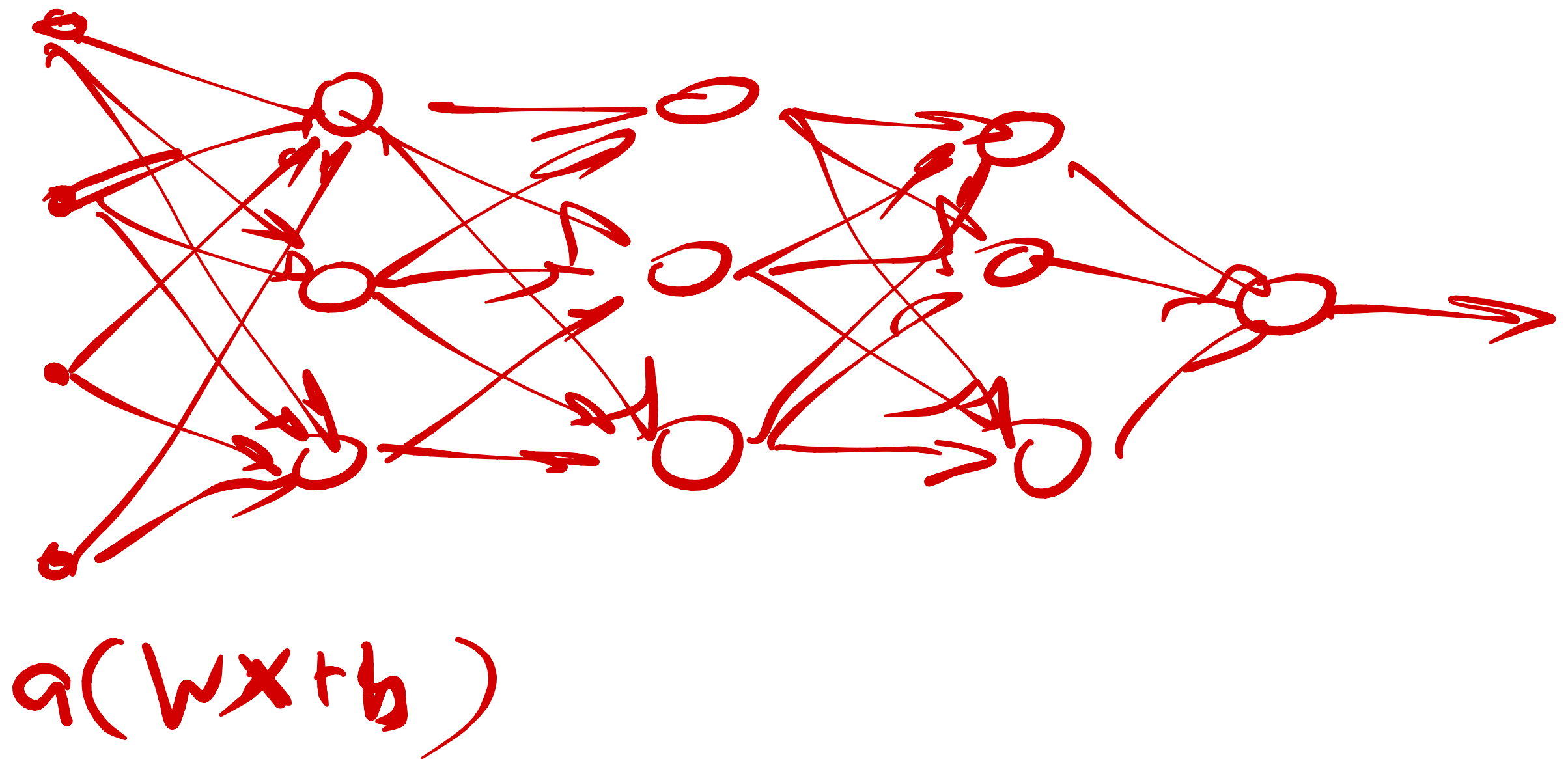


ReLU



$\max(0, z)$

Deep (= Multilayer) Networks



How does the supervision signal get to each weight?

- Backpropagation

$$f(g(x))$$

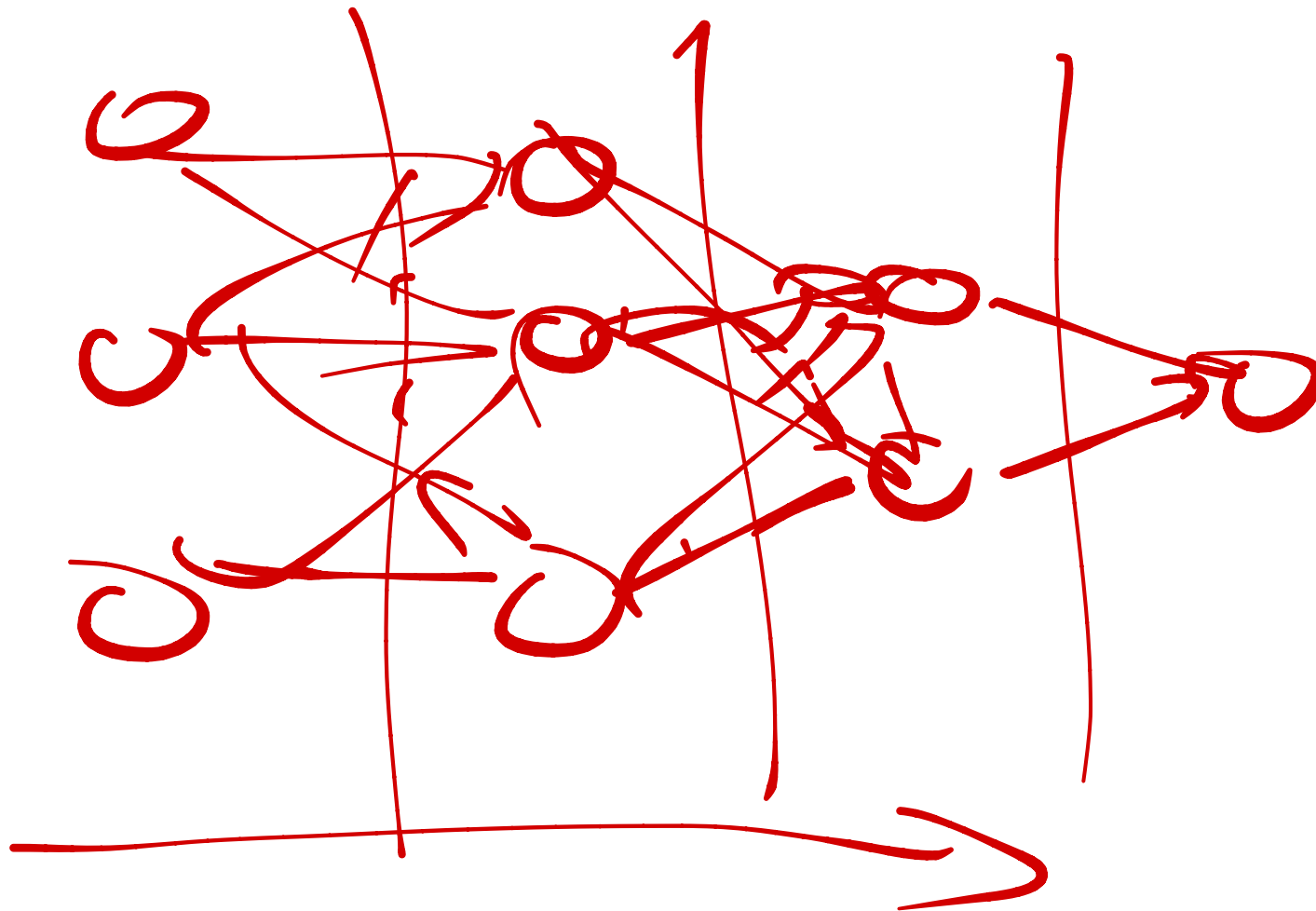
The Computation Graph

Theras

Tensorflow

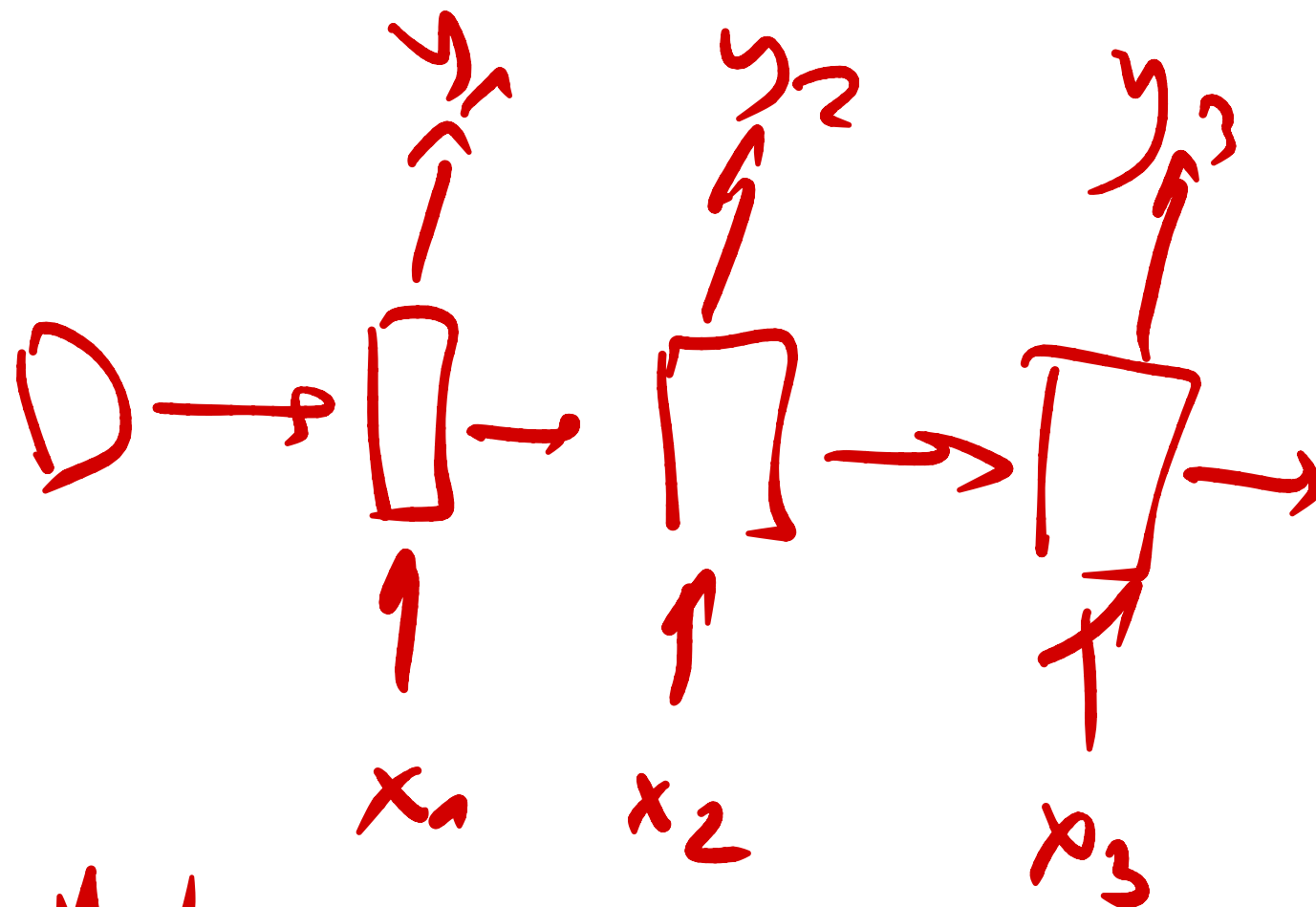
PyTorch

Fully Connected (Feed Forward) Network



Convolutional Networks

Recurrent Networks



RNN
LSTM

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advertisement

- The CL Colloquium! Wednesdays, 16-18h.
<https://github.com/compling-potsdam/sose19-cl-colloquium>

Week	Date	Presenter(s)	Title / Abstract
01	2019-04-10	<i>no talk scheduled</i>	
02	2019-04-17	David Schlangen	<i>Information on getting credit for this course</i>
03	2019-04-24	CL students	Poster Slam
04	2019-05-01	<i>public holiday</i>	
05	2019-05-08	Alan Nichols (RASA)	tba
06	2019-05-15	Shlomi Hod (Potsdam)	tba
07	2019-05-22	Alan Akbik (Zalando)	tba
08	2019-05-29	<i>no talk scheduled</i>	
09	2019-06-05	<i>no talk scheduled</i>	
10	2019-06-12	Staffan Larsson (Gothenburg)	tba
11	2019-06-19	Feiyu Xu (Lenovo) <i>to be confirmed</i>	tba
12	2019-06-26	Milica Gasic (Düsseldorf)	tba
13	2019-07-03	Raquel Fernández (Amsterdam)	tba
14	2019-07-10	<i>no talk scheduled</i>	
15	2019-07-17	<i>no talk scheduled</i>	