

# Deep learning frameworks and differentiable programming

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# Programming by example.

- Given some constraints on the desired behavior of a program (a set of input output pairs examples).
- Rather than writing the rules (program) that accomplish a desired behavior (for input A we want output to be B)
- Search the program space for a program that satisfies the constraints.

# How to search?

- Discrete search.
- Relax the problem and make the components differentiable so we can apply gradient based optimization.

# Making problem differentiable

- Think about a program as a function parameterized with a set of real numbers.
  - $Y = f(X; W)$ .
  - For each  $W$ , the function  $f$  maps input  $X$  to  $Y$  in a different way.
  - Search is for the right function becomes search for  $W$ .
- In addition, input and output spaces must be relaxed (made continuous) as well as the criteria that evaluates the fit of a particular function.

# Simple example

- Learn the OR function

X1	X2	Y
0	0	0
0	1	1
1	0	1
1	1	1

# Simple example ...

- Similarly, we can learn the AND function or the NAND function.

X1	X2	Y
0	0	1
0	1	1
1	0	1
1	1	0

# Composition of functions

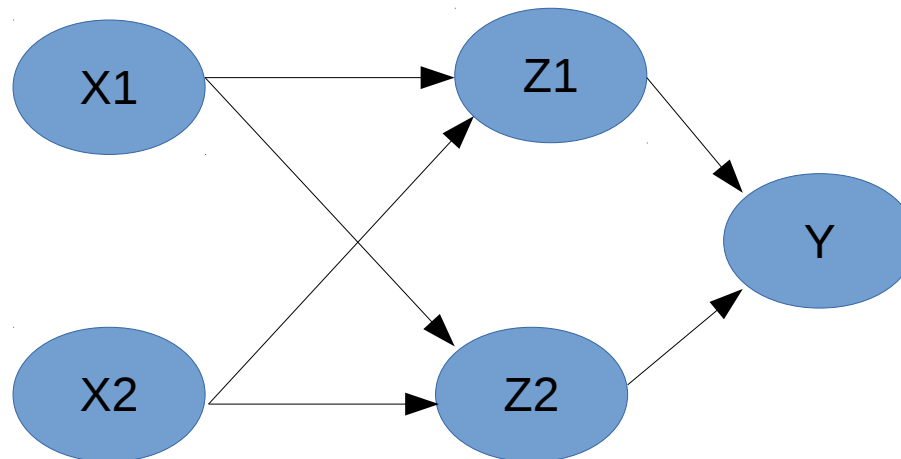
Composition of functions can give us more expressive power.

- Secondary Boolean functions such as XOR.

X1	X2	Y
0	0	0
0	1	1
1	0	1
1	1	0

# Composition of functions

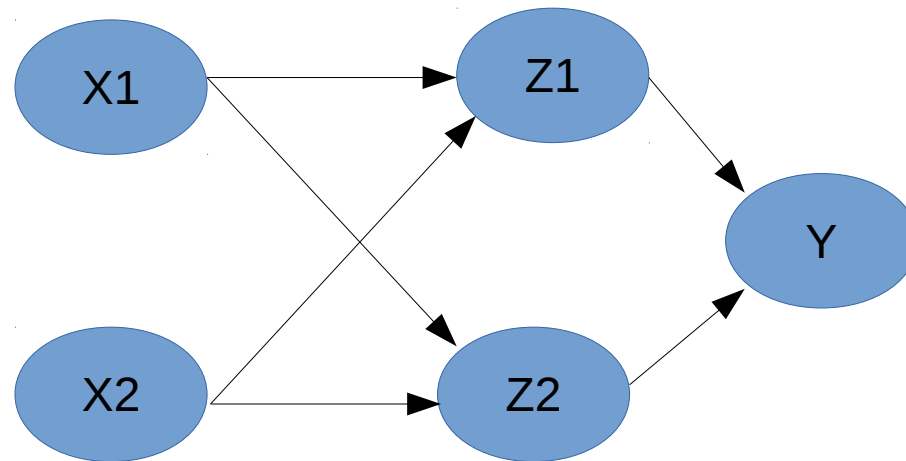
- Look at the expression for the XOR
$$Y = ((X1 \text{ OR } X2) \text{ OR } (X1 \text{ NAND } X2))$$
- Define intermediate variables
- $Z1 = (X1 \text{ OR } X2)$  and  $Z2 = (X1 \text{ NAND } X2)$
- $Y = (Z1 \text{ OR } Z2)$





# Composition of functions

- Look deeper into the derivatives of the function



- We want the derivatives of the fit function with respect to the parameters of the functions.

# Composition of functions

- Chain rule

