

Deep learning and differentiable programming

Luis Gonzalo Sanchez Giraldo

- https://youtu.be/_rdINNHLYaQ

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

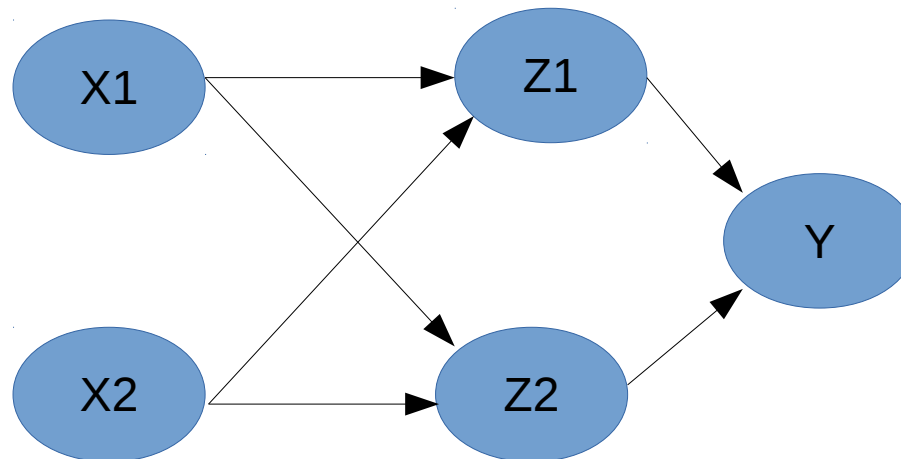
Composition of functions can give us more expressive power.

- Secondary Boolean functions such as XOR.

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

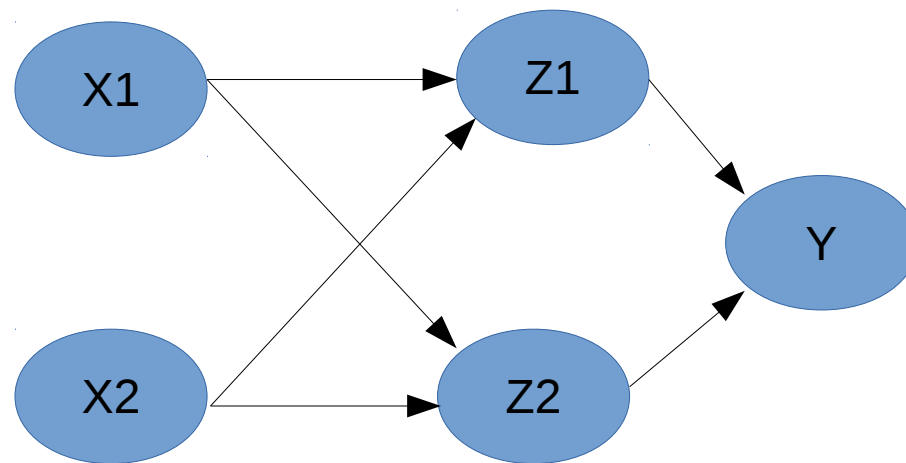
Composition of functions

- 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

