

Demystifying AI

Neural Networks

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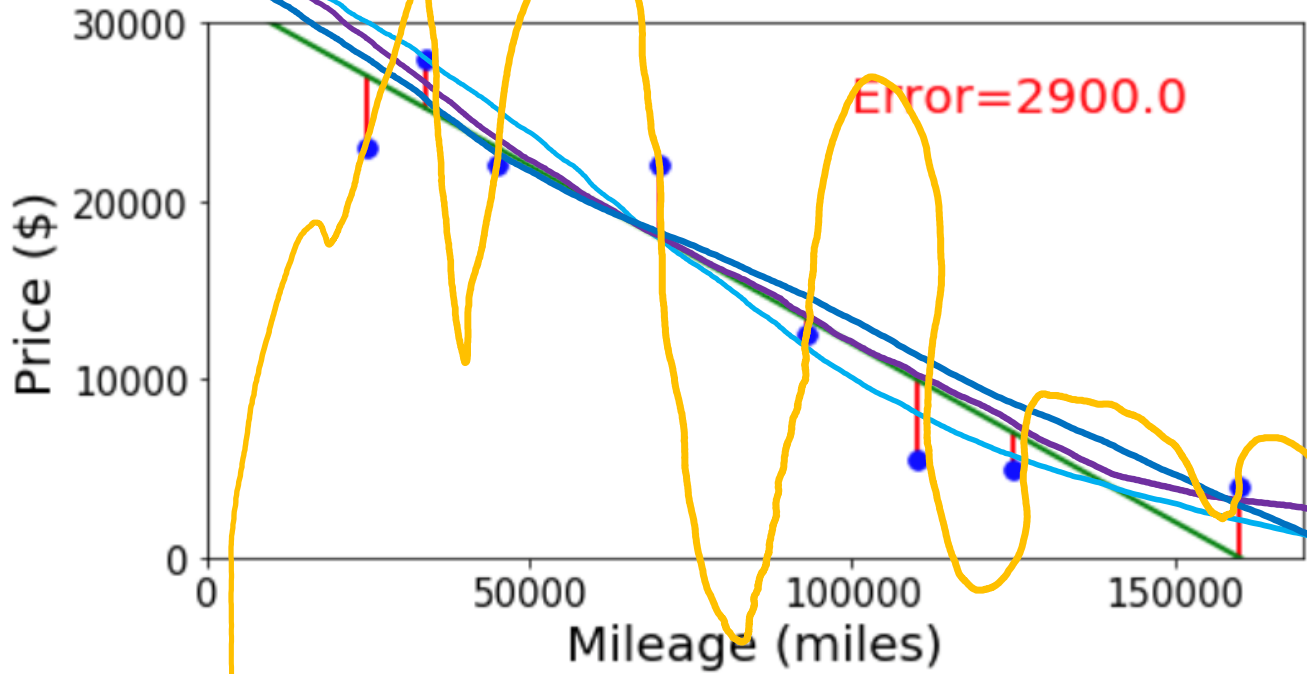
# Warm-up Exercise: Plotting Functions

Paper handout



# Linear Regression

## Selling my car example



# Regression for Non-linear data

Paper handout

# Linear Regression

## With N-D inputs

1D

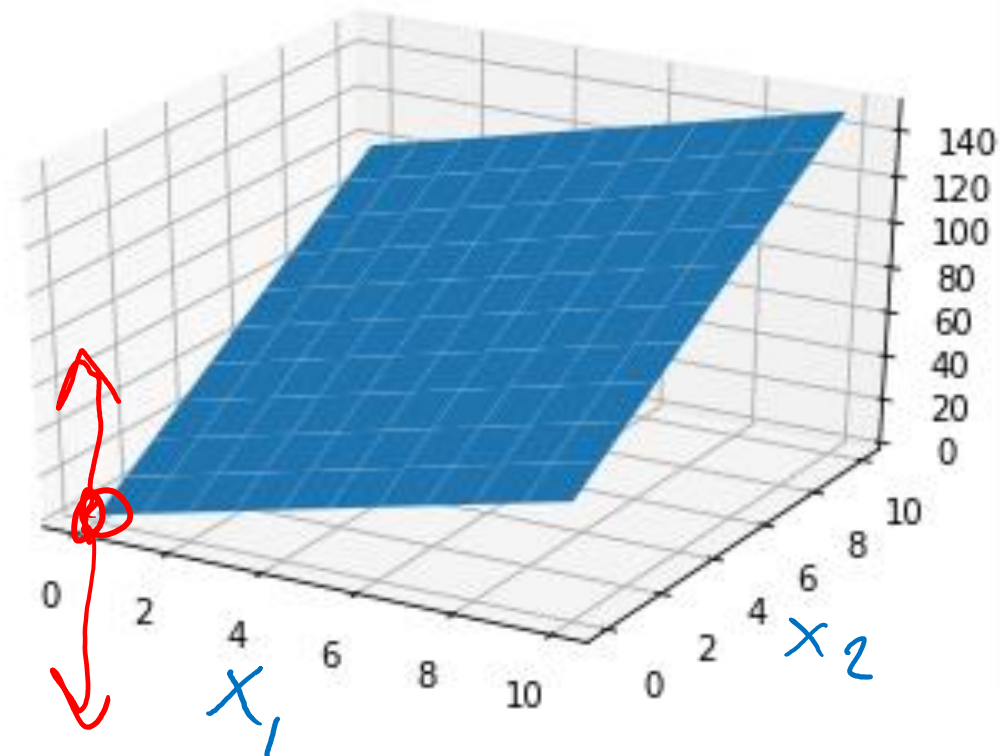
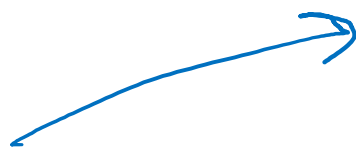
$$\hat{y} = w x_1 + b$$

2D

$$\hat{y} = w_1 x_1 + w_2 x_2 + \underline{b}$$

ND

$$\hat{y} = \left( \sum_{i=1}^N w_i x_i \right) + b$$



# Linear Regression

With N-D inputs

1-D linear function

$$y = w_1 x_1 + b$$

2-D linear function

$$y = w_1 x_1 + w_2 x_2 + b$$

3-D linear function

$$y = w_1 x_1 + w_2 x_2 + w_2 x_2 + b$$

N-D linear function

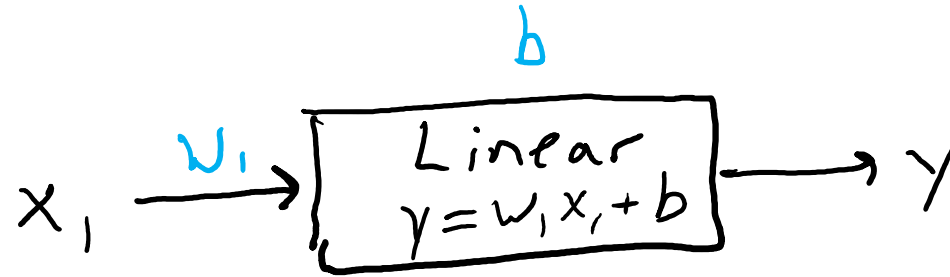
$$y = \sum_{i=1}^N w_i x_i + b$$

# Network Diagrams

## Linear functions

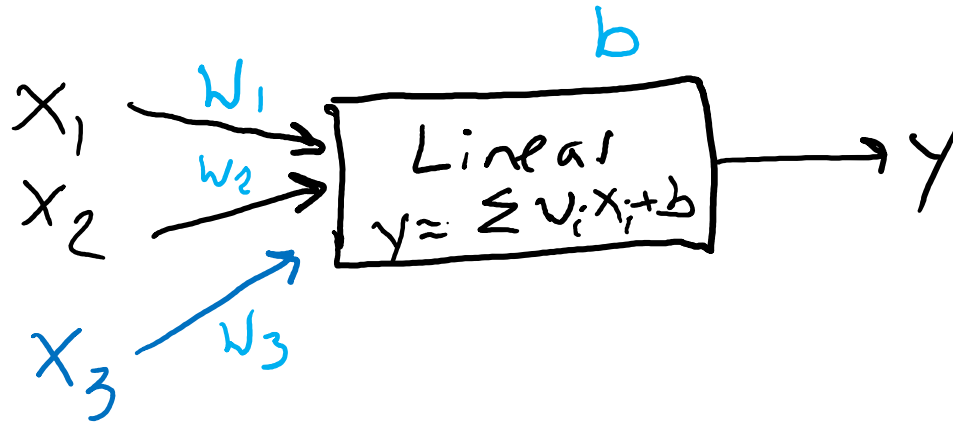
### 1-D linear function

$$y = w_1 x_1 + b$$



### 2-D linear function

$$y = \underbrace{w_1 x_1 + w_2 x_2}_{\text{sum}} + b$$



### 3-D linear function

$$y = w_1 x_1 + w_2 x_2 + w_2 x_2 + b$$

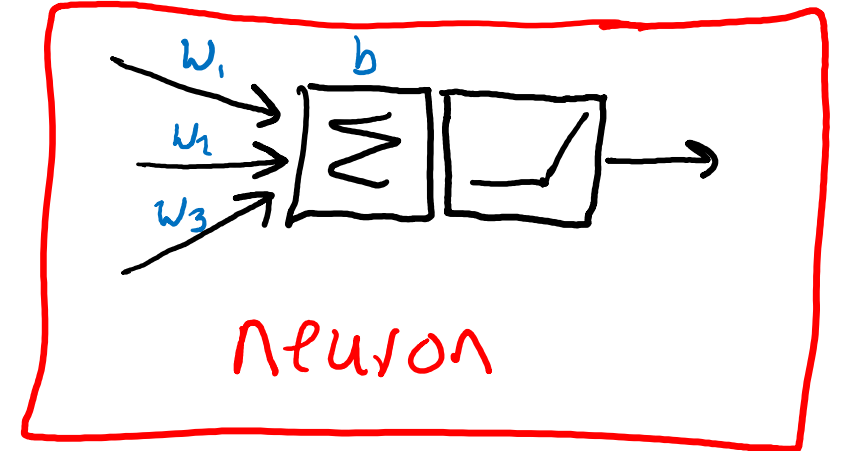
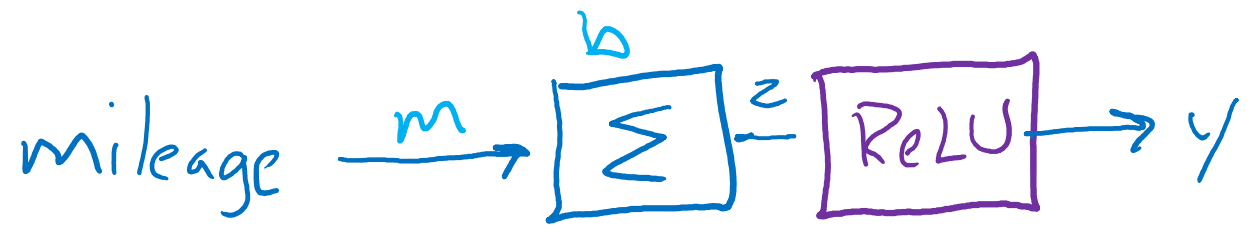
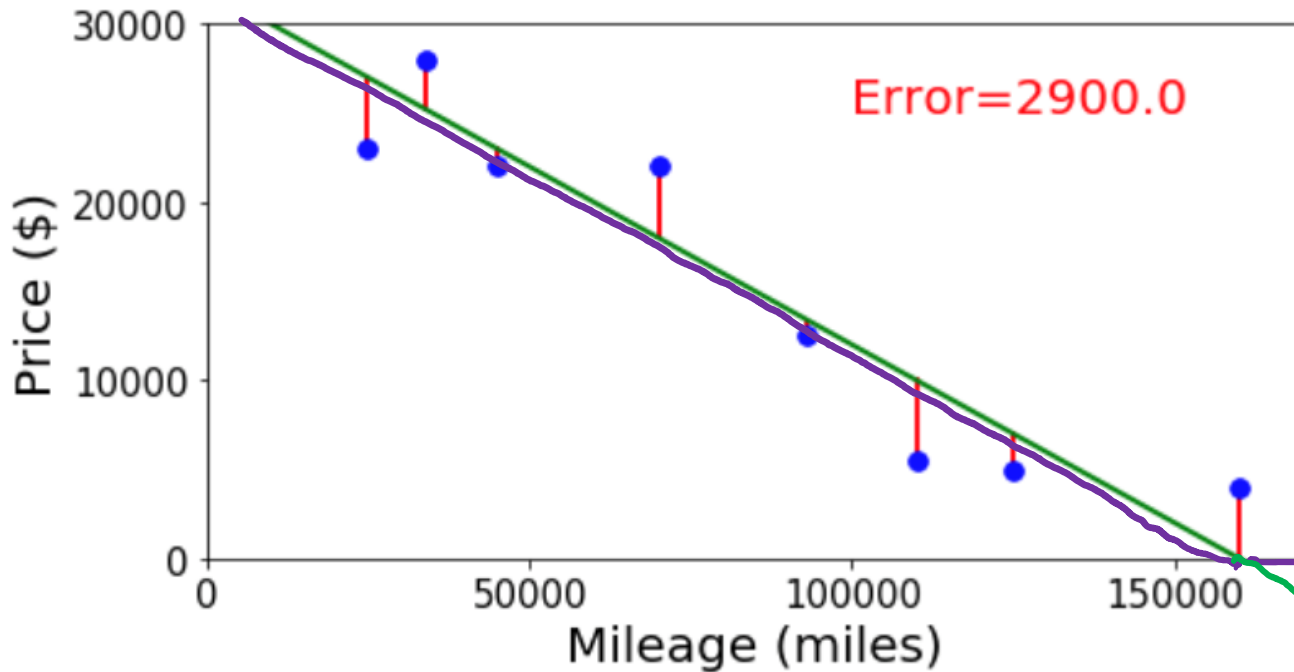


### N-D linear function

$$y = \underbrace{\sum_{i=1}^N w_i x_i}_{\text{sum}} + b$$

# Linear Regression

## Selling my car example



$$z = mx + b$$

$$a = \max(0, z)$$

$$a = \text{ReLU}(z)$$



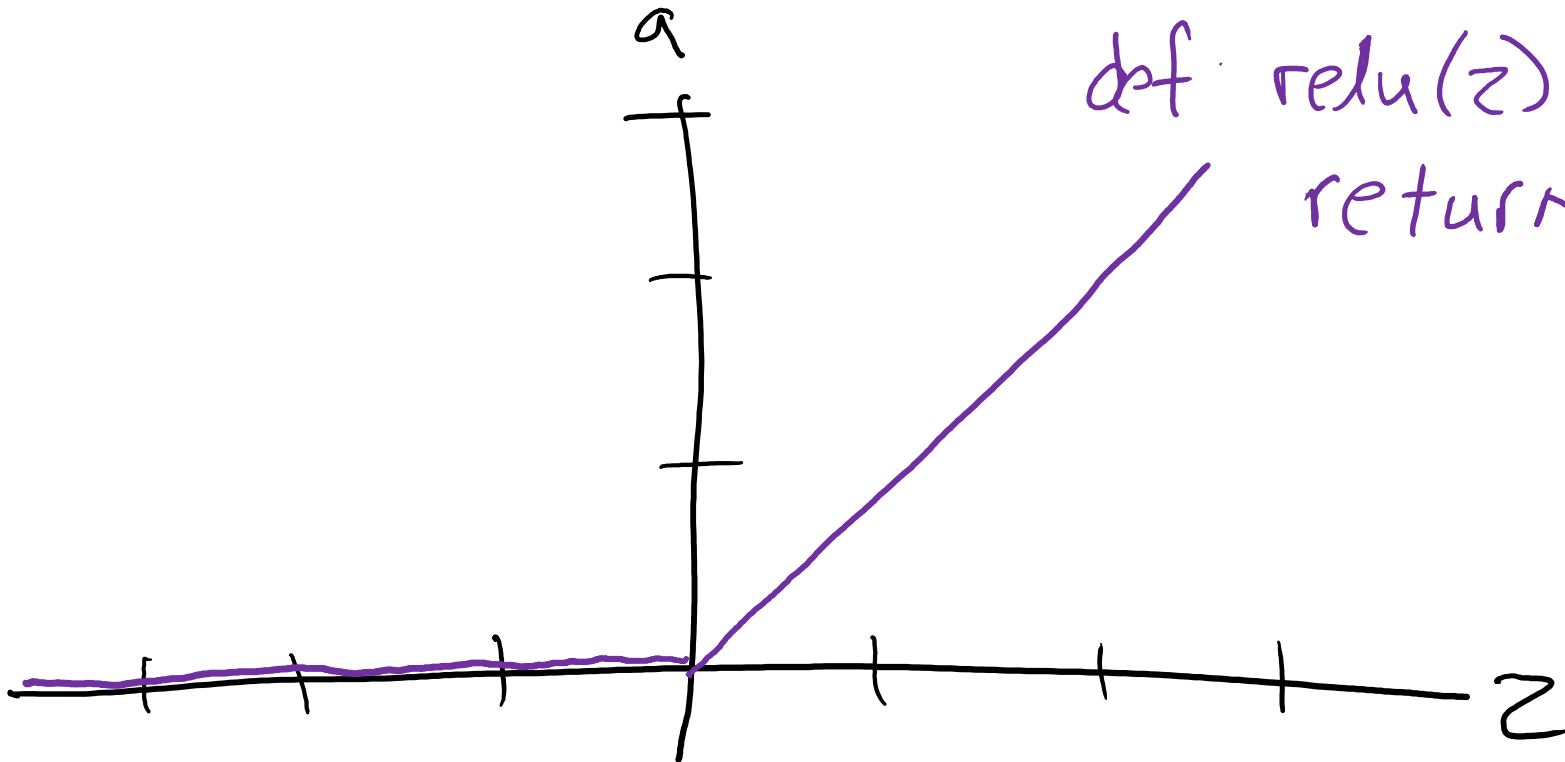
Linear plus ReLU  
Rectified linear unit

$$a = \text{ReLU}(z)$$

$$= \begin{cases} 0 & \text{if } z < 0 \\ z & \text{otherwise} \end{cases}$$

def relu(z)

return max(0, z)



# Plotting Functions

## Connecting functions

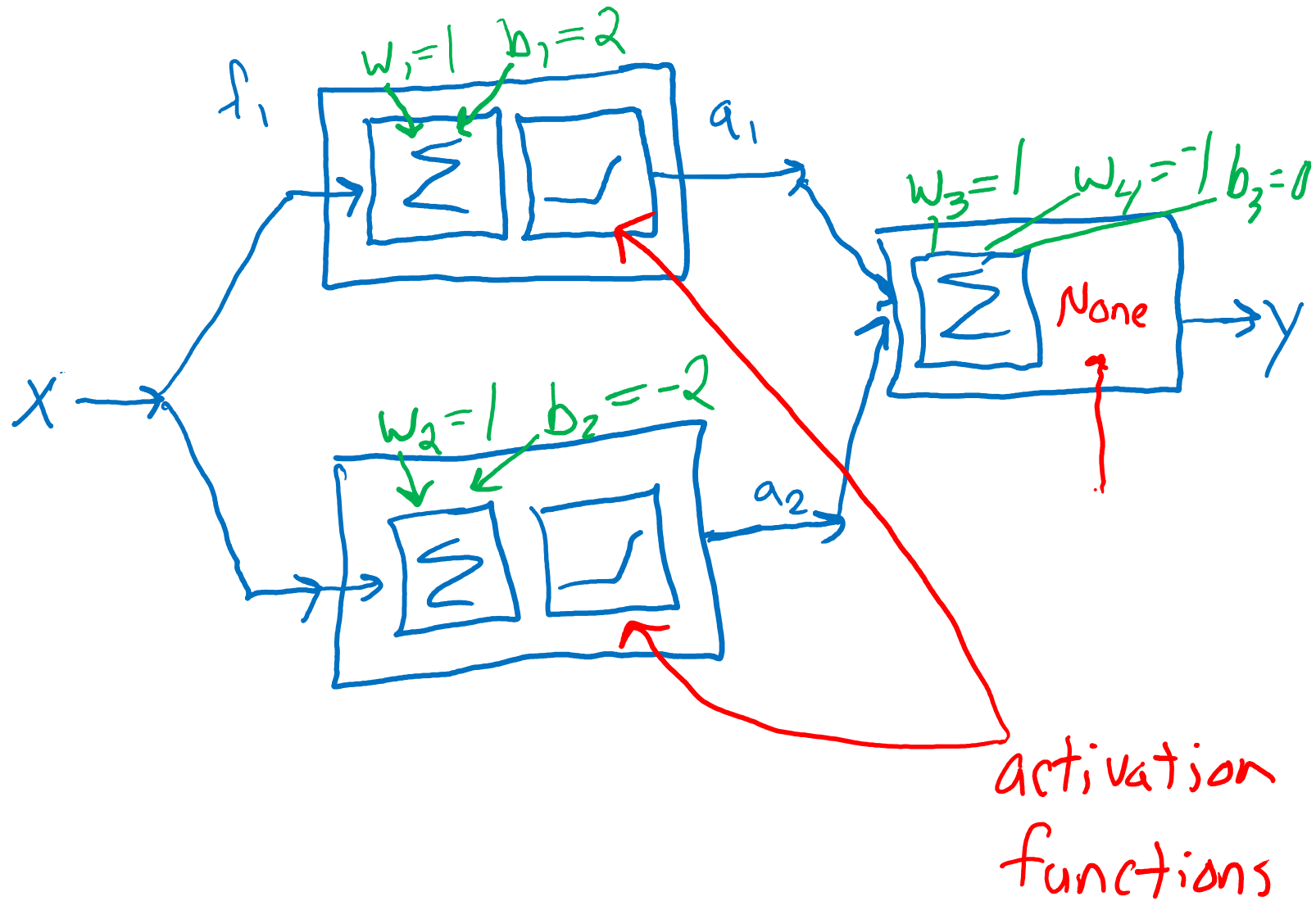
Plot the function  $f_3(x)$  vs  $x$

$$a_1 = f_1(x) = \max(0, 1x + 2)$$

$$a_2 = f_2(x) = \max(0, 1x - 2)$$

$$y = f_3(x) = f_1(x) - f_2(x)$$

$$1 \cdot a_1 + (-1) \cdot a_2 + 0$$

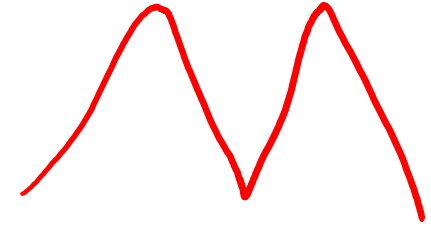


# Neural Networks

<https://www.cs.cmu.edu/~15181/tfp>

## Setup:

- Switch to the regression dataset that looks like the letter M
- Set the learning rate to 0.003



## Steps:

- Set up your architecture: add as many hidden layers and neurons per layer as you like
- Click the play button
- Observe the resulting fit and loss (mean squared error)
- Repeat to try to use as few neurons as possible and still get a good fit

# Three-neuron network

## Connecting functions

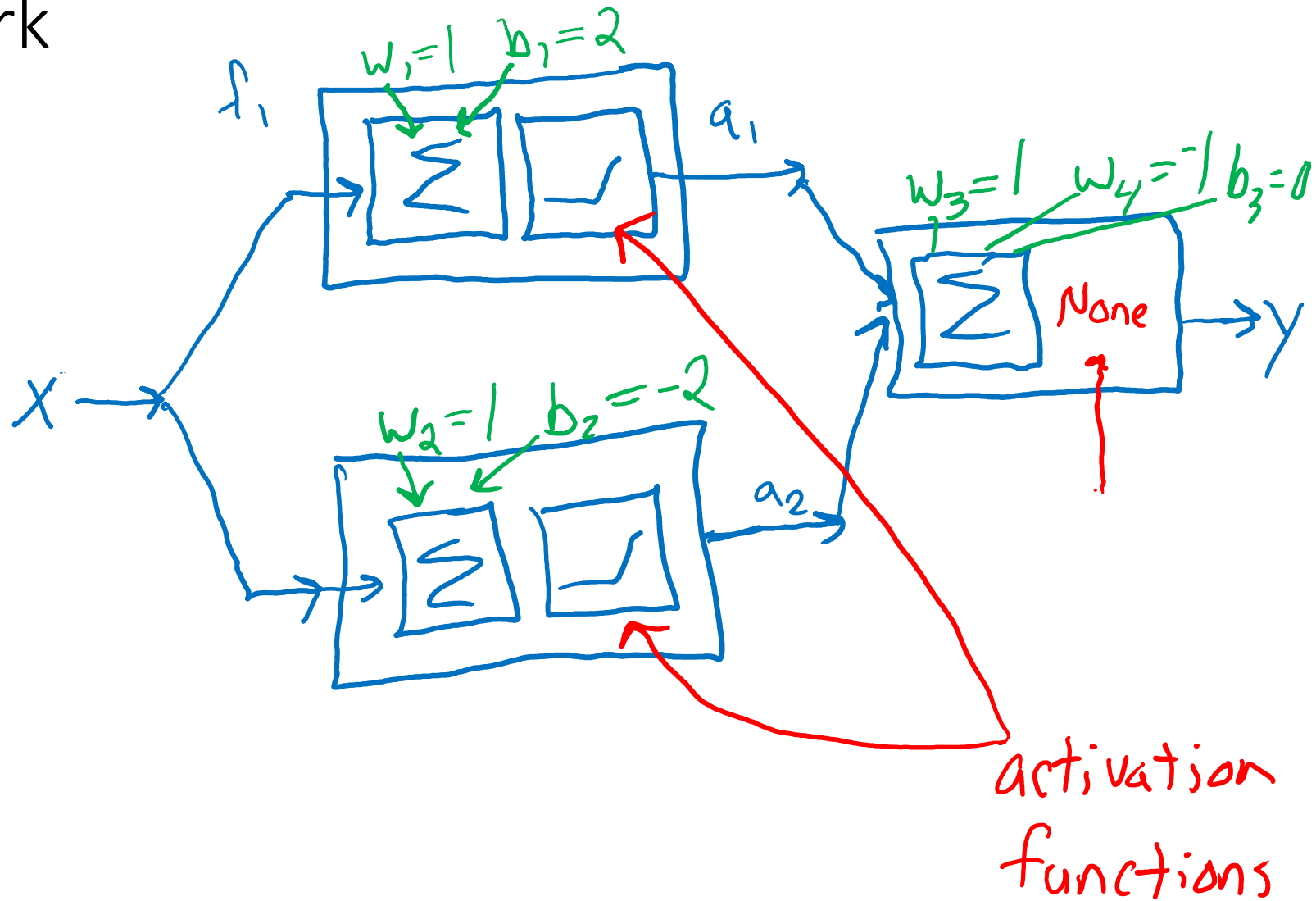
Plot the function  $f_3(x)$  vs  $x$

$$a_1 = f_1(x) = \max(0, 1x + 2)$$

$$a_2 = f_2(x) = \max(0, 1x - 2)$$

$$y = f_3(x) = f_1(x) - f_2(x)$$

$$1 \cdot a_1 + (-1) a_2 + 0$$



# Three-neuron network

## Connecting functions

Plot the function  $f_3(x)$  vs  $x$

$$f_1(x) = \max(0, 1x + 2)$$

$$f_2(x) = \max(0, 1x - 2)$$

$$f_3(x) = f_1(x) - f_2(x)$$

