

Homework 2

What is Deep learning?
It's a part of a broader family of machine learning methods based on artificial neural networks that represent learning.

What are RNNs?
They are computing systems inspired by biological neural networks that simulate animal brains.
ANN is based on a collection of units or nodes called artificial neurons.
Each neuron individually performs only a single computation.

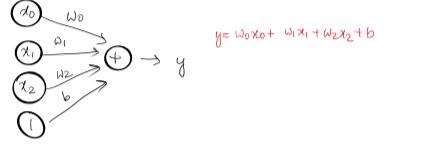
THE LINEAR UNIT
Individual component of Neural Network: A neuron.
It would look something like this:

That linear unit: $y = wx + b$ → Does this look familiar?
where:
• x is the input
• w is the weight (w is associated with the weight of connection)
• b is a special kind of weight called the bias (b does not have any input delay)
• y is the resulting output of the neuron (it sums up all the values coming)

SINGLE NEURAL MODELS ARE USELESS MODELS
To understand:
Let's say we have a dataset of 10 records.
Input: count of cups per serving - "cups"
Output: Calories per serving - "calories"
 $b = 90$
 $w = 25$ Let's assume
Using the given info, we can estimate the calorie content of a meal

$$\text{Inputs: } 2.5 \rightarrow \text{cups} \rightarrow 3b + w \cdot x \rightarrow \text{calories} \rightarrow 90 + (5 \cdot 2.5) = 102.5$$

MULTIPLE INPUTS



What is keras?
→ Open source software library that provides a Python interface for artificial neural networks.

LINER UNIT IN KERAS

How to create a model in Keras?
→ `Keras Sequential` → create a neural network as a stack of layers.
Let's look at the code:

```
from tensorflow import keras
from tensorflow.keras import layers
```

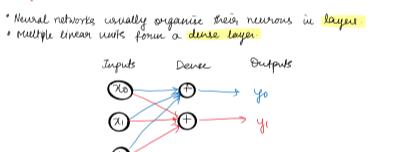

Create a network with 1 linear unit
model = keras.Sequential([
 layers.Dense(1, input_shape=(3,))]

* `units` → no. of outputs = 1 (in our case)
* `input_shape` → dimensions of the inputs

DEEP NEURAL NETWORKS

In this lesson, we will learn how we can build neural networks capable of learning the complex kinds of relationships deep neural networks are famous for.

Layers



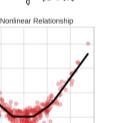
Through a deep stack of layers, a neural network can transform its inputs in more and more complex ways.

THE ACTIVATION FUNCTION

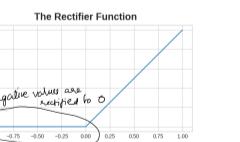
If I leave my dense layers with nothing in between, it's no better than a single dense layer.
We need something non-linear to make sure dense layers can more fully utilize the world of non-linearities.
That is where Activation Function comes in.



Without AF, neural networks can only learn linear relationships.



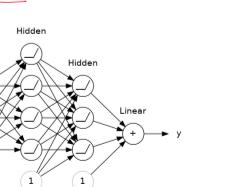
Activation Function - a function we apply to the layer of outputs.
Most common - ReLU (0,x)



When we attach the rectifier to a linear unit, we get a rectified linear unit or ReLU. (For this reason, it's common to call the rectifier function the "ReLU function".) Applying a ReLU activation to a linear unit means the output becomes $\max(0, w \cdot x + b)$, which we might draw in a diagram like:



STACKING DENSE LAYERS



BUILDING SEQUENTIAL MODELS

```
from tensorflow import keras
from tensorflow.keras import layers

model = keras.Sequential([
    # the hidden ReLU layers
    layers.Dense(units=4, activation='relu',
    input_shape=[2]),
    layers.Dense(units=4, activation='relu'),
    # the linear output layer
    layers.Dense(units=1)])
```

STOCHASTIC GRADIENT DESCENT

So far we saw what neurons and neural techniques are. We applied little pre-defined values of bias and weight. The network does not know anything yet. Thus, we are going to see how we can **train** the model.

In addition to training the network with data, we need:

- A loss function
- An optimizer

THE LOSS FUNCTION

The loss function measures the disparity between the target true value and the value the model predicts.
Different problems have different loss functions.
For regression problems, the mean absolute error or MAE is the common loss function.

THE OPTIMIZER

Now we have done setup.
To solve the problem, we want the network to solve now, we want to say how we want it to be solved. That's the job of the **optimizer**.

In a way, we can say all day learning algorithms belong to the stochastic gradient descent.

• Each iteration's sample of training data is called a minibatch.

• A complete round of training data is called epoch.

LET'S LOOK AT AN EXAMPLE!