Artificial Neural Networks (ANN) Multi Layer Perceptron (MLP)

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Multilayer Perceptron (MLP)

- Perceptron stand alone cannot solve many problems. Also simple binary problems: e.g. A XOR B!
- Solution is a layered architecture

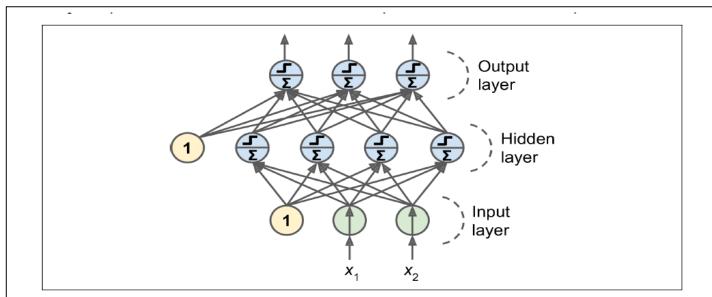


Figure 10-7. Architecture of a Multilayer Perceptron with two inputs, one hidden layer of four neurons, and three output neurons (the bias neurons are shown here, but usually they are implicit)

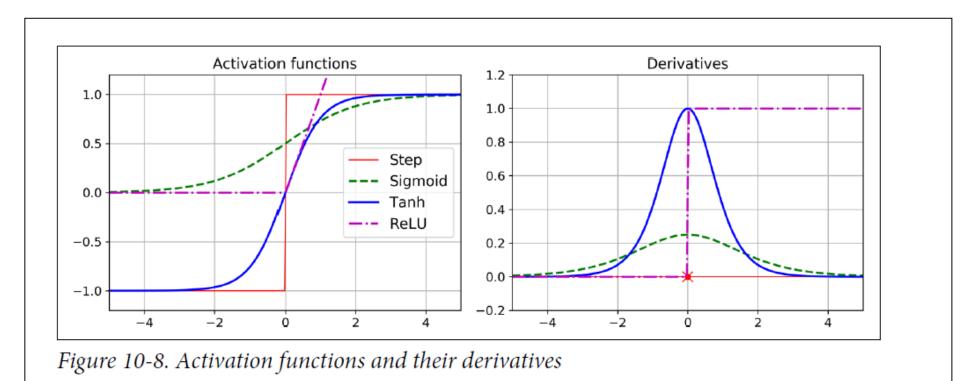
Problem: How to train? Don't worry we have backpropagation since 1986!

Backpropagation agorithm

- Take a mini-batch (typically 32 instance) one at a time
- Forward pass (prediction)
 - Send mini-batch to first layer
 - Calculate output
 - Send result to next layer
 - Finally we have the output in output layer
- Calculate the output error (using a loss function)
- Backward pass
 - Measure error gradient on connection weights
 - Performs Gradient Descent to change all connection weights
- Take next mini-batch; Repeat the steps
- Pass the training set many times called epochs
- Problem: Cannot use activation step function in Gradient Descent, Need another activation function!

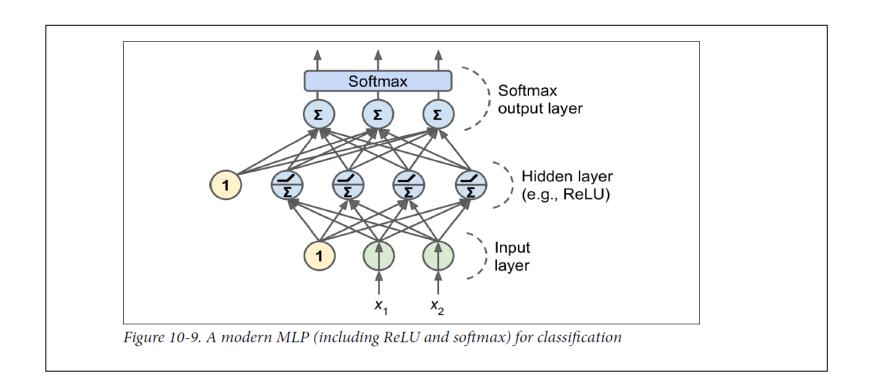
Activation functions

- Many types, 3 popular types in use
 - ReLU: Rectified Linear Units (maybe Softplus version)
 - Sigmoid, Sigmoid function from logistic regression (classification)
 - Tanh: Tangent hyperbolic



Multilayer Perceptron (MLP): Modern Classification

- Input layer with features with bias 1
- One or more hidden layers with neurons, bias 1 and using ReLU: Recitfied Linear Units
- Output layer with Sigmoid function (probability 0-1) and one neuron per class



MINST Fashion Dataset

- 10 classes: sneaker, coat,
- Feature is a 28x28 pixel picture
- 70.000 pictures: 60.000 in training set 10.000 test set



Keras model: Sequential Model Code

- Load the data
- Set up validation set for the epochs
- Flatten 2-dimensional matrix (array) to 1-dim array
- Create a model with two inner layers and one output layer
- Dense all neurons connected to next layer's neurons

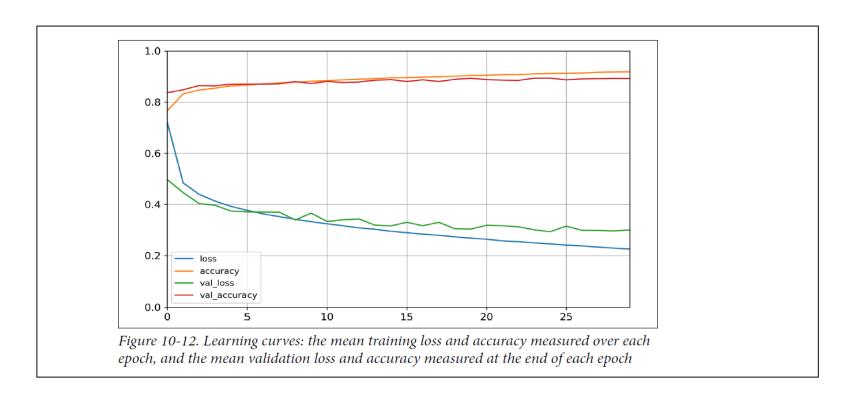
```
X_valid, X_train = X_train_full[:5000] / 255.0, X_train_full[5000:] / 255.0
y_valid, y_train = y_train_full[:5000], y_train_full[5000:]

model = keras.models.Sequential([
    keras.layers.Flatten(input_shape=[28, 28]),
    keras.layers.Dense(300, activation="relu"),
    keras.layers.Dense(100, activation="relu"),
    keras.layers.Dense(10, activation="softmax")
])
```

Now let's look at the whole program code

Learning curves: Accuracy and loss

- Accuracy: percentage correct
- Loss: Cross-Entropy (Chapter 4) for classification, MSE for regression



Fine tuning hyperparameters

- Number of layers
- Number of neurons
- Neurons distribution
- Activation function

Normally more layers better than many neurons Neurons more in start than later, Claudius' favourite: 200, 50, 50 10 ReLU is standard but you never know!

Code and Assignment!

- First a small round tour in the code,
- Then more assignments
- Later at home, MAYBE relax with a video or two

Assignments second round

- It is time for discussion and solving a few exercises in groups
- MLP Classfication Fashion Exercise
- Playground at tensorflow.org

