

Deep Learning

Big Data & Machine Learning Bootcamp - Keep Coding



Outline

1. Mini batch gradient descent
2. Gradient descent with momentum
3. Adam optimization algorithm
4. Hyperparameters
5. Softmax Regression
6. Deep Learning Frameworks

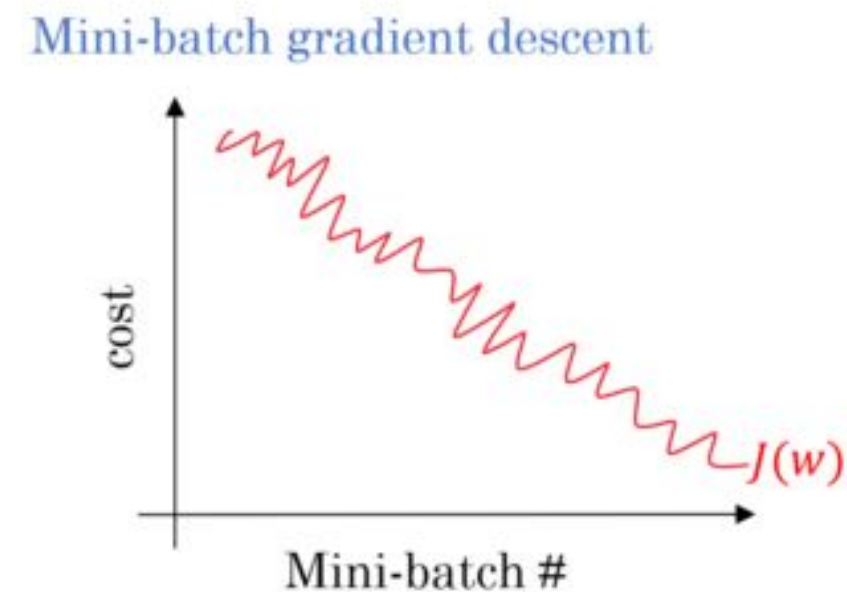
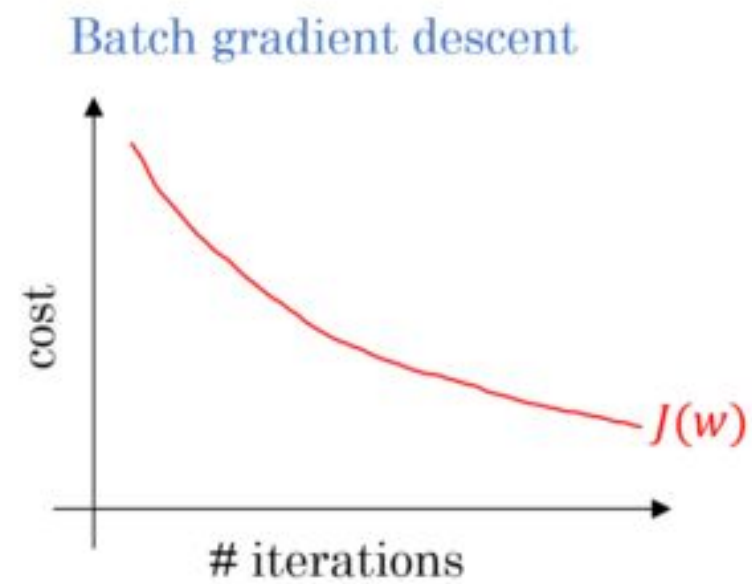


Mini batch gradient descent

One epoch is a single pass through the training set!

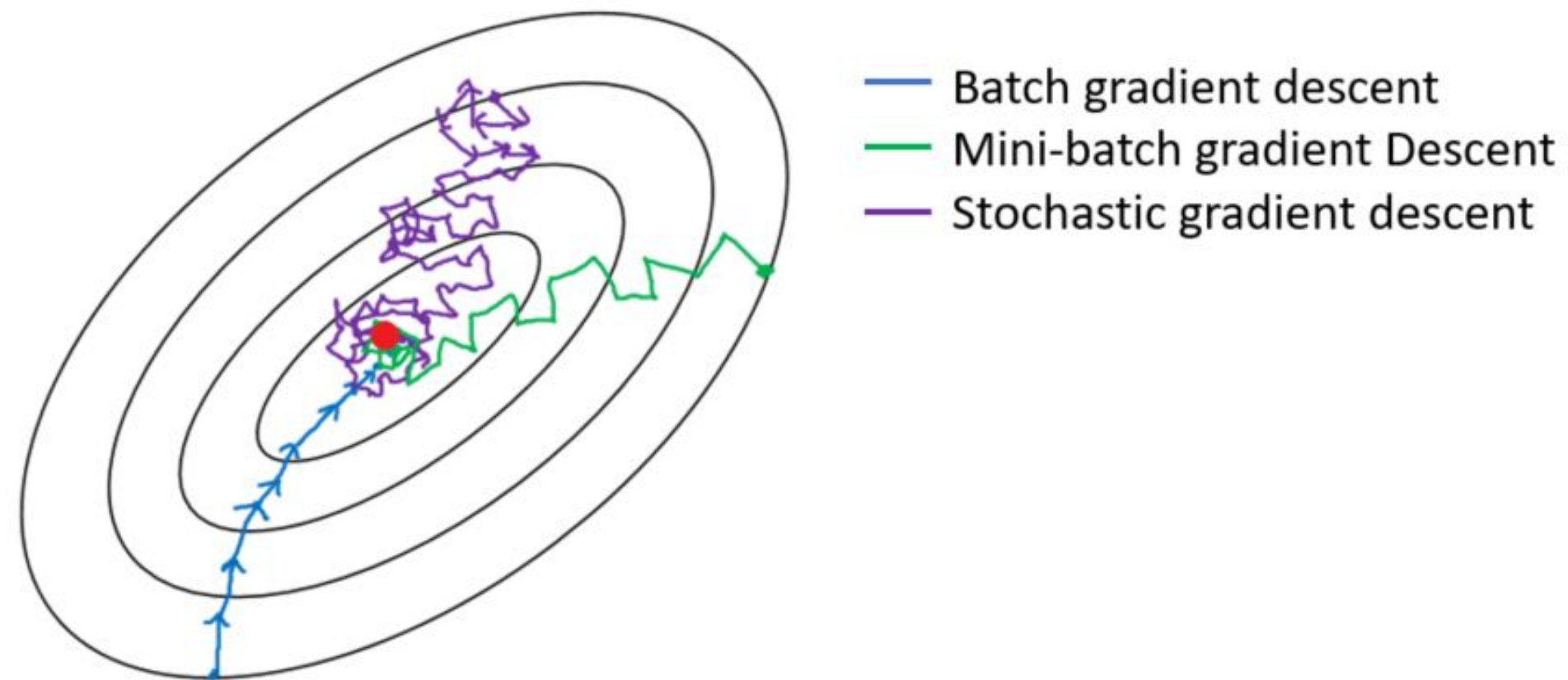
Using batch gradient descent, you only take one step per epoch.

Using mini batch gradient descent, you take #batches steps per epoch

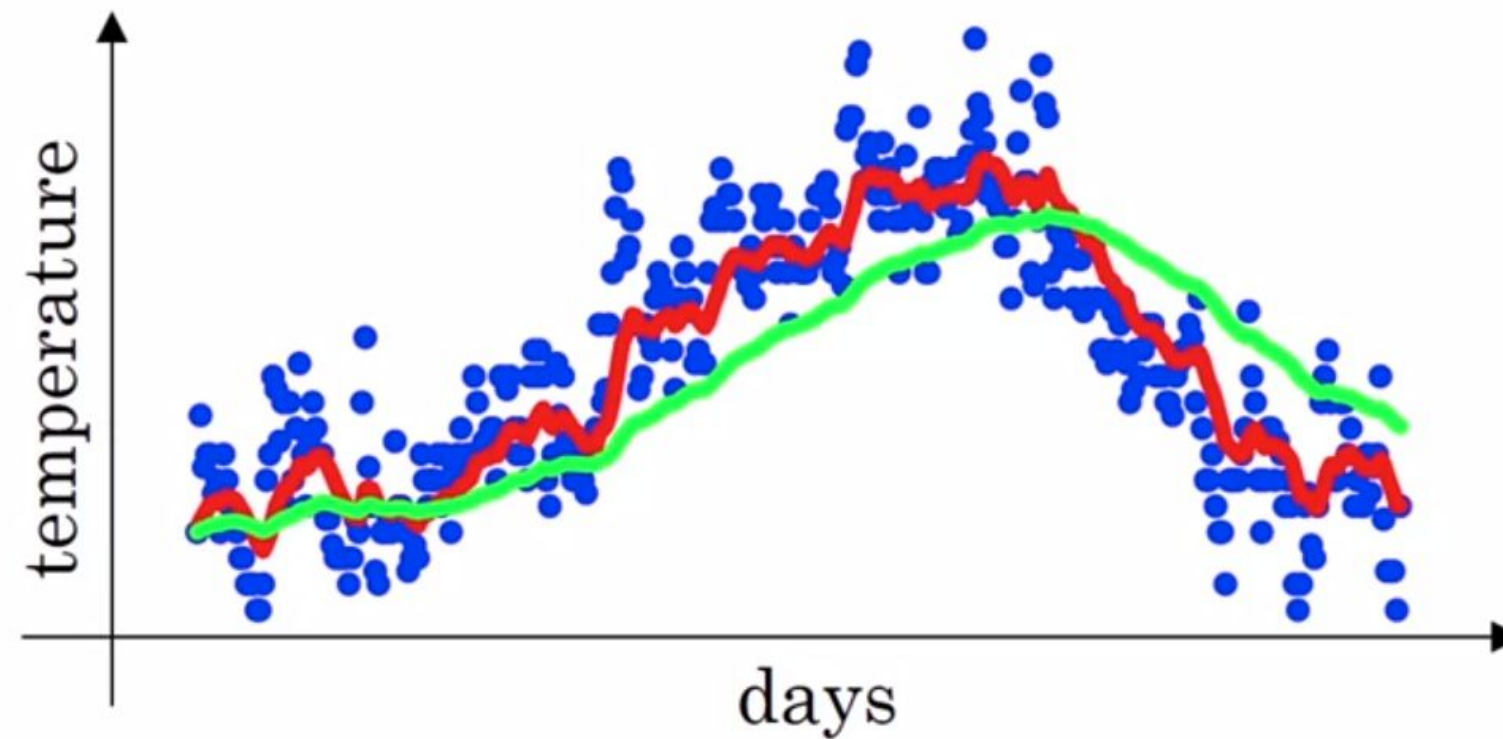


Mini batch gradient descent

Fastest learning occurs when using stochastic gradient descent!



Gradient descent with momentum



$$v_t = \beta v_{t-1} + (1 - \beta)\theta_t$$

Blue dots: Data

*Red line: $\beta = 0.9$ (**Intuition:** You focused on the last 10 values)*

*Blue line: $\beta = 0.98$ (**Intuition:** You focused on the last 50 values)*



Sources:
- Coursera

Gradient descent with momentum

On iteration t :

Compute dW, db on the current mini-batch

$$v_{dW} = \beta v_{dW} + (1 - \beta) dW$$

$$v_{db} = \beta v_{db} + (1 - \beta) db$$

$$W = W - \alpha v_{dW}, \quad b = b - \alpha v_{db}$$

Hyperparameters: α, β $\beta = 0.9$



Adam optimization algorithm

Adam or Adaptive Moment Estimation is probably the most used optimization algorithm!

It is basically a **combination of RMSProp and gradient descent with momentum** optimization algorithms!

For each Parameter w^j
(j subscript dropped for clarity)

$$\nu_t = \beta_1 * \nu_{t-1} - (1 - \beta_1) * g_t$$

$$s_t = \beta_2 * s_{t-1} - (1 - \beta_2) * g_t^2$$

$$\Delta\omega_t = -\eta \frac{\nu_t}{\sqrt{s_t + \epsilon}} * g_t$$

$$\omega_{t+1} = \omega_t + \Delta\omega_t$$

Sorry for the change in the nomenclature here :)

The essence is that there are three hyperparameters when using Adam: learning rate, β_1 and β_2 .

β_1 is commonly set to 0.9 and β_2 is commonly set to 0.99

η : Initial Learning rate

g_t : Gradient at time t along ω^j

ν_t : Exponential Average of gradients along ω_j

s_t : Exponential Average of squares of gradients along ω_j

β_1, β_2 : Hyperparameters



Hyperparameters

So far we have talked about many hyperparameters:

- Learning rate
- Momentum (~ 0.9)
- β_1, β_2 (Adam optimizer)
- Mini batch size
- Number of layers
- Number of hidden units
- Learning rate decay

Color means importance of the hyperparameters (Red: very important, Orange: mid importance and Green: low importance)

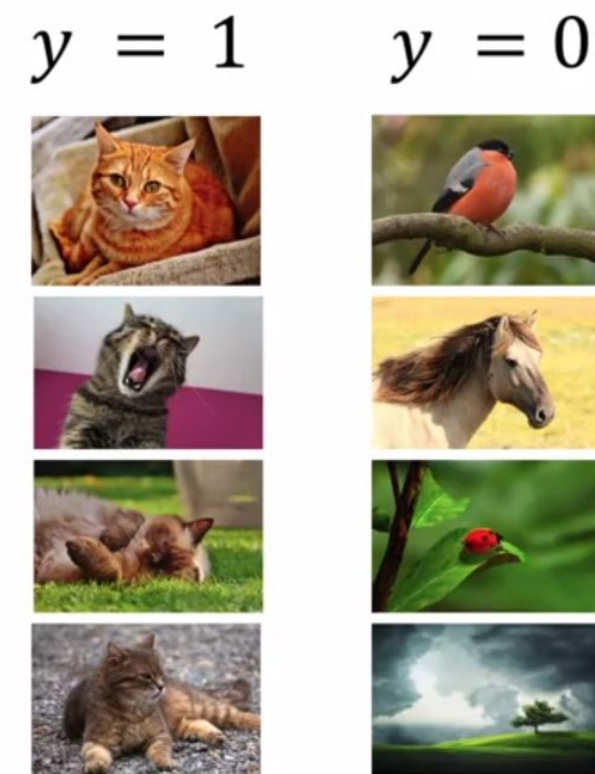
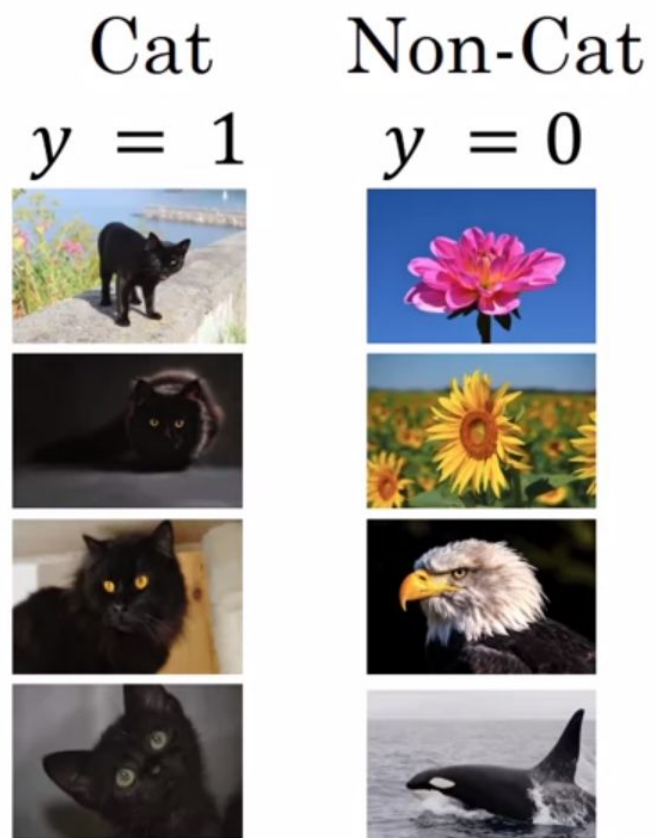


Sources:
- Coursera

Hyperparameters - Batch Normalization

But why **Batch Normalization (BN)** helps? Basically it makes the neural network robust to “**covariate shift**”

Say you train a neural network on black cats only. If we don't use BN, the performance of the system will be bad when testing on cats of different colors!



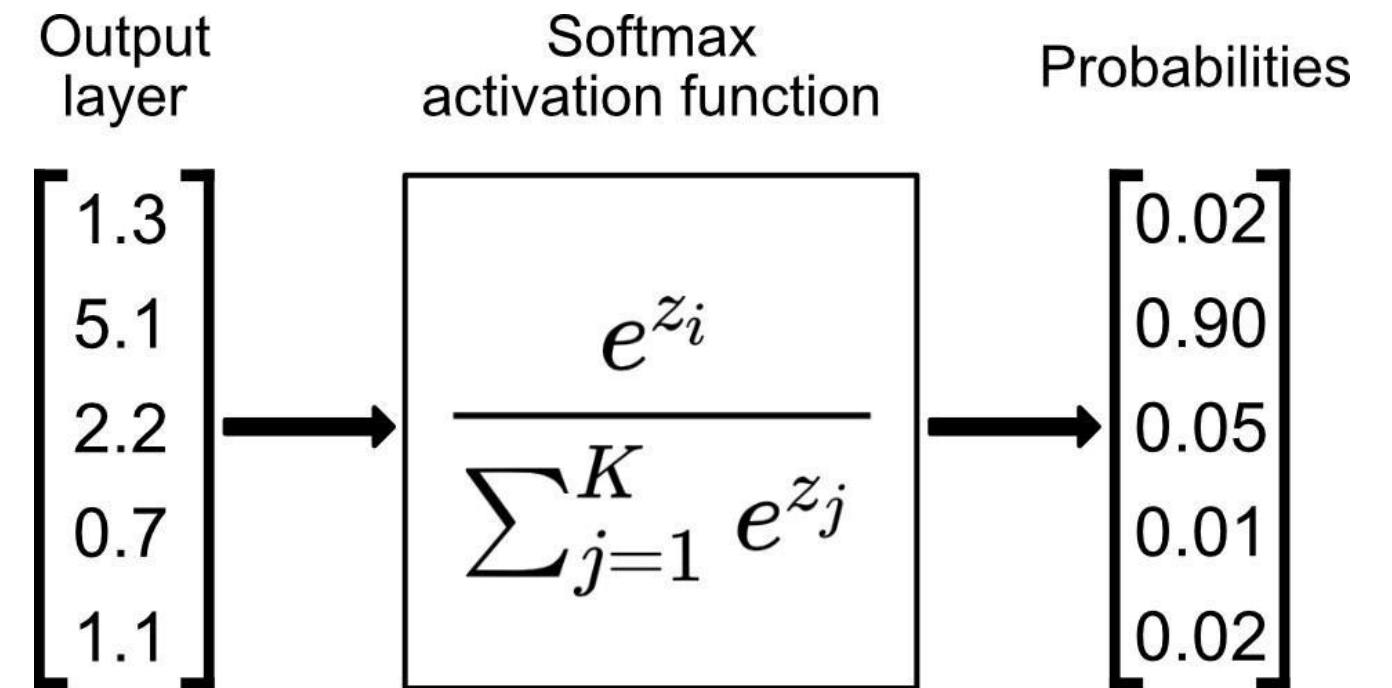
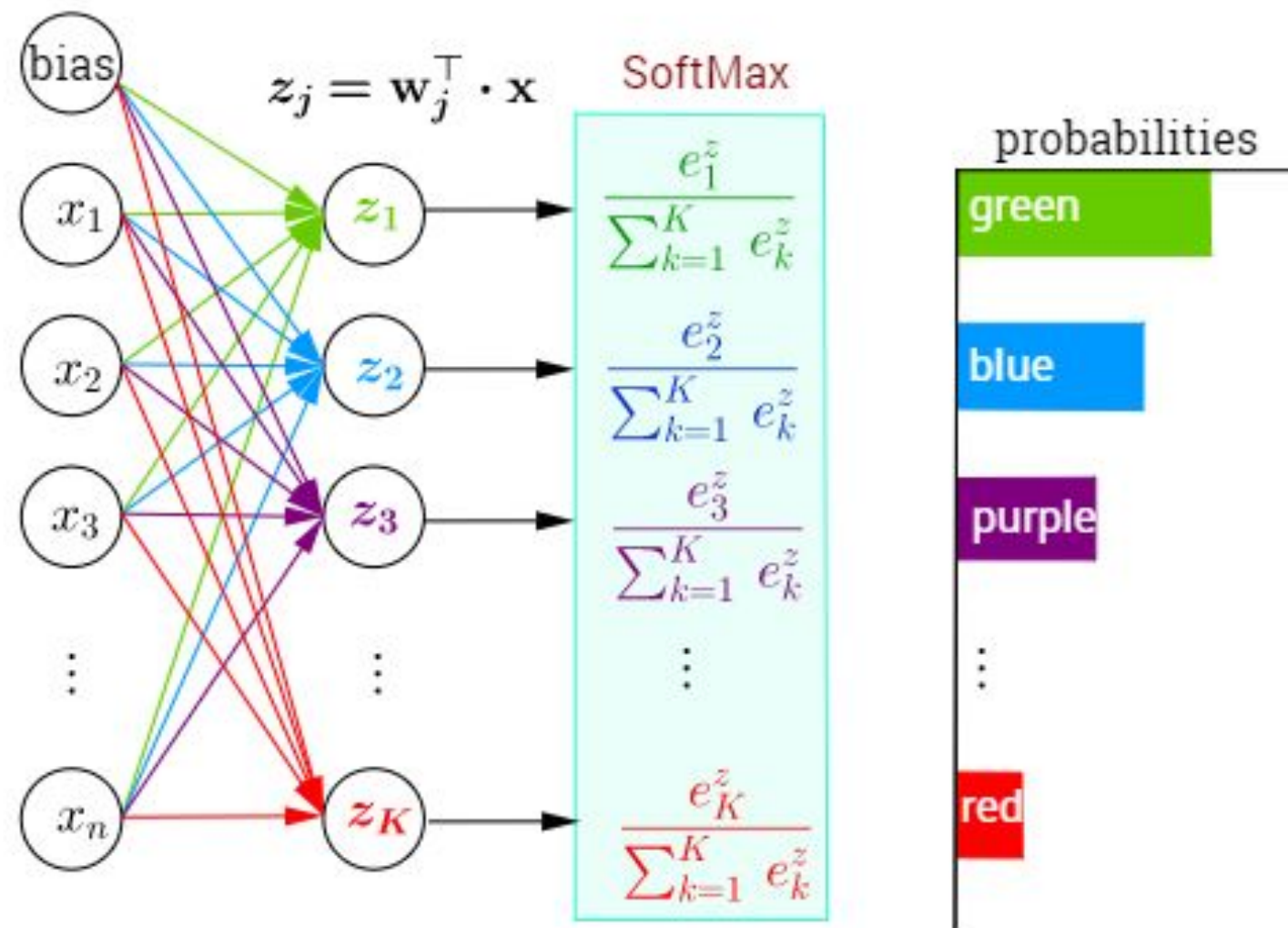
Sources:

- Coursera

- Ioffe, S. and Szegedy, C., 2015. Batch normalization: accelerating deep network training by reducing internal covariate shift.

Softmax Regression

The Softmax layer outputs probabilities:



- K is the number of classes!
- The output probabilities sum up to 1 as a probability distro does



Sources:

- <https://www.pinterest.co.uk/pin/849702654687897076/>
- <https://towardsdatascience.com/softmax-activation-function-explained-a7e1bc3ad60>

Softmax Regression

How do you encode the labels/classes when using softmax classification?

One-hot encoding

id	color
1	red
2	blue
3	green
4	blue



id	color_red	color_blue	color_green
1	1	0	0
2	0	1	0
3	0	0	1
4	0	1	0



Deep Learning Frameworks

The criteria to choose a deep learning framework is:

- Ease of programming (development and deployment)
- Running speed
- Open source

TensorFlow: <https://www.tensorflow.org/>

PyTorch: <https://pytorch.org/>

Keras: <https://keras.io/>

FastAI: <https://www.fast.ai/>

New GitHub Activity

