

Basics of Neural Network Programming

Vectorization

What is vectorization?

$$for i in ray (n-x):$$

$$2+= \omega [1] * x (1)$$



Basics of Neural Network Programming

More vectorization examples

Neural network programming guideline

Whenever possible, avoid explicit for-loops.

Neural network programming guideline

Whenever possible, avoid explicit for-loops.

$$U = AV$$

$$U_{i} = \sum_{j} \sum_{i} A_{i,j} V_{j}$$

$$U = np.zeros((n, i))$$

$$for i \dots \in ACITiJ*vC_{i}J$$

$$uCiJ += ACITC_{i}J*vC_{i}J$$

Vectors and matrix valued functions

Say you need to apply the exponential operation on every element of a matrix/vector.

$$v = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \rightarrow \mathbf{u} = \begin{bmatrix} \mathbf{e}^{\mathbf{v}_1} \\ \mathbf{e}^{\mathbf{v}_2} \end{bmatrix}$$

$$v = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \rightarrow u = \begin{bmatrix} e^{v_1} \\ e^{v_n} \end{bmatrix}$$

$$u = np \cdot exp(v) \leftarrow 1$$

$$np \cdot log(v)$$

$$np \cdot als(v)$$

$$np \cdot als(v)$$

$$np \cdot haximum(v, 0)$$

$$np \cdot haximum(v, 0)$$

$$v \neq v = 1$$

$$v = np \cdot exp(v)$$

$$np \cdot als(v)$$

$$np \cdot haximum(v, 0)$$

$$v \neq v \neq 1$$

Logistic regression derivatives

$$J = 0, \quad dw1 = 0, \quad dw2 = 0, \quad db = 0$$

$$\Rightarrow \text{ for } i = 1 \text{ to } n:$$

$$z^{(i)} = w^{T}x^{(i)} + b$$

$$a^{(i)} = \sigma(z^{(i)})$$

$$J += -[y^{(i)} \log \hat{y}^{(i)} + (1 - y^{(i)}) \log(1 - \hat{y}^{(i)})]$$

$$dz^{(i)} = a^{(i)}(1 - a^{(i)})$$

$$dw_{1} += x_{1}^{(i)}dz^{(i)}$$

$$dw_{2} += x_{2}^{(i)}dz^{(i)}$$

$$db += dz^{(i)}$$

$$J = J/m, \quad dw_{1} = dw_{1}/m, \quad dw_{2} = dw_{2}/m, \quad db = db/m$$

$$d\omega /= m$$



Basics of Neural Network Programming

Broadcasting in Python

Broadcasting example

Calories from Carbs, Proteins, Fats in 100g of different foods:

Apples Beef Eggs Potatoes

Carb
$$56.0$$
 0.0 4.4 68.0 1.2 104.0 52.0 8.0 1.8 135.0 99.0 0.9 13.4 135.0

Broadcasting example

$$\begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix} + \begin{bmatrix} 100 \\ 100 \\ 100 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} + \begin{bmatrix} 100 & 200 & 300 \\ 100 & 200 & 300 \end{bmatrix}$$

$$(m,n) \xrightarrow{(2,3)} (m,n) \xrightarrow{(2,3)} (m,n) \xrightarrow{(2,3)}$$

$$\begin{bmatrix}
1 & 2 & 3 \\
4 & 5 & 6
\end{bmatrix} +
\begin{bmatrix}
100 & 60 & 60 \\
200 & 60 & 60
\end{bmatrix} =
\begin{pmatrix}
(m, 1) & 6 & 6 \\
(m, n) & 6 & 6
\end{pmatrix}$$

Andrew Ng

General Principle

$$(M, n) \qquad + \qquad (1, n) \qquad \sim (M, n)$$

$$\frac{1}{modn'} \qquad + \qquad (M, 1) \qquad \sim (M, n)$$

$$(M, 1) \qquad + \qquad R$$

$$\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} \qquad + \qquad 100 \qquad = \begin{bmatrix} 101 \\ 102 \\ 103 \end{bmatrix}$$

$$\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} \qquad + \qquad 100 \qquad = \begin{bmatrix} 101 \\ 102 \\ 103 \end{bmatrix}$$

Mostlab/Octave: bsxfun



Basics of Neural Network Programming

A note on python/ numpy vectors

Python Demo

Python / numpy vectors

```
import numpy as np
a = np.random.randn(5)
a = np.random.randn((5,1))
a = np.random.randn((1,5))
assert(a.shape = (5,1))
```