



deeplearning.ai

# Basics of Neural Network Programming

---

## Vectorization

Vectorization is basically the art of getting rid of explicit for loops in your code. I

think the ability to perform vectorization has become a key skill

# What is vectorization?

$$z = \underbrace{w^T x}_{\text{dot product}} + b$$

$$w = \begin{bmatrix} \vdots \\ \vdots \\ \vdots \end{bmatrix} \quad x = \begin{bmatrix} \vdots \\ \vdots \\ \vdots \end{bmatrix}$$

$$w \in \mathbb{R}^{n_x}$$

$$x \in \mathbb{R}^{n_x}$$

Non-vectorized:

$$z = 0$$

for i in range(n-x):

$$z += w[i] * x[i]$$

$$z += b$$

Vectorized

$$z = \underbrace{\text{np.dot}(w, x)}_{w^T x} + b$$

parallelization instructions

→ GPU } SIMD - single instruction  
→ CPU } multiple data.

If you use built-in functions such as this np.function or other functions that don't require you explicitly implementing a for loop. It enables Python numpy to take much better advantage of parallelism to do your computations much faster. And this is true both computations on CPUs and computations on GPUs.



deeplearning.ai

# 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.

$$u = Av$$

$$u_i = \sum_j A_{ij} v_j$$

$$u = \text{np.zeros}(n, 1)$$

for i ... ←

for j ... ←

$$u[i] += A[i][j] * v[j]$$

$$u = \text{np.dot}(A, v)$$

# Vectors and matrix valued functions

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

$$v = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \rightarrow u = \begin{bmatrix} e^{v_1} \\ e^{v_2} \\ \vdots \\ e^{v_n} \end{bmatrix}$$

```
→ u = np.zeros((n,1))  
→ for i in range(n):  
    → u[i]=math.exp(v[i])
```

```
import numpy as np  
u = np.exp(v)  
  
np.log(v)  
np.abs(v)  
np.maximum(v, 0)  
v**2  
1/v
```

# Logistic regression derivatives

$$J = 0, \quad \boxed{\cancel{dw_1 = 0, dw_2 = 0}}, \quad db = 0$$

$$dw = np.zeros((n-x, 1))$$

→ for i = 1 to n:

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

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

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

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

$$\cancel{dw_1 += x_1^{(i)} dz^{(i)}}$$

$$\cancel{dw_2 += x_2^{(i)} dz^{(i)}}$$

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

$$n_x = 2$$

$$dw += x^{(i)} dz^{(i)}$$

$$J = J/m, \quad \boxed{\cancel{dw_1 = dw_1/m, dw_2 = dw_2/m}}, \quad db = db/m$$

$$dw /= m.$$