

Logistic regression

lab3

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- Problem: Recognizing cat pictures.
- Algorithm: Logistic Regression with neural network mindset
- Performance matrix: Accuracy

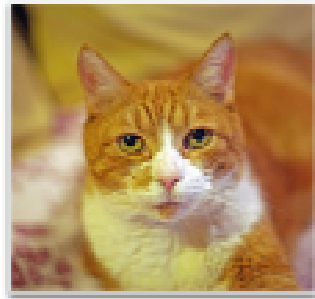
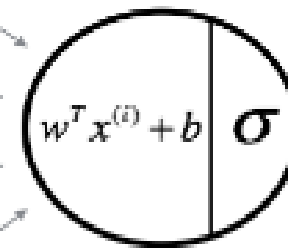
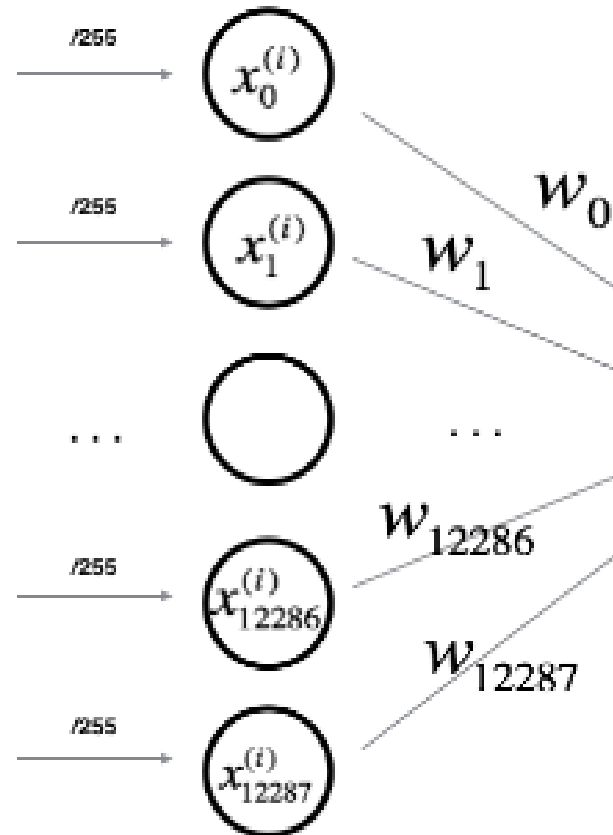


image2vector

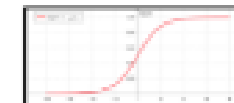
$\begin{pmatrix} 255 \\ 231 \\ \dots \\ 94 \\ 142 \end{pmatrix}$



0.73

$0.73 > 0.5$

"it's a cat"



Mathematical expression of the algorithm:

For one example $x^{(i)}$:

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

$$\hat{y}^{(i)} = a^{(i)} = \text{sigmoid}(z^{(i)})$$

$$\mathcal{L}(a^{(i)}, y^{(i)}) = -y^{(i)} \log(a^{(i)}) - (1 - y^{(i)}) \log(1 - a^{(i)})$$

The cost is then computed by summing over all training examples:

$$J = \frac{1}{m} \sum_{i=1}^m \mathcal{L}(a^{(i)}, y^{(i)})$$

Forward Propagation:

- You get X
- You compute $A = \sigma(w^T X + b) = (a^{(1)}, a^{(2)}, \dots, a^{(m-1)}, a^{(m)})$
- You calculate the cost function: $J = -\frac{1}{m} \sum_{i=1}^m y^{(i)} \log(a^{(i)}) + (1 - y^{(i)}) \log(1 - a^{(i)})$

Here are the two formulas you will be using:

$$\frac{\partial J}{\partial w} = \frac{1}{m} X(A - Y)^T$$
$$\frac{\partial J}{\partial b} = \frac{1}{m} \sum_{i=1}^m (a^{(i)} - y^{(i)})$$

- optimization function:

The goal is to learn w and b by minimizing the cost function J. For a parameter θ, the update rule

$$\underline{\theta} = \underline{\theta} - \underline{\alpha} \underline{d\theta},$$

where α is the learning rate.

- **Implement :**
 - Apply logistic regression as explained on *Titanic Passengers Dataset.*

