Ciência de Dados (Big Data Processing and Analytics)

Big Data Analytics – Mineração e Análise de Dados

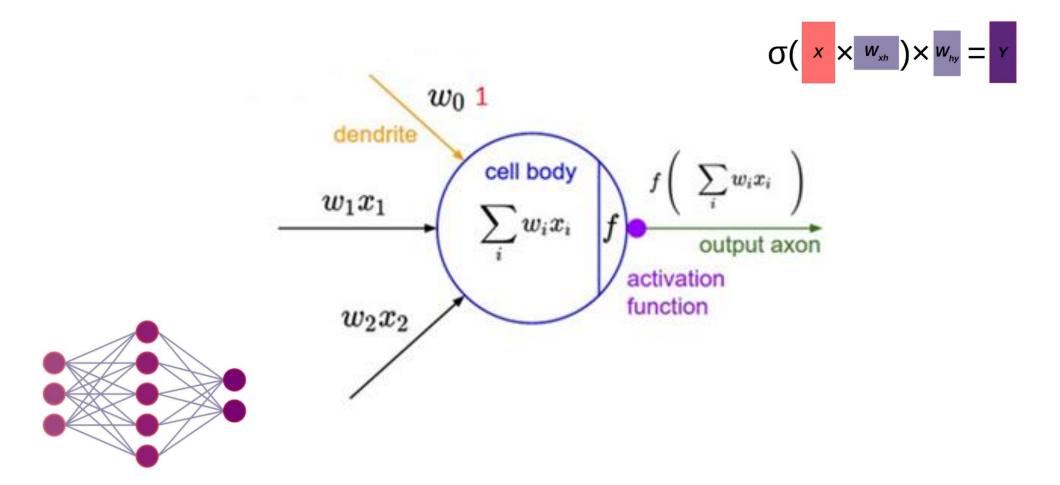




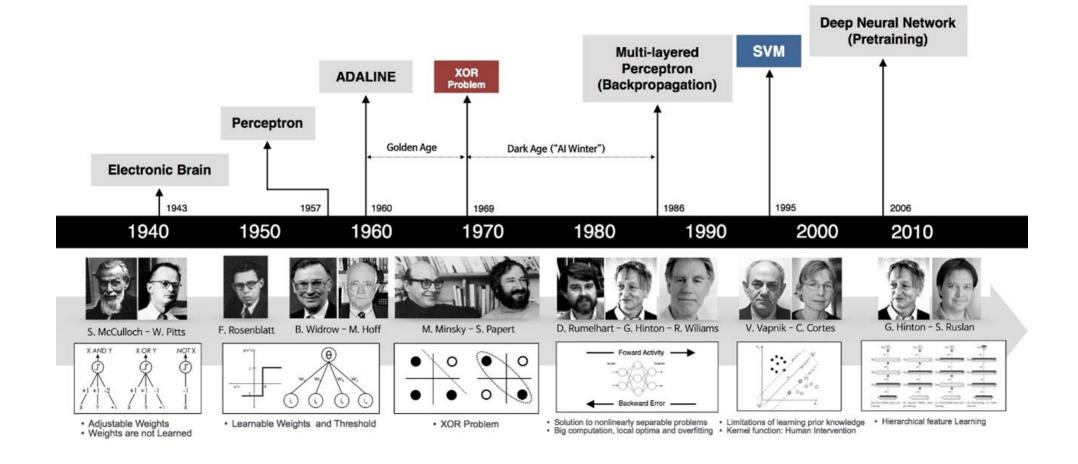
TRILHA 8 Redes Neurais e Deep Learning

Parte A

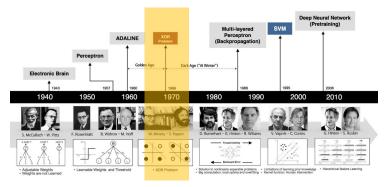
Neurônio Artificial

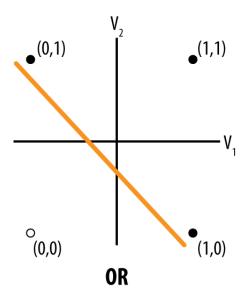


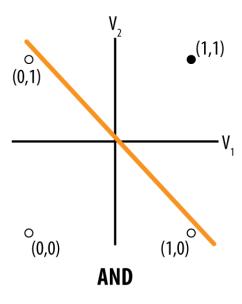
Redes Neurais: História

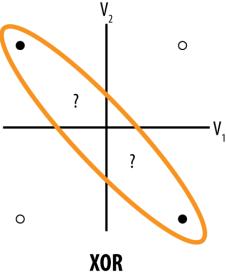


Redes Neurais: História

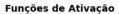


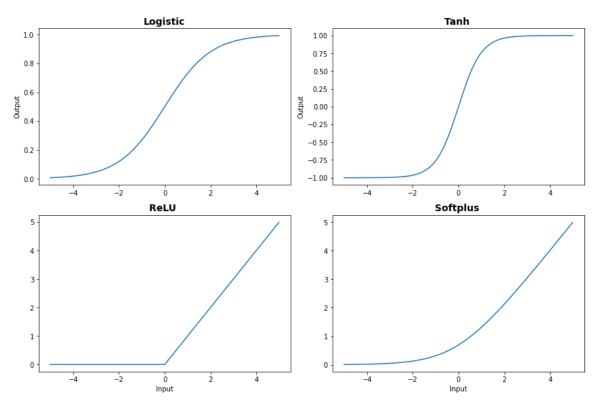


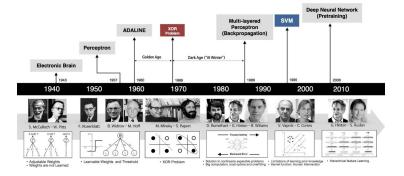




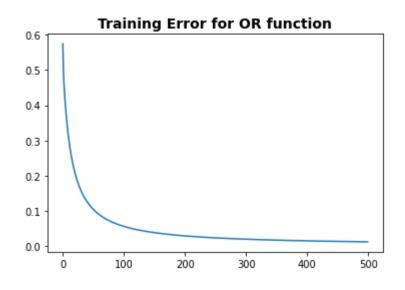
Funções de Ativação



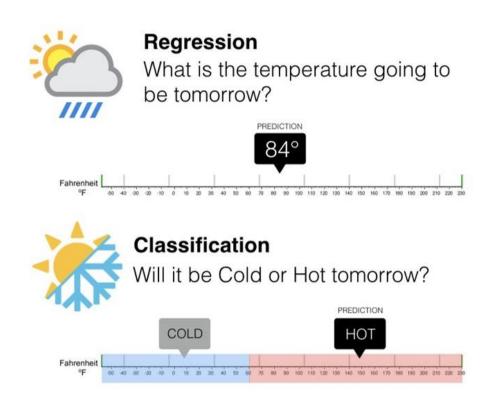




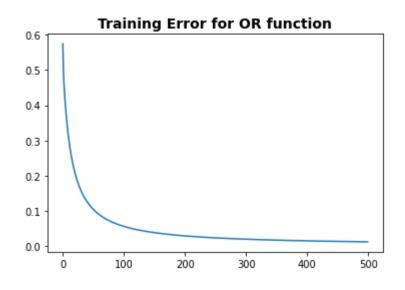
Aprendizado da Rede



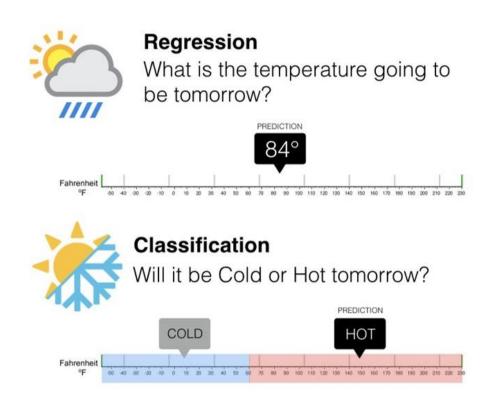
$$\sigma(x \times w_{xh}) \times w_{hy} = y$$



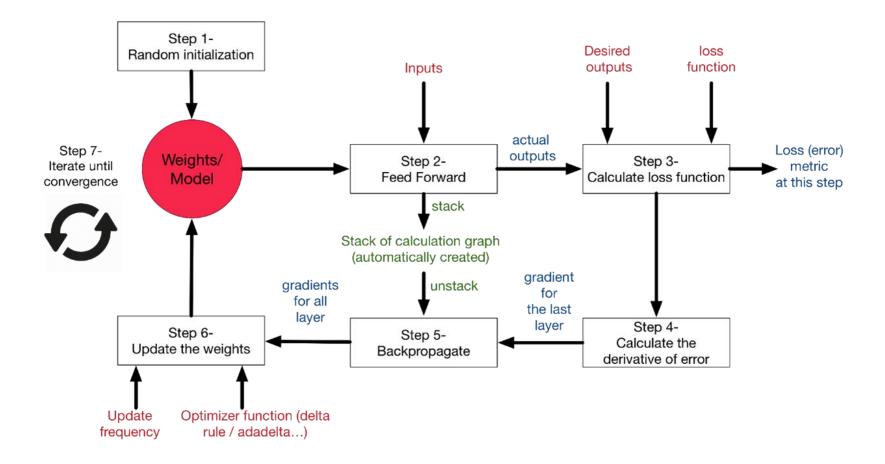
Aprendizado da Rede



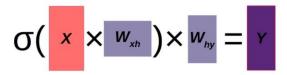
$$\sigma(x \times w_{xh}) \times w_{hy} = y$$

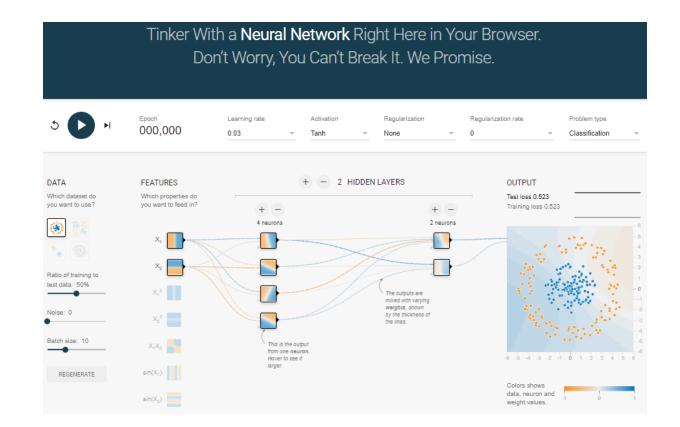


Backpropagation



Experimente

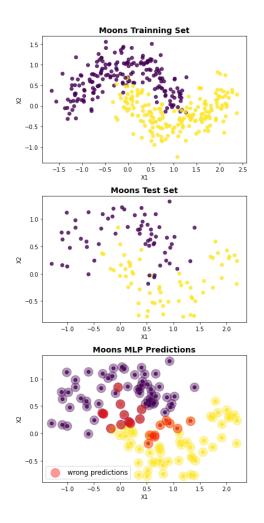




TRILHA 8 Redes Neurais e Deep Learning

Parte B

MLP Com o scikit-learn



```
Input Layer

Hidden Layer

Output Layer

Input node 1

Input node 2

Input node 2

Input node 2

Input node 3

Output 1

Output 1

Output 1

Output 2
```

```
from sklearn.neural_network import MLPClassifier
from sklearn.model_selection import train_test_split
from sklearn.datasets import make_moons

X, y = make_moons(n_samples=500, noise=0.25, random_state=1234)

X_train, X_test, y_train, y_test = train_test_split(X, y, stratify=y, random_state=1234)

mlp = MLPClassifier(activation='tanh', hidden_layer_sizes=(12, 12), random_state=1234)

mlp.fit(X_train, y_train)

y_pred = mlp.predict(X_test)

print("Accuracy on training set: {:.2f}".format(mlp.score(X_train, y_train)))
print("Accuracy on test set: {:.2f}".format(mlp.score(X_test, y_test)))
```

Frameworks de Deep Learning

	Keras	Pytorch	TensorFlow
API Level	High	Low	High and Low
Architecture	Simple, concise, readable	Complex, less readable	Not easy to use
Datasets	Smaller datasets	Large datasets, high performance	Large datasets, high performance
Debugging	Simple network, so debugging is not often needed	Good debugging capabilities	Difficult to conduct debugging
Does It Have Trained Models?	Yes	Yes	Yes
Popularity	Most popular	Third most popular	Second most popular
Speed	Slow, low performance	Fast, high-performance	Fast, high-performance
Written In	Python	Lua	C++, CUDA, Python

Keras Deep Learning

Modelo Sequencial

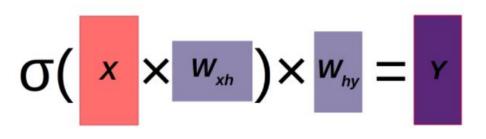
A construção do modelo segue basicamente os seguintes passos:

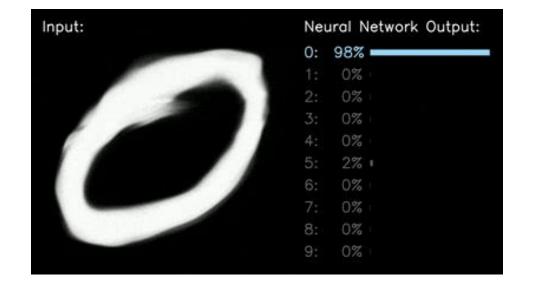
- 1. Defina o modelo.
- 2. Compile o modelo.
- 3. Treine o modelo.
- 4. Avalie o modelo.
- 5. Faça Predições.

Keras Deep Learning

```
# Define o Modelo
model = keras.Sequential(layers.Dense(X_train.shape[1], activation='relu', input_shape=[X_train.shape[1]])) # Entrada
model.add(layers.Dense(16, activation='relu'))
model.add(layers.Dense(y train.shape[1], activation='sigmoid')) # Saída
# Compila o Modelo
                                                                                                                                    loss
model.compile(loss='binary crossentropy', # Multiclass loss
                                                                                       0.70
                                                                                                                                    val loss
              optimizer='adam',
              metrics=['binary_accuracy'])
                                                                                       0.68
                                                                                       0.66
# Treina o Modelo
                                                                                       0.64
history = model.fit(
                                                                                       0.62
   X train, y train,
   batch_size=32,
                                                                                       0.60
   validation split=0.3,
   epochs=50,
                                                                                       0.58
   verbose=1,
                                                                                                     10
                                                                                                              20
                                                                                                                       30
```

Aplicações Redes Neurais





Aplicações Deep Learning

$$\sigma(x \times w_{xh}) \times w_{hy} = x$$



 Uma câmera instalada no veículo captura imagens do ambiente à sua volta. Essas imagens são registradas em uma malha bidimensional de sensores de luz



Cada um desses elementos armazena um valor numérico binário (010101, por exemplo), que corresponde à luminosidade incidente sobre eles



Essa imagem transformada em números é analisada para a identificação de áreas de interesse, eliminando elementos dispensáveis, como, por exemplo, o céu



A última etapa envolve a análise, a classificação e o reconhecimento dos dados que importam, como pessoas, outros carros, placas e sinais de trânsito

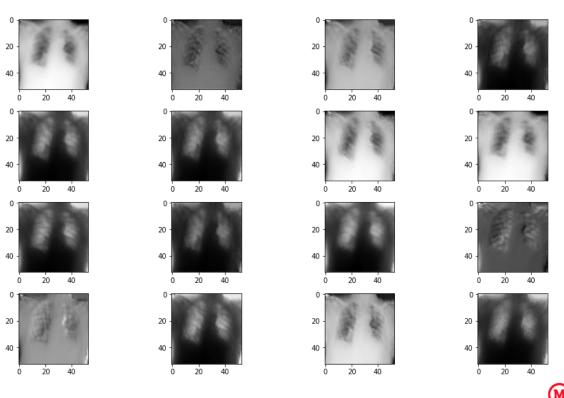


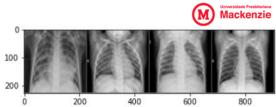
O sistema faz isso com auxílio de redes neurais que reconhecem padrões a partir de um treinamento inicial e são capazes de evoluir a partir da experiência



Aplicações Deep Learning

$$\sigma(x \times w_{xh}) \times w_{hy} = x$$





Real: VIRAL NORMAL NORMAL NORMAL

Aplicações Deep Learning

