

# **Ciência de Dados (Big Data Processing and Analytics)**

Big Data Analytics – Mineração e Análise de Dados



**Professor curador**  
Prof. Dr. Rogério de Oliveira





# **TRILHA 8**

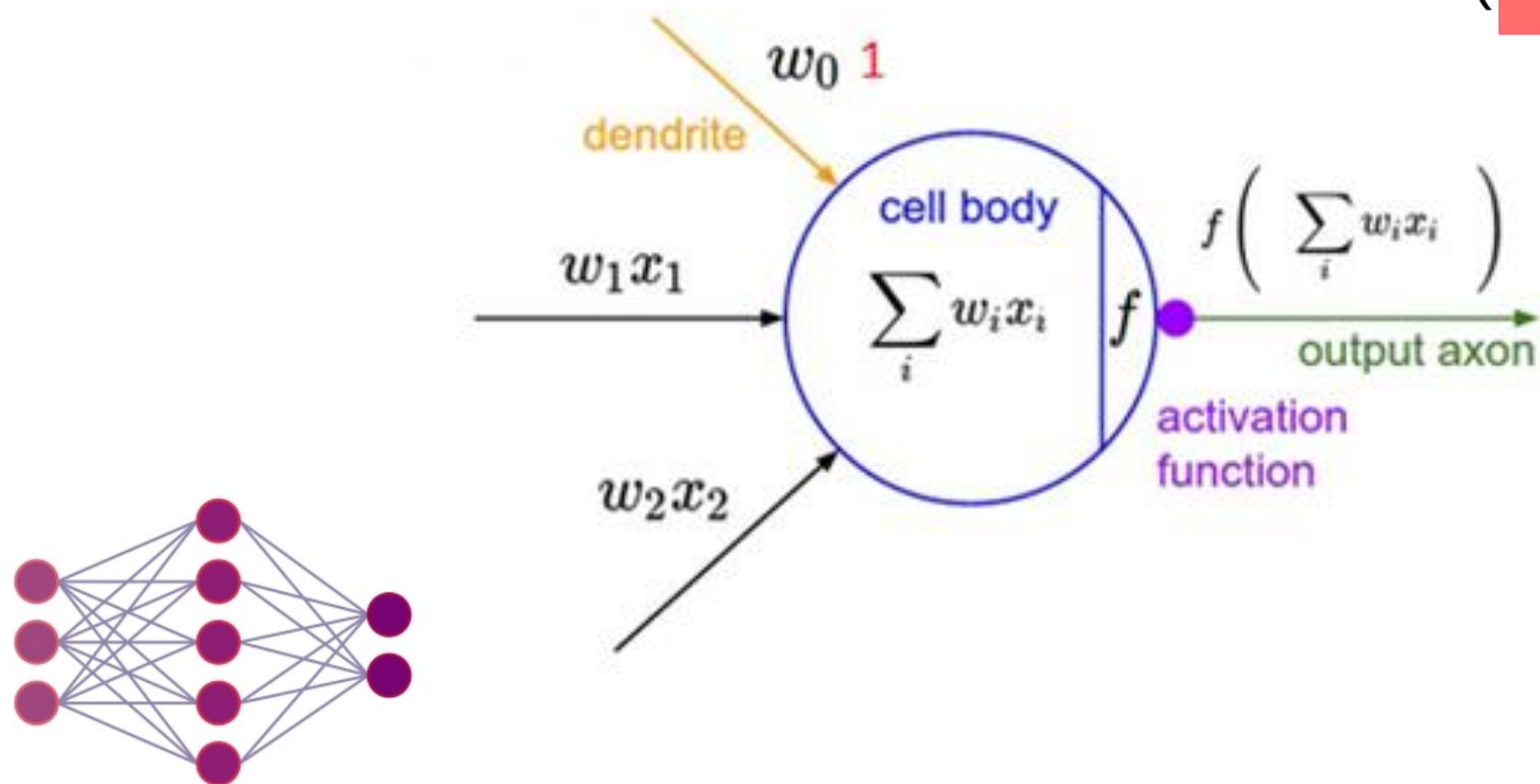
## **Redes Neurais e Deep Learning**

**Parte A**

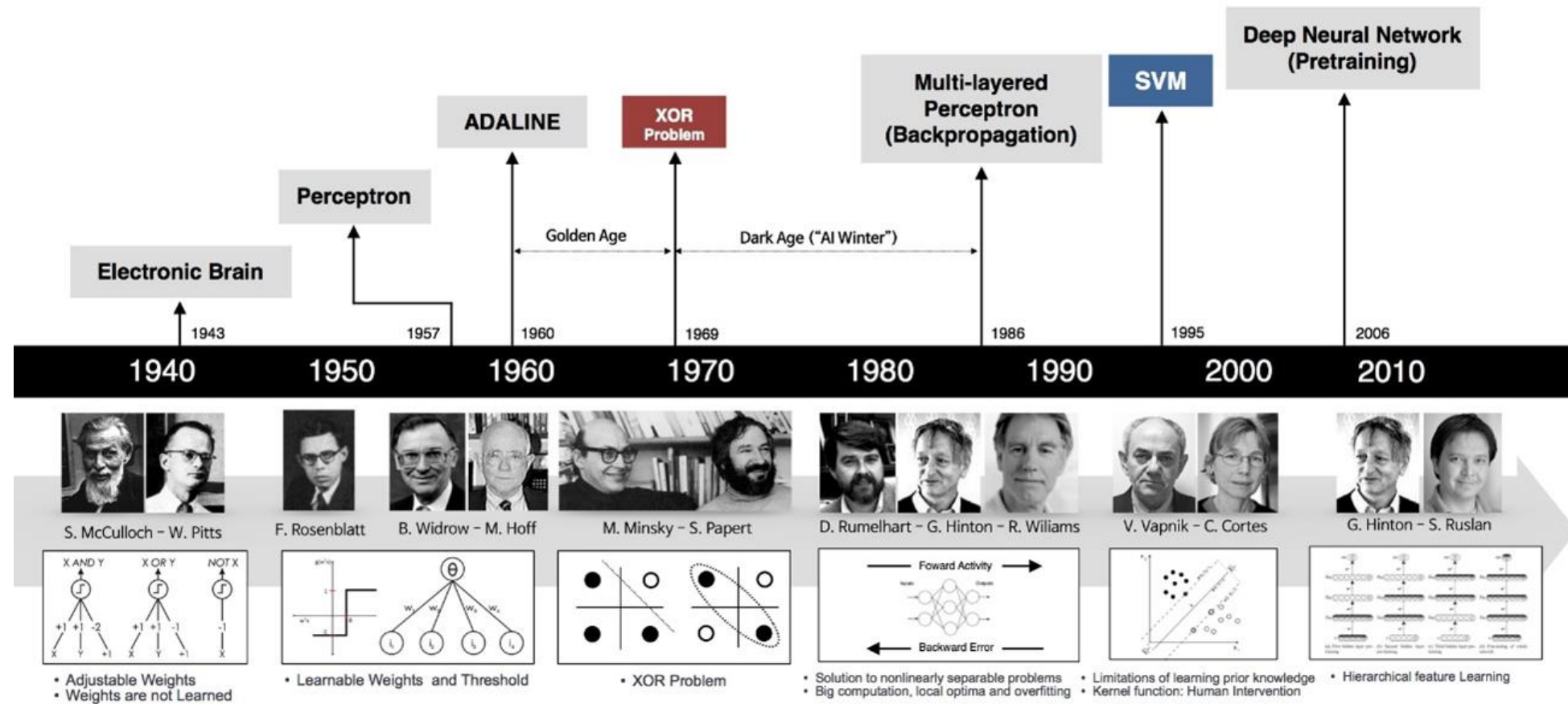
---

# Neurônio Artificial

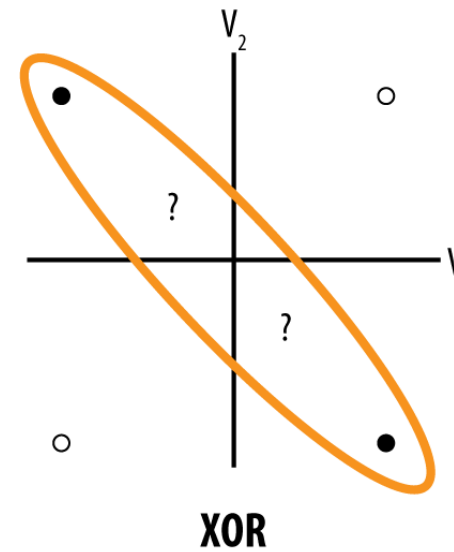
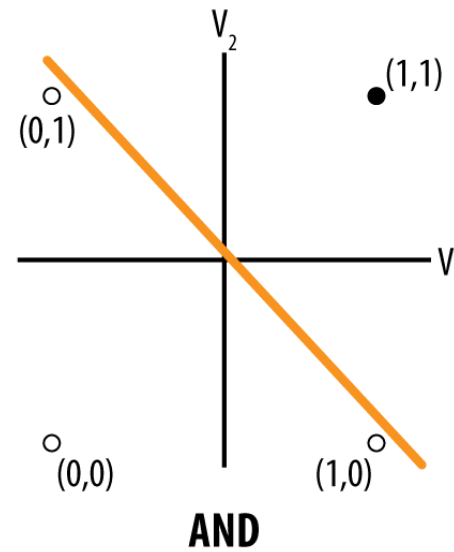
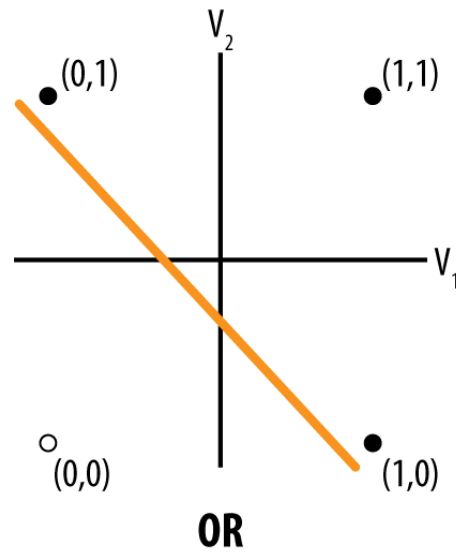
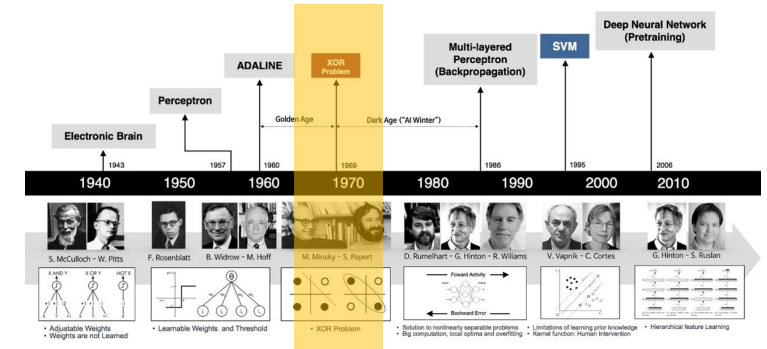
$$\sigma(x \times w_{xh}) \times w_{hy} = y$$



# Redes Neurais: História

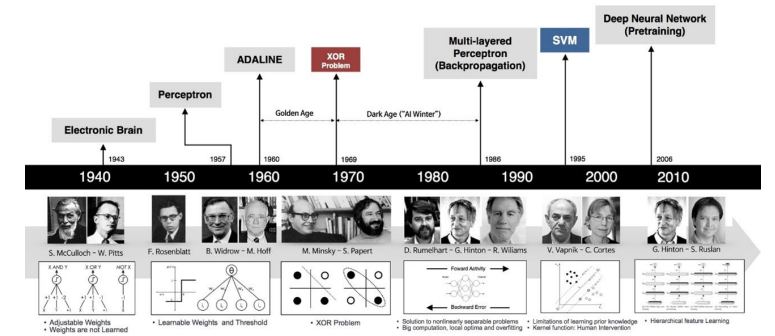
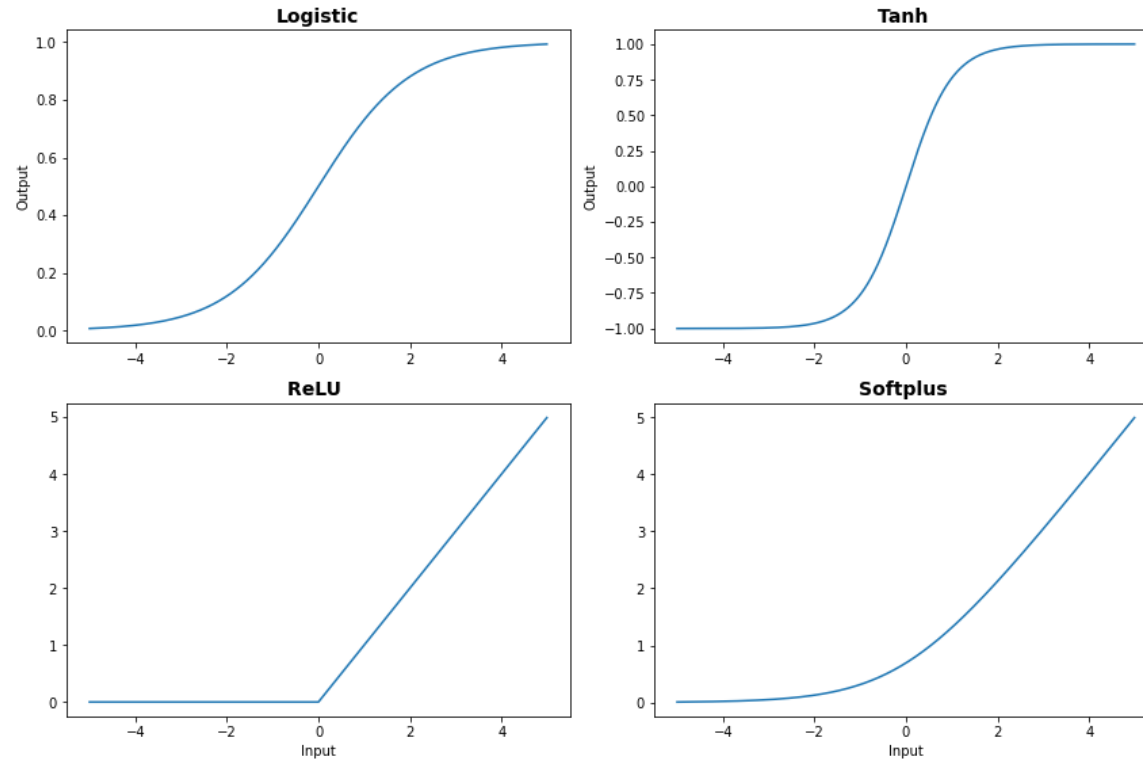


# Redes Neurais: História

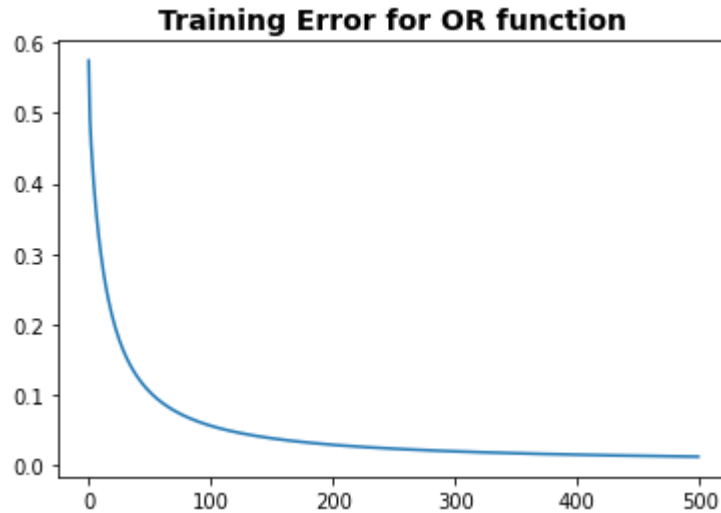


# Funções de Ativação

Funções de Ativação



# Aprendizado da Rede



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



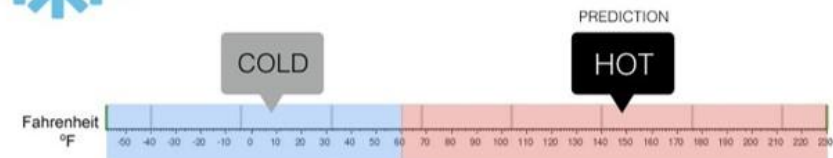
## Regression

What is the temperature going to be tomorrow?

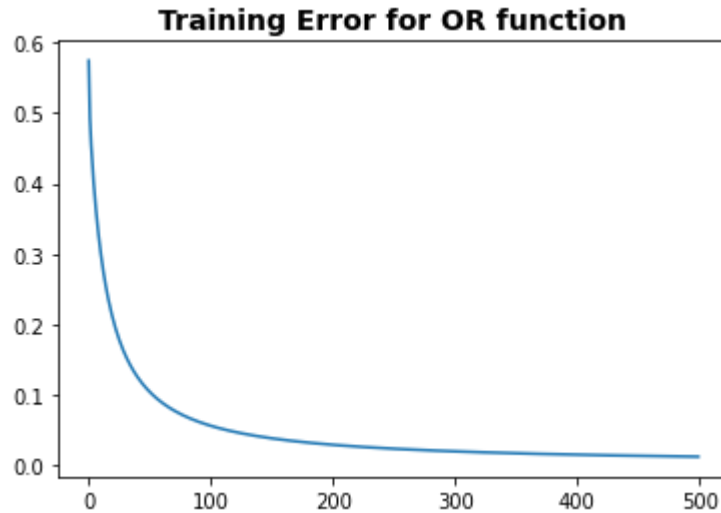


## Classification

Will it be Cold or Hot tomorrow?



# Aprendizado da Rede



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



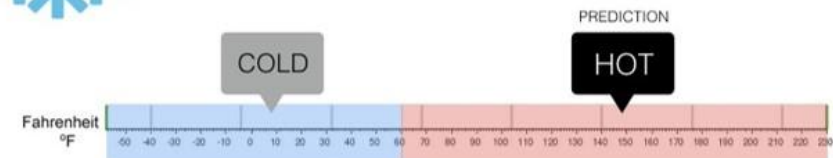
## Regression

What is the temperature going to be tomorrow?



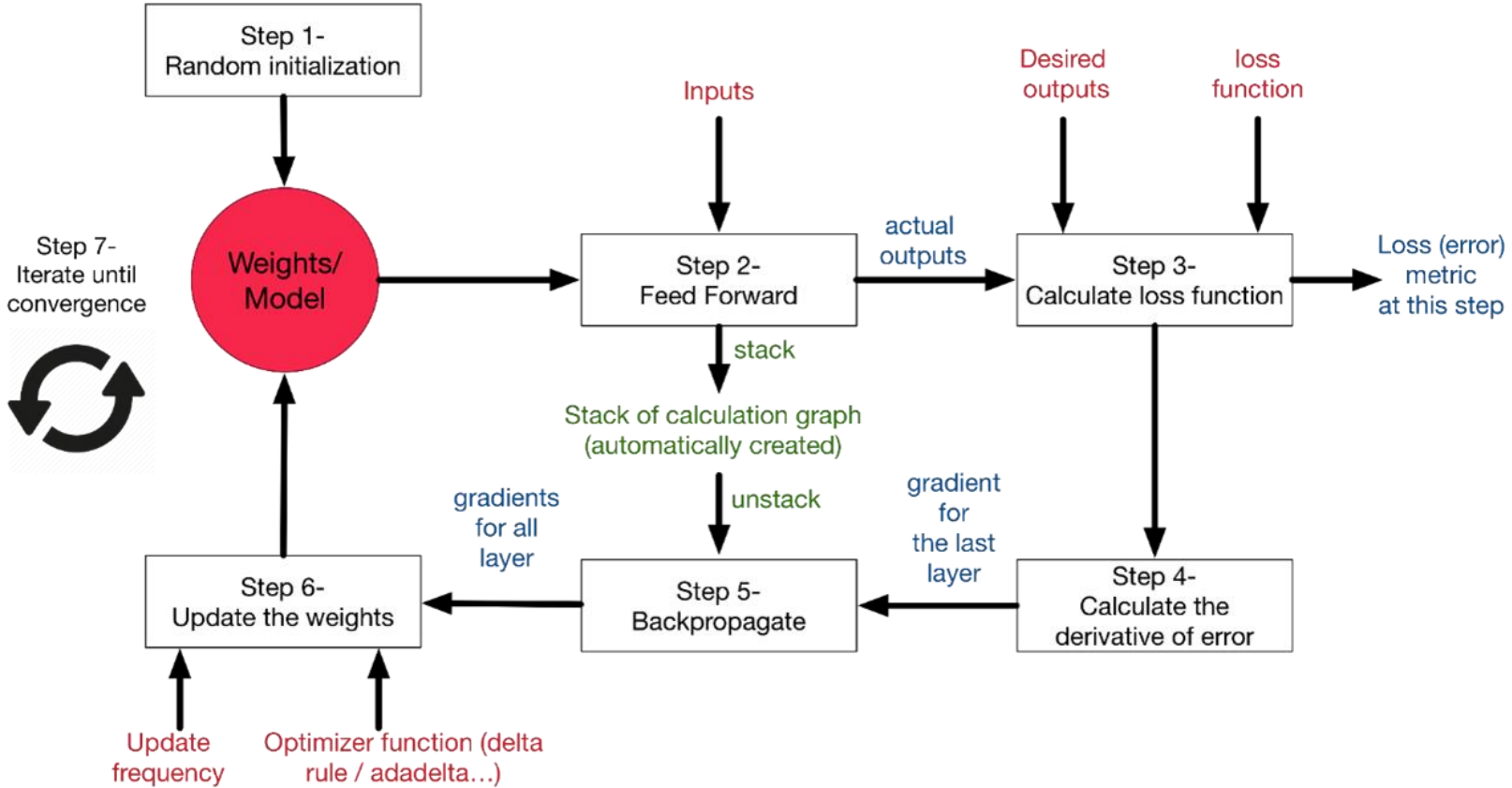
## Classification

Will it be Cold or Hot tomorrow?



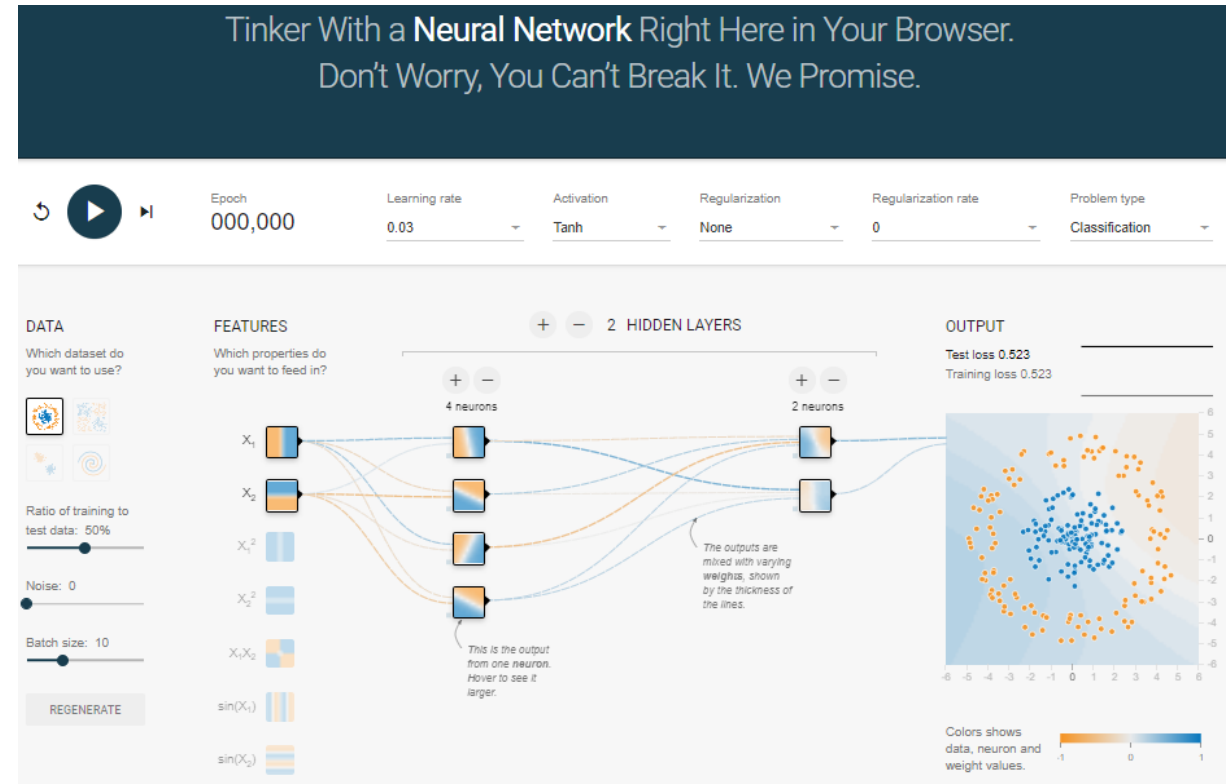


# Backpropagation



# Esperimente

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





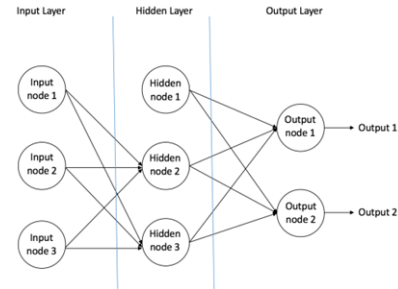
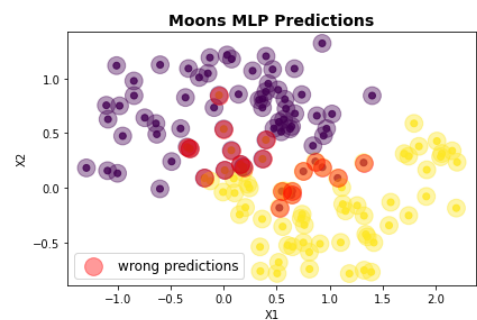
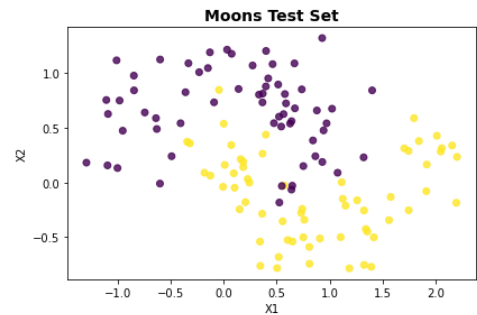
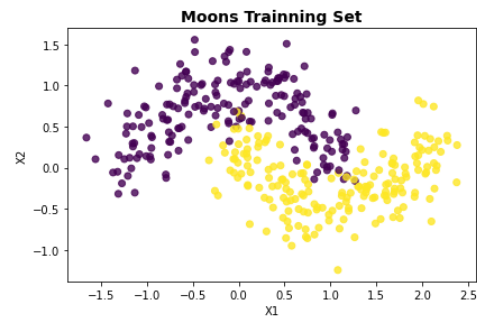
# **TRILHA 8**

## **Redes Neurais e Deep Learning**

**Parte B**

---

# MLP Com o scikit-learn



```
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*

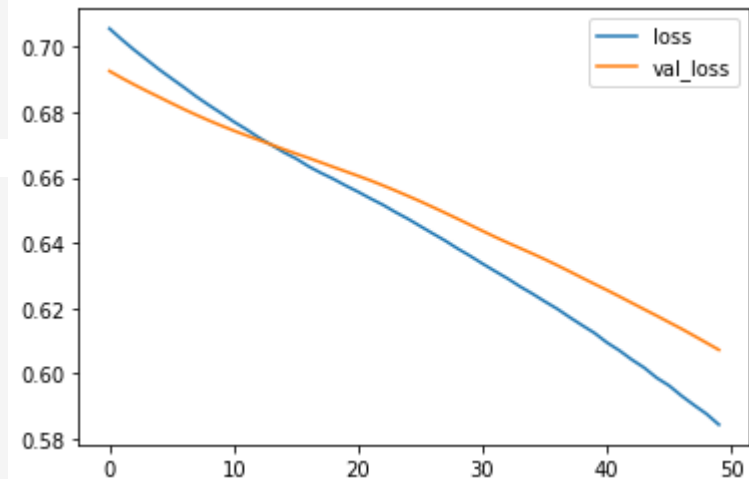
#-----

```
model.compile(loss='binary_crossentropy', # Multiclass loss
              optimizer='adam',
              metrics=['binary_accuracy'])
```

*# Treina o Modelo*

#-----

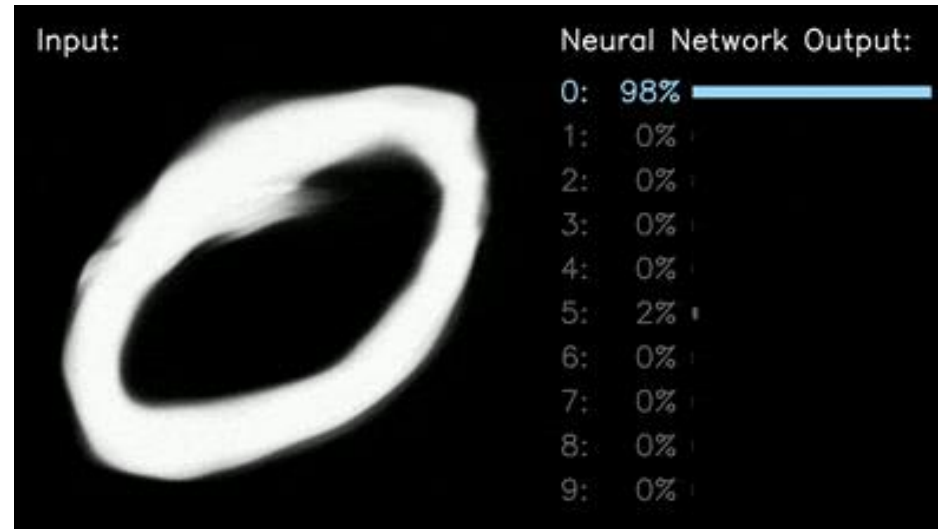
```
history = model.fit(
    X_train, y_train,
    batch_size=32,
    validation_split=0.3,
    epochs=50,
    verbose=1,
)
```



# Aplicações Redes Neurais



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





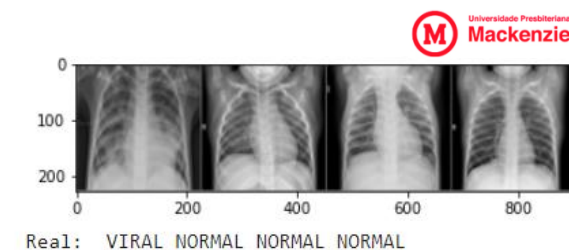
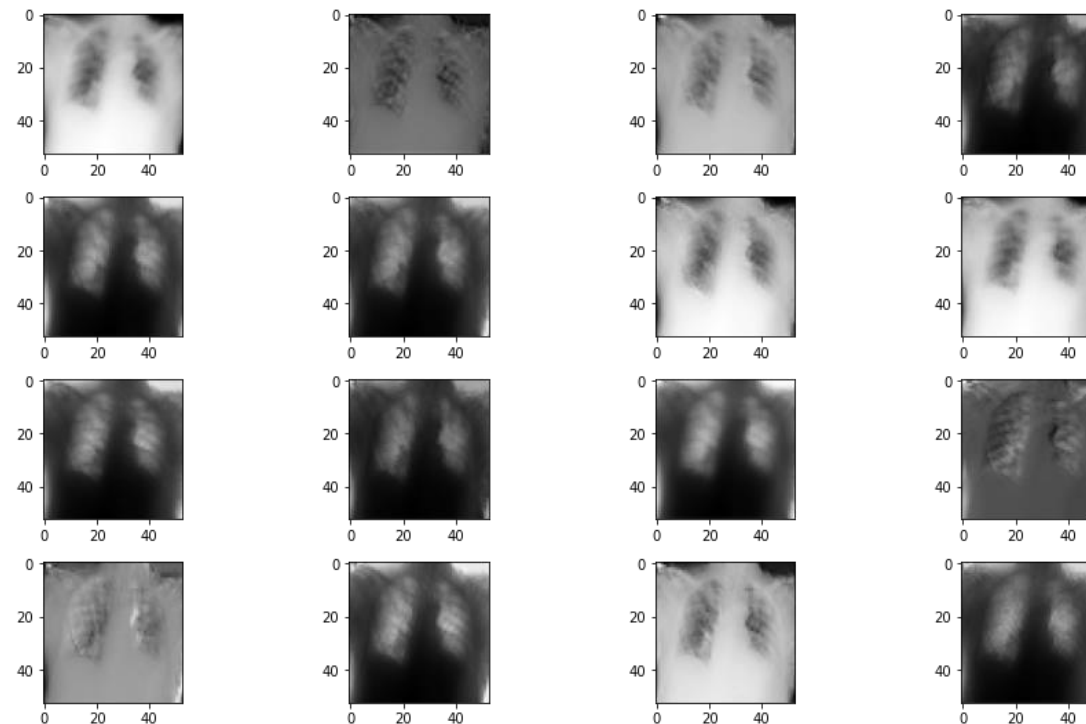
# Aplicações Deep Learning

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



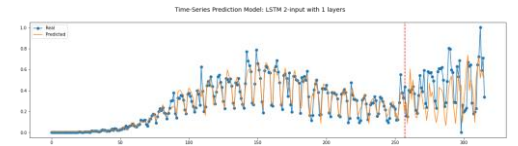
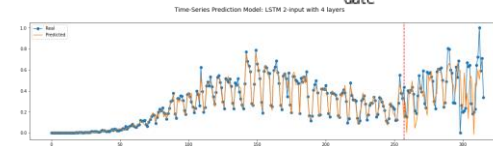
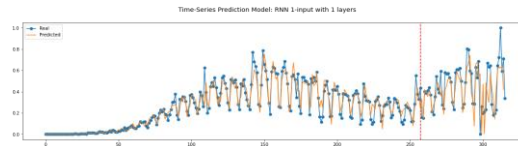
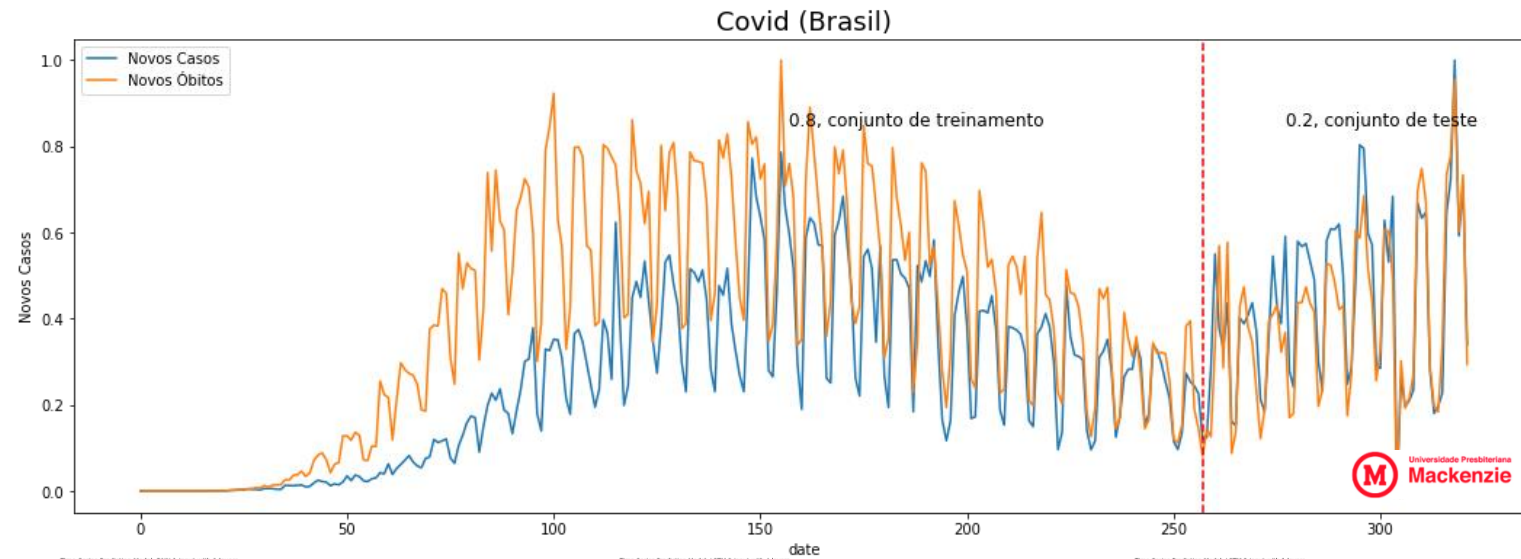
# Aplicações Deep Learning

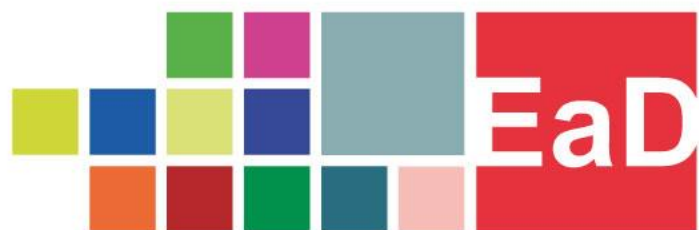
$$\sigma(\textcolor{red}{x} \times \textcolor{purple}{w}_{xh}) \times \textcolor{purple}{w}_{hy} = \textcolor{purple}{y}$$



# Aplicações Deep Learning

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





Universidade Presbiteriana  
**Mackenzie**