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✓ The Perceptron

✓ Implement the Perceptron using scikit-learn library

```

1  from sklearn.linear_model import Perceptron
2  import numpy as np
3
4  # Training data for AND gate
5  X = np.array([[0, 0], [0, 1], [1, 0], [1, 1]])
6  y = np.array([0, 0, 0, 1])
7
8  # Initialize and train Perceptron
9  model = Perceptron(max_iter=100, eta0=0.1, random_state=42)
10 model.fit(X, y)
11
12 # Results
13 print("Weights:", model.coef_)
14 print("Bias:", model.intercept_)
15 print("Predictions:", model.predict(X))

```

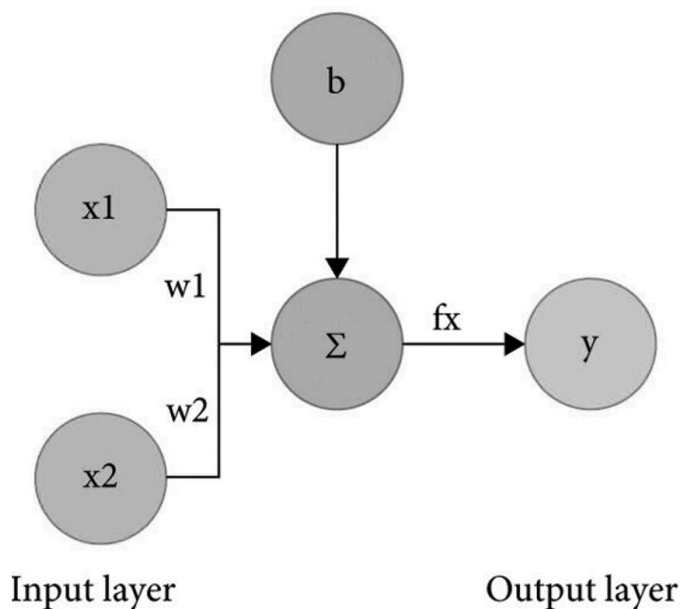
```

Weights: [[0.2 0.2]]
Bias: [-0.2]
Predictions: [0 0 0 1]

```

Note: In the scikit-learn Perceptron, the step function (also called the activation function) is a hard threshold function, and it's built-in.

```Python prediction = 1 if output >= 0 else 0



## ✓ The Mult-Layer Perceptron - MLP

### ✓ Solving the XOR Problem with a Neural Network

This code demonstrates how to build and train a simple neural network from scratch using NumPy to learn the XOR logic gate.

```

1 from sklearn.neural_network import MLPClassifier
2 import numpy as np
3
4 # XOR input and output
5 X = np.array([[0, 0], [0, 1], [1, 0], [1, 1]])
6 y = np.array([0, 1, 1, 0])
7
8 # Define MLP with 1 hidden layer of 2 neurons (minimal config for XOR)
9 mlp = MLPClassifier(hidden_layer_sizes=(2,), activation='tanh',
10 solver='adam', learning_rate_init=0.01,
11 max_iter=10000, random_state=42)
12
13
14 # Train the model
15 mlp.fit(X, y)
16
17 # Make predictions
18 predictions = mlp.predict(X)
19
20 print("Predictions:\n", predictions)
21 print("\nWeights (input to hidden):\n", "[w11 , w12]\n[w21 , w22]\n", mlp.coefs_[0])
22 print("\nBias hidden:\n", mlp.intercepts_[0])
23 print("\nWeights (hidden to output):\n", mlp.coefs_[1])
24 print("\nBias output:\n", mlp.intercepts_[1])
25

```

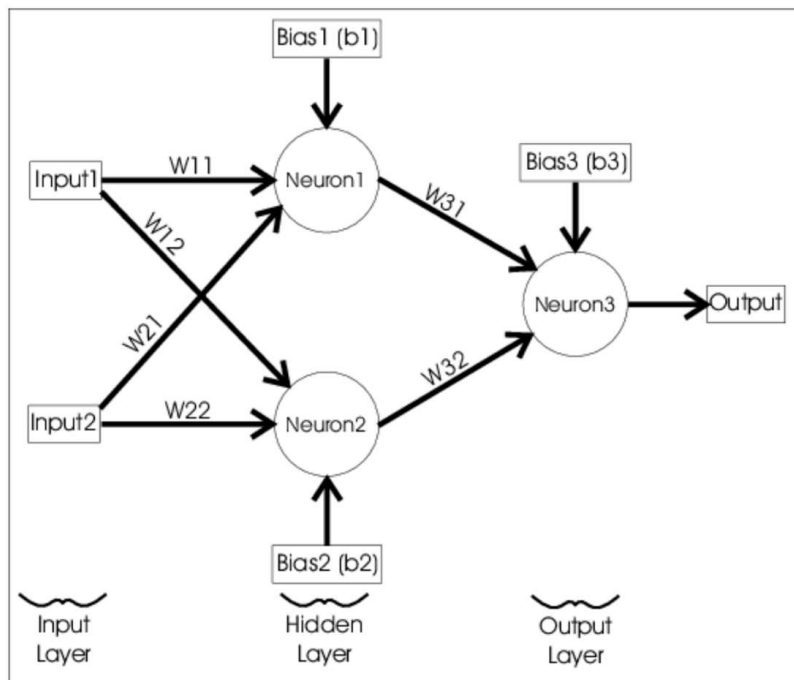
➞ Predictions:  
[0 1 1 0]

Weights (input to hidden):  
[ w11 , w12 ]  
[ w21 , w22 ]  
[[ 2.7144501 3.27401218]  
[-2.73418453 -3.17014048]]

Bias hidden:  
[ 1.21994174 -1.63451199]

Weights (hidden to output):  
[[-4.37775211]  
[ 4.46553876]]

Bias output:  
[3.61855675]



The Neural Network Model to solve the XOR Logic (from: <https://stopsmokingaids.me/>)

## ✓ Building Word Embeddings from Scratch

We can build word embeddings from scratch using a corpus of our own and using gensim library to build Word2Vec representations.

### ✓ Example 1 - Simple Tokenized Corpus

```
1 %pip install gensim
```

```
Collecting gensim
 Downloading gensim-4.3.3-cp310-cp310-win_amd64.whl (24.0 MB)
Collecting smart-open>=1.8.1
 Using cached smart_open-7.1.0-py3-none-any.whl (61 kB)
Collecting numpy<2.0,>=1.18.5
 Downloading numpy-1.26.4-cp310-cp310-win_amd64.whl (15.8 MB)
Collecting scipy<1.14.0,>=1.7.0
 Downloading scipy-1.13.1-cp310-cp310-win_amd64.whl (46.2 MB)
Collecting wrapt
 Using cached wrapt-1.17.2-cp310-cp310-win_amd64.whl (38 kB)
Installing collected packages: wrapt, numpy, smart-open, scipy, gensim
Attempting uninstall: numpy
 Found existing installation: numpy 2.2.5
 Uninstalling numpy-2.2.5:
 Successfully uninstalled numpy-2.2.5
Attempting uninstall: scipy
 Found existing installation: scipy 1.15.2
 Uninstalling scipy-1.15.2:
 Successfully uninstalled scipy-1.15.2
Successfully installed gensim-4.3.3 numpy-1.26.4 scipy-1.13.1 smart-open-7.1.0 wrapt-1.17.2
Note: you may need to restart the kernel to use updated packages.
WARNING: The script f2py.exe is installed in 'c:\Users\me\AppData\Local\Programs\Python\Python310\Scripts' which is not on PATH.
Consider adding this directory to PATH or, if you prefer to suppress this warning, use --no-warn-script-location.
WARNING: You are using pip version 21.2.3; however, version 25.0.1 is available.
You should consider upgrading via the 'c:\Users\me\AppData\Local\Programs\Python\Python310\python.exe -m pip install --upgrade pip' command
```

```
1 from gensim.models import Word2Vec
2
3 # Sample corpus
4 sentences = [
5 ['data', 'science', 'is', 'fun'],
6 ['machine', 'learning', 'is', 'powerful'],
7 ['data', 'and', 'learning', 'are', 'related']
8]
9
10 # Train the model
11 model = Word2Vec(sentences, vector_size=50, window=2, min_count=1, workers=2)
12
13 # Access the embedding for a word
14 print("Vector for 'data':\n", model.wv['data'])
15
16 # Find similar words
17 print("Words similar to 'data':", model.wv.most_similar('data'))
```

```
Vector for 'data':
[-0.01723938 0.00733148 0.01037977 0.01148388 0.01493384 -0.01233535
 0.00221123 0.01209456 -0.0056801 -0.01234705 -0.00082045 -0.0167379
-0.01120002 0.01420908 0.00670508 0.01445134 0.01360049 0.01506148
-0.00757831 -0.00112361 0.00469675 -0.00903806 0.01677746 -0.01971633
 0.01352928 0.00582883 -0.00986566 0.00879638 -0.00347915 0.01342277
 0.0199297 -0.00872489 -0.00119868 -0.01139127 0.00770164 0.00557325
 0.01378215 0.01220219 0.01907699 0.01854683 0.01579614 -0.01397901
-0.01831173 -0.00071151 -0.00619968 0.01578863 0.01187715 -0.00309133
 0.00302193 0.00358008]
Words similar to 'data': [('are', 0.16563551127910614), ('fun', 0.13940520584583282), ('learning', 0.1267007291316986), ('powerful', 0.0
```

### ✓ Example 2 - Simple Untokenized Corpus

The code below uses the gensim library to build word embeddings using Word2Vec models from scratch.

It uses a text corpus to learn word similarities.

```

1 %pip install nltk
2
3 import nltk
4 nltk.download('punkt_tab')

```

Requirement already satisfied: nltk in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (3.9.1)  
Requirement already satisfied: click in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from nltk) (8.1.8)  
Requirement already satisfied: regex>=2021.8.3 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from nltk) (2024.11.8)  
Requirement already satisfied: tqdm in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from nltk) (4.67.1)  
Requirement already satisfied: joblib in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from nltk) (1.4.2)  
Requirement already satisfied: colorama in c:\users\me\appdata\roaming\python\python310\site-packages (from click->nltk) (0.4.6)  
Note: you may need to restart the kernel to use updated packages.  
WARNING: You are using pip version 21.2.3; however, version 25.0.1 is available.  
You should consider upgrading via the 'c:\Users\me\AppData\Local\Programs\Python\Python310\python.exe -m pip install --upgrade pip' command.  
[nltk\_data] Downloading package punkt to  
[nltk\_data] C:/Users/me/AppData/Roaming/nltk\_data...  
[nltk\_data] Package punkt is already up-to-date!

```

1 import gensim
2 from gensim.models import Word2Vec
3 from nltk.tokenize import word_tokenize
4
5 # Sample corpus
6 sentences = [
7 "Large language models are transforming business applications",
8 "Natural language processing helps computers understand human language",
9 "Word embeddings capture semantic relationships between words",
10 "Neural networks learn distributed representations of words",
11 "Businesses use language models for various applications",
12 "Customer service can be improved with language technology",
13 "Modern language models require significant computing resources",
14 "Language models can generate human-like text for businesses"
15]
16
17 # Tokenize the sentences
18 tokenized_sentences = [word_tokenize(sentence.lower()) for sentence in sentences]
19
20 # Train Word2Vec model
21 model = Word2Vec(
22 sentences=tokenized_sentences,
23 vector_size=100, # Embedding dimension
24 window=5, # Context window size
25 min_count=1, # Minimum word frequency
26 workers=4 # Number of threads
27)
28
29 # Save the model
30 model.save("word2vec.model")
31
32 # Find the most similar words to "language"
33 similar_words = model.wv.most_similar("language", topn=5)
34 print("Words most similar to 'language':")
35 for word, similarity in similar_words:
36 print(f"{word}: {similarity:.4f}")
37
38 # Vector for a specific word
39 word_vector = model.wv["business"]
40 print(f"\nVector for 'business' (first 10 dimensions):\n{word_vector[:10]}")
41
42 # Word analogies
43 analogy_result = model.wv.most_similar(
44 positive=["business", "language"],
45 negative=["models"],
46 topn=3
47)
48 print("\nAnalogy results:")
49 for word, similarity in analogy_result:
50 print(f"{word}: {similarity:.4f}")

```

Words most similar to 'language':  
natural: 0.2196  
between: 0.2167  
resources: 0.1955  
distributed: 0.1696  
significant: 0.1522

Vector for 'business' (first 10 dimensions):

```
[0.00816812 -0.00444303 0.00898543 0.00825366 -0.00443522 0.00030311
 0.00427449 -0.00392632 -0.00555997 -0.00651232]
```

```
Analogy results:
neural: 0.2595
natural: 0.2004
resources: 0.1899
```

## Word Similarities Examples

### Use Fake Embeddings

```
1 import numpy as np
2 from sklearn.metrics.pairwise import cosine_similarity
3
4 # Fake word vectors (3D for simplicity)
5 word_vectors = {
6 "king": np.array([0.8, 0.65, 0.1]),
7 "queen": np.array([0.78, 0.66, 0.12]),
8 "man": np.array([0.9, 0.1, 0.1]),
9 "woman": np.array([0.88, 0.12, 0.12]),
10 "apple": np.array([0.1, 0.8, 0.9]),
11 }
12 def similarity(w1, w2):
13 return cosine_similarity([word_vectors[w1]], [word_vectors[w2]])[0][0]
14
15 print("Similarity(king, queen):", similarity("king", "queen"))
16 print("Similarity(man, woman):", similarity("man", "woman"))
17 print("Similarity(king, apple):", similarity("king", "apple"))
```

```
→ Similarity(king, queen): 0.9995995265529728
 Similarity(man, woman): 0.999399810286
 Similarity(king, apple): 0.5514092058274782
```

### Use Real Embeddings - Gensim library

```
1 import gensim.downloader as api
2 from gensim.models import Word2Vec
3
4 # Load pre-trained Word2Vec model
5 word2vec_model = api.load("word2vec-google-news-300")
6
7 # Find similar words
8 similar_words = word2vec_model.most_similar('computer', topn=5)
9 print("Words similar to 'computer':", similar_words)
10
11 # Word analogies
12 result = word2vec_model.most_similar(positive=['woman', 'king'], negative=['man'], topn=1)
13 print("king - man + woman =", result)
14
15 # Train your own Word2Vec model
16 sentences = [["cat", "say", "meow"], ["dog", "say", "woof"]]
17 model = Word2Vec(sentences, vector_size=100, window=5, min_count=1, workers=4)
18
19 # Get vector for a word
20 cat_vector = model.wv['cat']
21 print("Vector for 'cat':", cat_vector[:5]) # Show first 5 dimensions
```

```
→ Words similar to 'computer': [('computers', 0.7979382276535034), ('laptop', 0.6640493869781494), ('laptop_computer', 0.6548869013786316)
 king - man + woman = [('queen', 0.7118191123008728)]
 Vector for 'cat': [-0.00713902 0.00124103 -0.00717672 -0.00224462 0.0037193]
```

### Use Real Embeddings - Spacy library

Download spacy and the required libraries

```
1 %pip install spacy
2 !python -m spacy download en_core_web_md
```

Requirement already satisfied: spacy in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (3.8.5)

Requirement already satisfied: typer<1.0.0,>=0.3.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: pydantic!=1.8,!<1.8.1,<3.0.0,>=1.7.4 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: srsly<3.0.0,>=2.4.3 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: packaging>=20.0 in c:\users\me\appdata\roaming\python\python310\lib\site-packages (from spacy) (24.2)

Requirement already satisfied: spacy-legacy<3.1.0,>=3.0.11 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: langcodes<4.0.0,>=3.2.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: requests<3.1.0,>=2.13.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: spacy-loggers<2.0.0,>=1.0.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: thinc<8.4.0,>=8.3.4 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: weasel<0.5.0,>=0.1.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: requests<3.0.0,>=2.13.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: cymem<2.1.0,>=2.0.2 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: Jinja2 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy) (3.1.6)

Requirement already satisfied: tqdm<5.0.0,>=4.38.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: setuptools in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy) (57.4.0)

Requirement already satisfied: wasabi<1.2.0,>=0.9.1 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: numpy>=1.19.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy) (2.2.2)

Requirement already satisfied: catalogue<2.1.0,>=2.0.6 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from spacy)

Requirement already satisfied: language-data>=1.2 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from language-data)

Requirement already satisfied: marisa-trie>=1.1.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from language-data)

Requirement already satisfied: pydantic-core==2.33.1 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from pydantic)

Requirement already satisfied: typing-extensions>=4.12.2 in c:\users\me\appdata\roaming\python\python310\lib\site-packages (from pydantic)

Requirement already satisfied: typing-inspection>=0.4.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from pydantic)

Requirement already satisfied: annotated-types>=0.6.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from pydantic)

Requirement already satisfied: charset-normalizer<4,>=2 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from requests)

Requirement already satisfied: urllib3<3,>=1.21.1 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from requests)

Requirement already satisfied: idna<4,>=2.5 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from requests)

Requirement already satisfied: certifi>=2017.4.17 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from requests)

Requirement already satisfied: blis<1.4.0,>=1.3.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from thinc)

Requirement already satisfied: confection<1.0.0,>=0.0.1 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from thinc)

Requirement already satisfied: colorama in c:\users\me\appdata\roaming\python\python310\lib\site-packages (from tqdm)

Requirement already satisfied: click>=8.0.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from typer)

Requirement already satisfied: shellingham>=1.3.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from typer)

Requirement already satisfied: rich>=10.11.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from typer)

Requirement already satisfied: pygments<3.0.0,>=2.13.0 in c:\users\me\appdata\roaming\python\python310\lib\site-packages (from rich)

Requirement already satisfied: markdown-it-py>=2.2.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from rich)

Requirement already satisfied: mdurl<=0.1 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from markdown-it-py)

Requirement already satisfied: smart-open<8.0.0,>=5.2.1 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from smart-open)

Requirement already satisfied: cloudpathlib<1.0.0,>=0.7.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from smart-open)

Requirement already satisfied: MarkupSafe>=2.0 in c:\users\me\appdata\local\programs\python\python310\lib\site-packages (from Jinja2)

Note: you may need to restart the kernel to use updated packages.

WARNING: You are using pip version 21.2.3; however, version 25.0.1 is available.

You should consider upgrading via the 'c:\Users\me\AppData\Local\Programs\Python\Python310\python.exe -m pip install --upgrade pip' command.

Collecting en-core-web-md==3.8.0

Downloading [https://github.com/explosion/spacy-models/releases/download/en\\_core\\_web\\_md-3.8.0/en\\_core\\_web\\_md-3.8.0-py3-none-any.whl](https://github.com/explosion/spacy-models/releases/download/en_core_web_md-3.8.0/en_core_web_md-3.8.0-py3-none-any.whl) (3.8MB)

Installing collected packages: en-core-web-md

Successfully installed en-core-web-md-3.8.0

✓ Download and installation successful

You can now load the package via spacy.load('en\_core\_web\_md')

WARNING: You are using pip version 21.2.3; however, version 25.0.1 is available.

You should consider upgrading via the 'c:\Users\me\AppData\Local\Programs\Python\Python310\python.exe -m pip install --upgrade pip' command.

```
1 import spacy
2 nlp = spacy.load("en_core_web_md")
3
4 word1 = nlp("king")
5 word2 = nlp("queen")
6 print("Similarity:", word1.similarity(word2))
```

Similarity: 0.38253095611315674

