

Goal: Use the perceptron algorithm to iterate until convergence

Tools: Numpy

Data:

- Positive class (+1): (1,3)
- Negative class (-1): (-1,4)

Task: Determine the number of updates until convergence using the perceptron algorithm. You must iterate over data points in the order: [(1,3),(-1,4)]. Your output should be the sequence of updates in the form $\mathbf{w}_i = [w_1, \dots, w_n]$

Perceptron Algorithm:

1. **Input:** Training data set: $(\mathbf{x}_1, y_1), \dots, (\mathbf{x}_N, y_N)$

2. **Initialization:**

- Weight vector: $\mathbf{w}^{(s)} = [0, \dots, 0]$
- Iteration count: $s = 0$

3. **Update Loop:**

- While there exists an instance $i \in [N]$ such that the prediction is incorrect, i.e., $y\mathbf{w}^{(s)} \cdot \mathbf{x} \leq 0$:
- Update weight:
 - For positive instances: $\mathbf{w}^{(s+1)} = \mathbf{w}^{(s)} + \mathbf{x}$
 - For negative instances: $\mathbf{w}^{(s+1)} = \mathbf{w}^{(s)} - \mathbf{x}$
- Increment counter: $s = s + 1$

4. **Output:** Final weight vector: $\mathbf{w}^{(s)}$

```
In [6]: # Import necessary libraries
import numpy as np
import matplotlib.pyplot as plt

# Define training data (vector form x = [x1, x2])
# Positive class (+1): (1, 3)
# Negative class (-1): (-1, 4)
# Order: [(1,3), (-1,4)]
X = np.array([[1, 3],          # (1, 3), y = +1
              [-1, 4]])      # (-1, 4), y = -1
y = np.array([1, -1]) # classes
```

```

# Initialize weight vector:  $w(s) = [0, \dots, 0]$ 
w = np.zeros(2)

# Initialize iteration counter:  $s = 0$ 
update_sequence = []
s = 0

# Update Loop: While there exists instance  $i$  such that  $y \cdot w \cdot x \leq 0$ 
converged = False
while not converged:
    misclassified = False
    for i in range(len(X)): # Iterate data in order
        # Prediction incorrect if  $y \cdot w(s) \cdot x \leq 0$ 
        if y[i] * np.dot(w, X[i]) <= 0:
            # For positive instances:  $w^{(s+1)} = w^{(s)} + x$ 
            # For negative instances:  $w^{(s+1)} = w^{(s)} - x$ 
            if y[i] == 1:
                w = w + X[i]
            else:
                w = w - X[i]
            s = s + 1
            update_sequence.append(w.copy())
            misclassified = True
            break # Restart from beginning of the dataset after each update
    if not misclassified:
        converged = True

# Output: Final weight vector  $w(s)$ 
print(f"Number of updates until convergence: {s}")
print("\nSequence of weight updates:")
for i, w_i in enumerate(update_sequence):
    print(f"\tw_{i+1} = [{', '.join(f'{x:.1f}' for x in w_i)}]")
print(f"\nFinal weight vector: w = [{', '.join(f'{x:.1f}' for x in w)}]")

# Plot points and final decision boundary
plt.figure()

# Plot data points
plt.scatter(1, 3, c='blue', s=100, marker='o', label='Positive (+1)', zorder=5)
plt.scatter(-1, 4, c='red', s=100, marker='s', label='Negative (-1)', zorder=5)

# Plot decision boundary:  $w \cdot x = 0 \rightarrow w_1 \cdot x_1 + w_2 \cdot x_2 = 0 \Rightarrow x_2 = -w_1 \cdot x_1 / w_2$ 
x1_line = np.linspace(-5, 5, 100)
if w[1] != 0:
    x2_line = (-w[0] * x1_line) / w[1]
    plt.plot(x1_line, x2_line, 'k-', linewidth=2, label='Decision boundary')
else:
    # Vertical line when  $w_2=0$ :  $x_1 = 0$ 
    plt.axvline(x=0, color='k', linewidth=2, label='Decision boundary')

plt.xlabel('$x_1$')
plt.ylabel('$x_2$')
plt.xlim(-5, 5)
plt.ylim(-5, 5)
plt.legend()

```

```
plt.grid(True, alpha=0.3)
plt.title('Perceptron: Data Points and Decision Boundary')
plt.tight_layout()
plt.show()
```

Number of updates until convergence: 5

Sequence of weight updates:

```
w_1 = [1.0, 3.0]
w_2 = [2.0, -1.0]
w_3 = [3.0, 2.0]
w_4 = [4.0, -2.0]
w_5 = [5.0, 1.0]
```

Final weight vector: $w = [5.0, 1.0]$

