

perceptron

August 29, 2024

1 The Perceptron

Created Aug. 2024 for the FSU Course: *Machine Learning in Physics* H. B. Prosper

The perceptron, defined by

$$y = g(x A^T + b), \quad (1)$$

is the basic computational element of neural networks. The output y and the bias b are row matrices, while the input x is a matrix consisting of one or more rows.

1.1 Tips

- Use **esc r** to disable a cell
- Use **esc y** to reactivate it
- Use **esc m** to go to markdown mode. **Markdown** is the typesetting language used in jupyter notebooks.
- In a markdown cell, double tap the mouse or glide pad (on your laptop) to go to edit mode.
- Shift + return to execute a cell (including markdown cells).
- If the equations don't typeset, try double tapping the cell again, and re-execute it.

1.2 Import modules

```
[1]: import numpy as np
import torch
```

1.3 Input Data

The input data can consist of one or more rows. We make the data type (**dtype**) `float32` to be compatible with PyTorch.

```
[2]: x = np.array([-2.0, 1.0, 4.0], dtype=np.float32)
```

1.4 Perceptron Parameters

```
[3]: # Perceptron 1
A1 = np.array([2, -3, 1], dtype=np.float32)
b1 = -5

# Perceptron 2
```

```
A2 = np.array([1, 2, 3], dtype=np.float32)
b2 = -4
```

1.4.1 Perceptron 1: $z_1 = x A_1^T + b_1$; $y_1 = g(z_1)$

```
[4]: z1 = x @ A1.T + b1
     y1 = np.maximum(0, z1)
     z1, y1
```

```
[4]: (-8.0, 0.0)
```

1.4.2 Perceptron 2: $z_2 = x A_2^T + b_2$; $y_2 = g(z_2)$

```
[5]: z2 = x @ A2.T + b2
     y2 = np.maximum(0, z2)
     z2, y2
```

```
[5]: (8.0, 8.0)
```

1.5 Multi-node Perceptron

Construct a 2-node perceptron with 3 inputs.

```
[6]: A = np.vstack([A1, A2]) # vertical stack
     b = np.hstack([b1, b2]) # horizontal stack
     A.shape, A, b.shape, b
```

```
[6]: ((2, 3),
      array([[ 2., -3.,  1.],
             [ 1.,  2.,  3.]], dtype=float32),
      (2,),
      array([-5, -4]))
```

Compute output of perceptron

```
[7]: z = x @ A.T + b
     y = np.maximum(0, z)
     z, y
```

```
[7]: (array([-8.,  8.]), array([0., 8.]))
```

1.6 Do the same using PyTorch

```
[8]: # instantiate a linear function with 3 inputs and 2 outputs
linear = torch.nn.Linear(3, 2)

# instantiate a relu function that will apply itself
# elementwise to its matrix argument.
relu = torch.nn.ReLU()

# compute output of a (3, 2) perceptron
with torch.no_grad(): # disable gradient calculation
    linear.weight.copy_(torch.tensor(A))
    linear.bias.copy_(torch.tensor(b))
    z = linear(torch.tensor(x)) # compute linear function
    y = relu(z)
z, y
```

```
[8]: (tensor([-8.,  8.]), tensor([0.,  8.]))
```

```
[ ]:
```