Aprendizado Profundo

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Pytorch: Como funciona

- Tensors
- Datasets
- DataLoader
- Processo de Treinamento
- Processo de avaliação



Dataset

```
class CustomDataset(Dataset):
  def __init__(self, ...):
  def __len__(self):
      \cdots \ \cdot
  def __getitem__(self, idx):
```

Dataloader

```
DataLoader(

training_data or test_data,

batch_size,

...
```

Dataset

número de exemplos: 1000000

divisão: 70 treino, 20 validação, 10 teste

- Arquitetura

```
(0,1,0) (1,1,0) Esfera de raio 0.1 (1,0,0) (1,0,0)
```

```
RosemblattPerceptron(
   (perceptron): Linear(in_features=3, out_features=8, bias=True)
)
```

- Acurácia: 100%

```
Test Error:
Accuracy: 100.0%, Avg loss: 1.321413
```

Questão 2 a) XOR

- Dataset

número de exemplos: 4

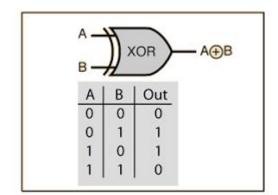
divisão: 4 treino, 4 teste

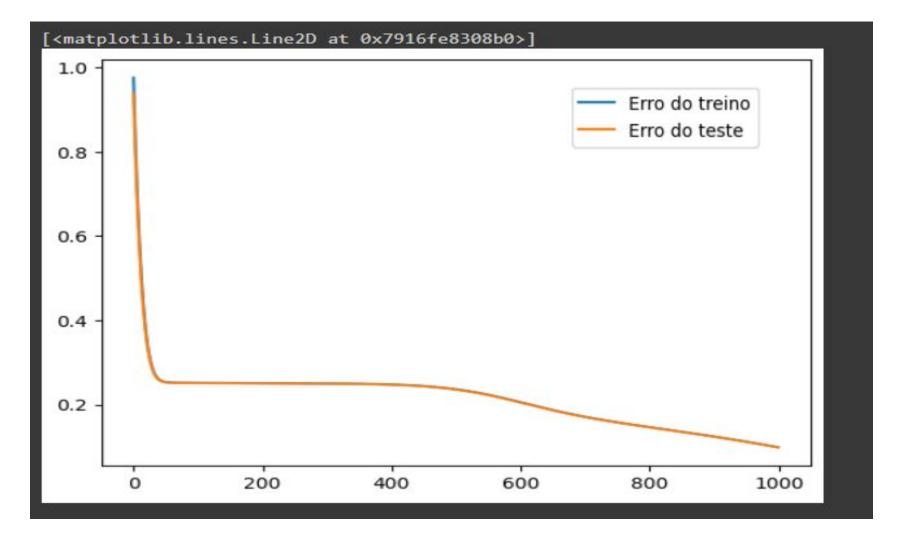
Arquitetura: Adam, Ir=0.005, MSE

```
MultiLayerPerceptron(
    (linear): Linear(in_features=2, out_features=2, bias=True)
    (Sigmoid): Sigmoid()
    (linear2): Linear(in_features=2, out_features=1, bias=True)
)
```

- Acurácia:

```
Epoch 1000
------
loss: 0.001433 [ 1/ 4]
Test Error:
Accuracy: 75.0%, Avg loss: 0.098248
```





b)
$$f(x) = log_{10}(x)$$
, onde $1 \le x \le 10$

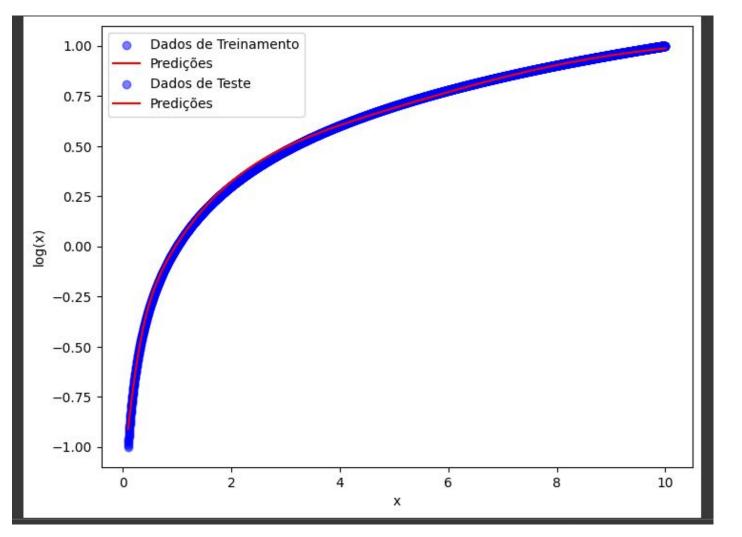
- Dataset:
 - 5000 exemplos, divisão 85/15
 - batch_size = 64
- Arquitetura: Adam, Ir=0.01, MSE

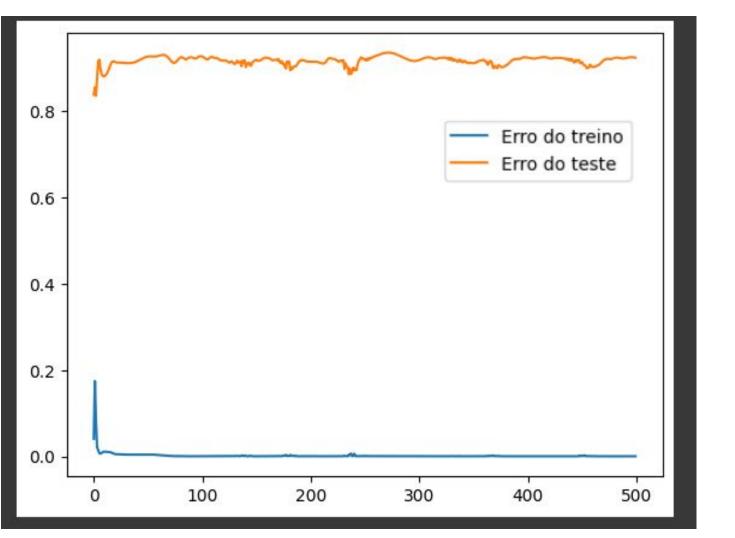
```
0 1 2 3 4 5 6 7 8 9 10
```

```
LogMultiLayerPerceptron(
   (linear): Linear(in_features=1, out_features=128, bias=True)
   (Sigmoid): Sigmoid()
   (linear2): Linear(in_features=128, out_features=1, bias=True)
)
```

- Loss

```
Test Error:
Avg loss: 0.923831
```





C)
$$f(x) = 10x^5 + 5x^4 + 2x^3 - 0.5x^2 + 3x + 2$$

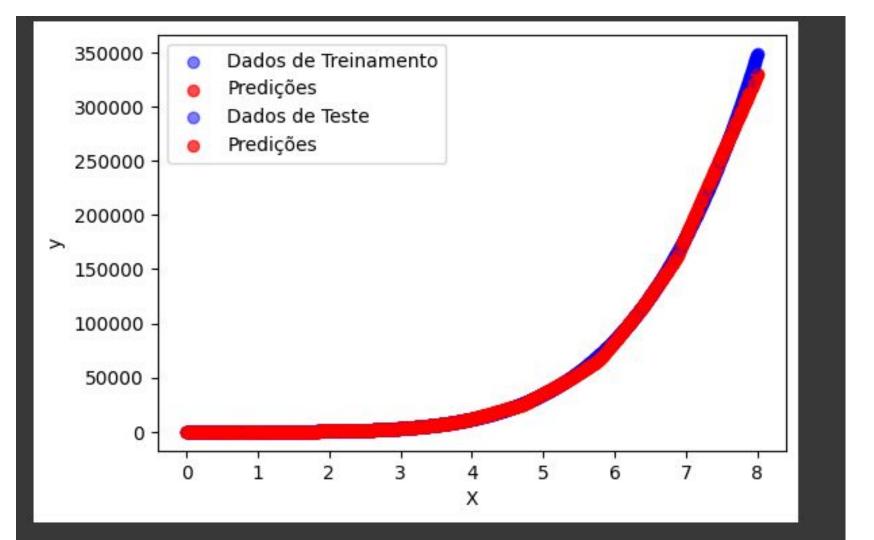
onde $0 \le x \le 8$

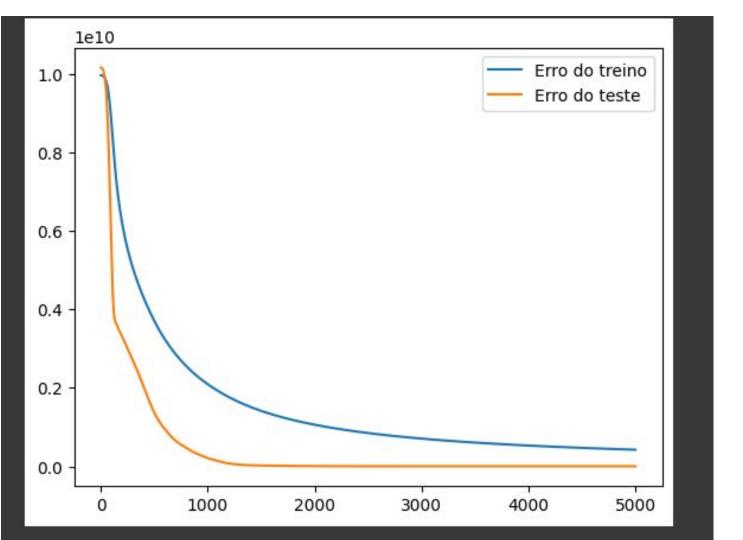
- Dataset:
 - 2000 exemplos, divisão aleatória 80/20
 - batch_size = 32
- Arquitetura: Adam, Ir=0.01, MSE

```
PoynomialModel(
    (linear): Linear(in_features=1, out_features=128, bias=True)
    (ReLU): ReLU()
    (linear2): Linear(in_features=128, out_features=64, bias=True)
    (ReLU2): ReLU()
    (linear3): Linear(in_features=64, out_features=1, bias=True)
)
```

Loss

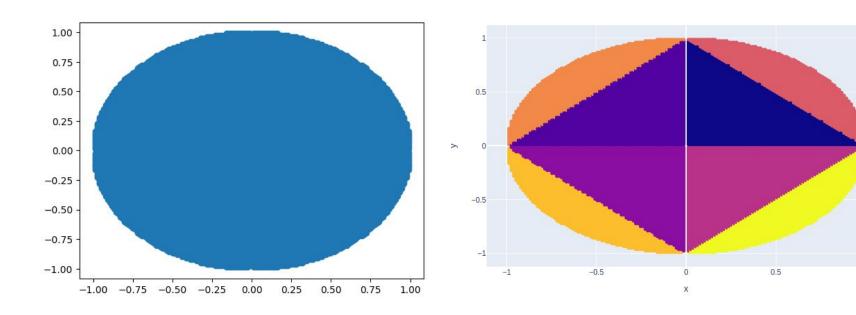
```
Epoch 5000
------
Test Error:
Avg loss: 24364401.406250
```





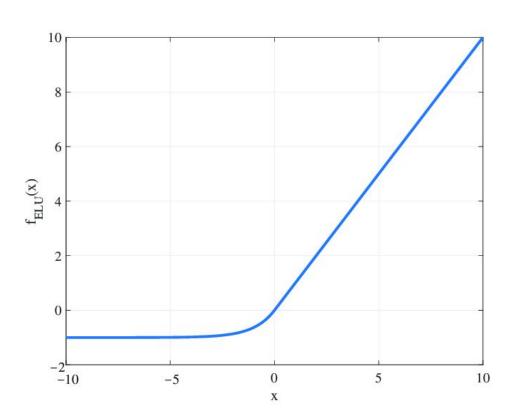
Dataset

- número de exemplos: 17246, divisão aleatória 80/20



- Arquitetura

```
ClassesNeuralNet(
   (elu): ELU(alpha=1.0)
   (linear1): Linear(in_features=2, out_features=256, bias=True)
   (linear2): Linear(in_features=256, out_features=256, bias=True)
   (linear3): Linear(in_features=256, out_features=64, bias=True)
   (linear4): Linear(in_features=64, out_features=16, bias=True)
   (linear5): Linear(in_features=16, out_features=8, bias=True)
)
```



a) Sem momentum

loss: Cross Entropy / Optimizer: SGD / Learning rate: 0.005

Batch size: 128 / N_Epochs: 1000

Acurácia:

```
loss: 1.434185 [ 128/13796]
loss: 1.413840 [12928/13796]
Test Error:
Accuracy: 83.7%, Avg loss: 1.440140
```

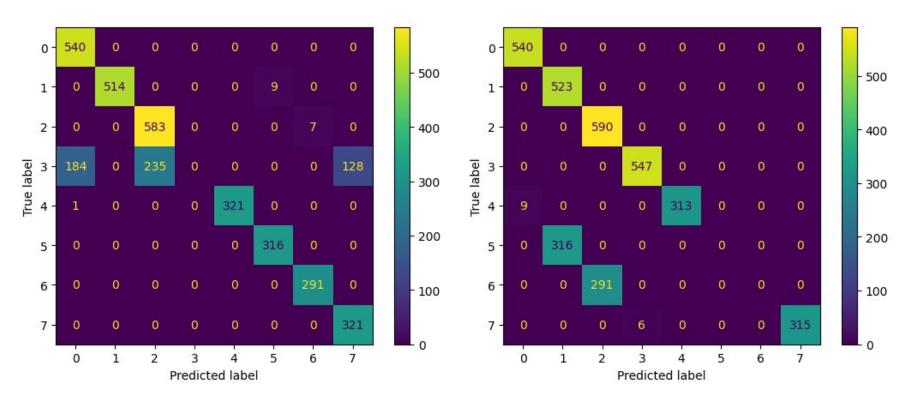
a) Com momentum (momentum = 0.5)

loss: Cross Entropy / Optimizer: SGD / Learning rate: 0.005

Batch size: 128 / N_Epochs: 1000

Acurácia:

```
loss: 1.446448 [ 128/13796]
loss: 1.425820 [12928/13796]
Test Error:
Accuracy: 82.0%, Avg loss: 1.454916
```



Sem momentum

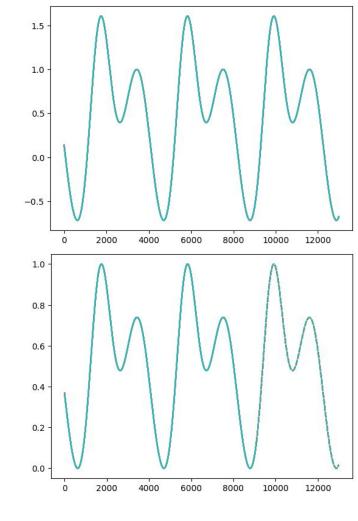
Com momentum

- Dataset
 - número de exemplos: 12988, divisão 70/30

$$x(n) = sen^2(n) + cos(n+cos(n)).$$

$$k = 10$$

$$x_{scaled} = rac{x - x_{min}}{x_{max} - x_{min}}$$



Arquitetura

loss: MSELoss / Optimizer: Adam / Learning rate: 0.001

Batch size: 64 / N_Epochs: 1000

```
TemporalSeriesNeuralNet(
   (linear_layers): Sequential(
      (0): Linear(in_features=10, out_features=32, bias=True)
      (1): Tanh()
      (2): Linear(in_features=32, out_features=64, bias=True)
      (3): Tanh()
      (4): Linear(in_features=64, out_features=64, bias=True)
      (5): Tanh()
      (6): Linear(in_features=64, out_features=32, bias=True)
      (7): Tanh()
      (8): Linear(in_features=32, out_features=3, bias=True)
    )
)
```

