# M9-L1-P1

November 7, 2023

## 1 M9-L1 Problem 1

Here, you will implement three loss functions from scratch in numpy: MAE, MSE, and MAPE.

```
[]: import numpy as np
    y_gt1 = np.array([1,2,3,4,5,6,7,8,9,10])
    y_pred1 = np.array([1,1.3,3.1,4.6,5.9,5.9,6.4,9.2,8.1,10.5])
    y_gt2 = np.array([-3.23594765, -3.74125693, -2.3040903, 0.
      -30190142, -1.68434859, 1.10160357, 0.8587438, 1.76546802, 3.13787123, 3.
      472990216, 5.89871795, 6.06406803, 6.28329118, 7.46406525, 8.21246221, 10.
      423145281, 9.39080133, 10.76761316, 10.45903557, 9.61872736, 13.68392163, 14.
      475332509, 14.00530973, 17.87581523, 15.01028079, 17.36899084, 17.99463433, 11.
      420.57318325, 21.36834867, 20.91252318, 21.99432414, 21.58696173, 21.
      435253687, 23.84400704, 25.20685402, 27.13938159, 27.97005662, 27.23893581, 11
      428.18254573, 28.29488138, 28.78200226, 29.35433587, 33.86996731, 32.2681256
      4, 33.19828933, 33.24215413, 36.13102571, 34.59822336, 36.85796679, 37.
      403382637, 39.17478129, 39.13565951, 39.32441832, 41.33545414, 42.65055409, U
      43.1473253 , 44.24186584, 44.1636577 , 45.29382449, 45.84269107, 47.
      40.1418421, 47.41917695, 47.36462649, 50.12692109, 50.40629987, 50.03646832, L
      452.98803478, 52.47654002, 54.29436964, 55.83010066, 56.08857887, 57.9575825
      4, 56.44194186, 58.93769518, 58.7091293 , 59.3817281 , 60.53226145, 61.
      465814444, 62.88444817, 62.52171885, 65.44628103, 65.86970284, 64.72638258, L
      468.60946432, 69.87568716, 70.01716341, 69.51704486, 69.48480293, 72.
      46859314, 71.86955033, 74.3537582 , 74.19817397, 75.82512388, 76.0634371 , ⊔
      477.27222973, 77.43474244, 80.06869878, 79.26832623, 80.40198936])
```

```
y_pred2 = np.array([-3.17886560e+00, -3.72628642e+00, -2.28154027e+00, -2.
 42424242e-06, 2.96261368e-01, -1.70080838e+00, 1.09113641e+00,
 →60043722e-01, 1.76729042e+00, 3.12498677e+00, 3.72452933e+00,
 481293300e+00, 6.01791742e+00, 6.27564586e+00, 7.43093457e+00, 8.
 418505900e+00, 1.00785853e+01, 9.41006754e+00, 1.07339029e+01, 1.
 →05483666e+01, 9.86429504e+00, 1.35944803e+01, 1.46257911e+01, 1.
 41092530e+01, 1.74700758e+01, 1.52285866e+01, 1.73610430e+01, 1.
 $0283176e+01, 2.02578402e+01, 2.10543695e+01, 2.08801196e+01, 2.
 419111495e+01, 2.17786086e+01, 2.17754891e+01, 2.39269636e+01, 2.
 $1674432e+01, 2.68054871e+01, 2.76337491e+01, 2.73444399e+01, 2.
 -82677426e+01, 2.85915692e+01, 2.91907133e+01, 2.98552019e+01, 3.
 -32092384e+01, 3.24325813e+01, 3.33437229e+01, 3.36586115e+01, 3.
 $\square$58501097e+01, 3.51566050e+01, 3.69363787e+01, 3.73654528e+01, 3.
 $\text{90232127e+01}$, 3.93355670e+01$, 3.97886962e+01$, 4.13471034e+01$, 4.
 $\text{\text{\cute{24678677e+01}}}, 4.31186248e+01$, 4.41080463e+01$, 4.44437982e+01$, 4.
 $4.54581242e+01, 4.61509657e+01, 4.71832256e+01, 4.78047650e+01, 4.
 -81822755e+01, 5.00379827e+01, 5.06088232e+01, 5.08521636e+01, 5.
 $\text{\pi}27428151e+01$, 5.29526597e+01$, 5.42661662e+01$, 5.54230479e+01$, 5.
 460162341e+01, 5.72972123e+01, 5.71389028e+01, 5.87005639e+01, 5.
 $\text{91111760e+01}$, 5.98988234e+01$, 6.08826528e+01$, 6.18502423e+01$, 6.
 428491288e+01, 6.32501917e+01, 6.48567227e+01, 6.55629719e+01, 6.
 $\sqrt{57207391e+01}$, 6.75883810e+01, 6.85509197e+01, 6.91918142e+01, 6.
 →96421235e+01, 7.02288144e+01, 7.17044458e+01, 7.21593122e+01, 7.
 -34448231e+01, 7.40436375e+01, 7.50845851e+01, 7.57923722e+01, 7.
 →67262442e+01, 7.74266118e+01, 7.86387737e+01, 7.91677250e+01,
 →00787815e+01])
```

#### 1.1 Mean Absolute Error

Complete the definition for MAE(y\_gt, y\_pred) below.

$$\text{MAE} = \frac{1}{n} \sum_{i=1}^{n} |y_i - \hat{y}_i| = \frac{1}{n} \sum_{i=1}^{n} |e_i|$$

MAE(y\_gt1, y\_pred1) should return 0.560.

```
[]: def MAE(y_gt, y_pred):
    return np.sum(abs(y_gt - y_pred)) / len(y_gt)

print(f"MAE(y_gt1, y_pred1) = {MAE(y_gt1, y_pred1):.3f}")
print(f"MAE(y_gt2, y_pred2) = {MAE(y_gt2, y_pred2):.3f}")

MAE(y_gt1, y_pred1) = 0.560
MAE(y_gt2, y_pred2) = 0.290
```

### 1.2 Mean Squared Error

Complete the definition for MSE(y\_gt, y\_pred) below.

$$\text{MSE} = \frac{1}{n} \sum_{i=1}^{n} (y_i - \hat{y}_i)^2 = \frac{1}{n} \sum_{i=1}^{n} (e_i)^2 = \frac{1}{n} e^T e$$

MSE(y\_gt1, y\_pred1) should return 0.454.

```
[]: def MSE(y_gt, y_pred):
    return np.sum(np.power((y_gt - y_pred),2)) / len(y_gt)

print(f"MSE(y_gt1, y_pred1) = {MSE(y_gt1, y_pred1):.3f}")

print(f"MSE(y_gt2, y_pred2) = {MSE(y_gt2, y_pred2):.3f}")

MSE(y_gt1, y_pred1) = 0.454

MSE(y_gt2, y_pred2) = 0.174
```

## 1.3 Mean Absolute Percentage Error

Complete the definition for MAPE(y\_gt, y\_pred, epsilon) below.

$$MAE = \frac{1}{n} \sum_{i=1}^{n} \frac{|y_i - \hat{y}_i|}{|y_i| + \varepsilon} = \frac{1}{n} \sum_{i=1}^{n} \frac{|e_i|}{|y_i| + \varepsilon}$$

MAPE(y\_gt1, y\_pred1, 1e-6) should return 0.112.

 $MAPE(y_gt2, y_pred2) = 0.032$ 

```
[]: def MAPE(y_gt, y_pred, epsilon=1e-6):
    return np.sum(abs(y_gt - y_pred)/(abs(y_gt) + epsilon)) / len(y_gt)

print(f"MAPE(y_gt1, y_pred1) = {MAPE(y_gt1, y_pred1):.3f}")

print(f"MAPE(y_gt2, y_pred2) = {MAPE(y_gt2, y_pred2):.3f}")

MAPE(y_gt1, y_pred1) = 0.112
```