## Neural Network from Scratch

## June 13, 2025

```
[83]: import numpy as np
      class NeuralNetwork():
          #Set random seed
          np.random.seed(42)
          def __init__(self, X, y, n_hidden_neurons, output_act_fn="linear", u
       →error_fn="mse"):
              self.y=y
              self.X=X
              self.n_input_neurons = X.shape[1]
              self.n_hidden_neurons = n_hidden_neurons
              self.output_act_fn = output_act_fn
              self.error_fn = error_fn
          #initialize weights and biases with random values (array of random numbers)
              self.input_hidden_weights = np.random.randn(self.n_input_neurons,self.
       →n_hidden_neurons)
              self.hidden_biases = np.random.randn(self.n_hidden_neurons)
              self.hidden_output_weights = np.random.randn(self.n_hidden_neurons,1)
              self.output_bais = np.random.randn(1)
          # activation function
          def activition(self, x, act_fn):
              if act_fn == 'sigmoid' :
                  return 1/(1 + np.exp(-x))
              elif act fn == 'relu' :
                  return np.maximum(0,x)
              elif act fn == 'linear':
                  return x
              else:
                  raise Exception('Unknown activation function')
          def activition_derivitive(self, x, act_fn):
```

```
if act_fn == 'sigmoid' :
          return x * (1-x)
      elif act_fn == 'relu' :
          return np.where(x>0,1,0)
      elif act_fn == 'linear':
          return 1
      else:
          raise Exception('Unknown activation function')
      # Forward Propagation
  def forward pass(self, X):
      #input layer
      self.input = X
           #Hidden layer
      self.hidden = self.activition(np.dot(self.input, self.
⇔input_hidden_weights) + self.hidden_biases , "relu")
          #Output layer
      self.output = self.activition(np.dot(self.hidden, self.
hidden_output_weights) + self.output_bais , self.output_act_fn)
      return self.output
  def error(self, y_true, y_pred):
      if self.error_fn == 'mse':
          return np.mean(np.square(y_true - y_pred))
      elif self.error fn == 'cross entropy':
          return -np.mean(y_true * np.log(y_pred+0.00001)+(1 - y_true)* np.
\hookrightarrowlog(1 - y_pred+0.00001))
      else:
          raise Exception('Unknown error function')
  def error_derivitive(self, y_true, y_pred):
      if self.error_fn == 'mse':
          return 2* (y_pred - y_true) / y_true.size
      elif self.error_fn == 'cross_entropy':
          return (y_pred - y_true) / (y_pred * (1- y_pred+0.00001)* y_true.
⇔size)
      else:
          raise Exception('Unknown activation function')
      # Backpropagation
  def backward_pass(self, X, y_true, y_pred, learning_rate):
      # output layer
      self.output_error = self.error_derivitive(y_true, y_pred) * self.
activition_derivitive(y_pred , self.output_act_fn)
      self.output_bais -= learning_rate * np.sum(self.output_error , axis=0)
```

```
→output_error)
                  # Hidden layer
              self.hidden_error = np.dot(self.output_error, self.
       →hidden output weights.T) * self.activition derivitive(self.hidden, "relu")
              self.hidden_biases -= learning_rate * np.sum(self.hidden_error,axis=0)
              self.input_hidden_weights -= learning_rate * np.dot(X.T,self.
       ⇔hidden_error)
                  # return weights and biases
              return self.input_hidden_weights, self.hidden_biases, self.
       →hidden_output_weights, self.output_bais
              # Training
          def train(self, X, y, learning_rate, epochs):
              for epoch in range(epochs):
                  y_pred = self.forward_pass(X)
                  wih, bh, who , bo =self.backward_pass(X, y, y_pred, learning_rate)
                  if epoch % 500 == 0:
                      print('Epoch: {}, Loss: {:.3f}'.format(epoch, self.error(y, __
       →y pred)))
              print('Trainig complete!')
              return wih, bh, who, bo
              # Predition
          def predict(self, X):
              if self.error fn == "mse":
                  return self.forward_pass(X)
              elif self.error_fn == 'cross_entropy':
                  return np.where(self.forward_pass(X) > 0.5, 1, 0)
[66]: from sklearn.datasets import make regression
      X, y = make_regression(n_samples=1000, n_features=3, noise=20, __
       →random_state=42)
      y = y.reshape(-1,1)
      from sklearn.model_selection import train_test_split
      X_train, X_test, y_train, y_test = train_test_split(X,y,random_state=42)
      from sklearn.preprocessing import StandardScaler
      scaler = StandardScaler()
      X_train = scaler.fit_transform(X_train)
      X_test = scaler.transform(X_test)
```

self.hidden\_output\_weights -= learning\_rate \* np.dot(self.hidden.T,self.

```
[67]: nn = NeuralNetwork(X_train, y_train, n_hidden_neurons=128,__
       →output_act_fn='linear',error_fn='mse')
      wih, bh, who, bo= nn.train(X_train, y_train, learning_rate=0.001, epochs=5000)
     Epoch: 0, Loss: 13892.784
     Epoch: 500, Loss: 374.455
     Epoch: 1000, Loss: 367.597
     Epoch: 1500, Loss: 362.922
     Epoch: 2000, Loss: 359.279
     Epoch: 2500, Loss: 356.310
     Epoch: 3000, Loss: 354.433
     Epoch: 3500, Loss: 352.962
     Epoch: 4000, Loss: 351.569
     Epoch: 4500, Loss: 350.323
     Trainig complete!
[68]: # Weights and biases
      print('Input-hidden weights:\n', wih)
      print('Hidden-output weights:\n', who)
      print('Hidden biases:\n', bh)
      print('Output bias:\n', bo)
     Input-hidden weights:
       \begin{bmatrix} 0.46218215 & 0.32542526 & 0.16580122 & 1.95488085 & 0.38762215 & -0.20067982 \end{bmatrix} 
        2.37748297 1.29989384 -2.04491936 0.99280742 -1.00687122 -0.45065981
        0.24196227 \ -2.53614614 \ -2.86502429 \ -0.57946922 \ -0.97141953 \ \ 0.43746702
       -1.1375737 -2.10772852 1.68217445 -1.30138492 0.0675282 -1.46967514
       -0.85697598 0.04103586 -0.18329947 0.27823312 0.29107773 -1.26135854
       -0.29820138 2.74711965 -0.94704823 -1.47916521 1.36182059 -1.38830393
       -0.14503344 -2.32309602 -1.27638632 0.53413518 1.37159347 0.25528276
        0.06449208 - 0.4329507 - 2.22006145 - 0.54760222 - 0.35777331 0.84362734
        0.56385418 -0.65745027 0.81709272 -0.34187398 -0.38302591 0.90215247
                    0.82033207 - 2.03080346 - 0.66764091 0.45827357 2.0147918
        1.3512
       -0.42215082 -0.18260957 -0.91808413 -1.86473578 1.12831257 1.94606514
       -1.14753947 0.31739748 0.58125847 -1.36779867 1.01385854 3.32589176
        0.17883453 2.03555612 -3.01400746 0.62750092 0.13721074 -0.990968
        0.81978882 -2.19759049 -0.09314316 1.0601566
                                                        2.66800386 -0.15633042
       -0.87627891 -0.86861731 0.78476595 0.29750958 -0.26421085 -0.72215849
        0.27824087 2.35579203 -0.54494303 0.59275396 -1.33860321 -2.44785674
       -0.53712489 1.68819686 0.00511346 0.43626686 -1.46287535 -0.443088
       -0.04317059 -0.79546518 -0.02259556 1.44727207 2.90546878 0.97093419
       -0.42545192 -0.14622313 -1.76017047 0.83785549 0.11687791 2.78604817
       -0.29754586 0.03559718 -0.03297419 -1.11915353 1.11644524 2.31861803
        1.27818612 -1.21913875 1.15585473 -1.90729882 1.08559308 2.23150204
       -2.71706934 -1.39488468]
      [-0.00442579 -0.94091384 -2.27458144 0.54909627 -1.03743309 0.31757436]
       -1.06957905 1.99265692 -0.89502155 0.42080245 0.30863459 -1.24499542
```

```
0.22745993  0.29548771  -2.65461159  -0.04847006  0.32516244  0.37562201
 -1.01047592 -2.67174119 0.79936286 -0.04671264 0.25049285 0.15920662
 -1.48683963 0.09516851 0.81307274 -0.75585487 2.6197359
                                                          0.03797221
 -1.35885333 1.88079317 -1.4391623
                                    0.11762717 1.7808744 -1.19762485
  1.44405059 0.59427192 0.75905219 2.03933201 0.20573006 -0.72073368
 -0.70176757 -0.9378861 -0.06616551 0.45844735 0.5542515
                                                          0.75592334
  0.26788389 2.25313165 0.16985023 2.7788275
                                               1.35987084 -0.85993899
 -1.49953486 0.29078571 -0.98026318 1.11075019 0.61490588 0.85108741
 -0.71318754 -2.61000742 -0.36714651 0.28794195 0.53611805 -1.32912275
 -0.69410131 1.56653742 -0.78166424 -0.05178262 0.27903486 -1.30846656
             0.4631465
  0.76569207 -0.0861997
                         0.61062185 4.00623855 1.12663503 1.35869671
  0.94341078 \quad 0.49281619 \quad -0.36869928 \quad 0.63838179 \quad -0.42062862 \quad -0.61557853
 -0.31910665 0.91337097 2.74870441 -1.56153159 -0.49854119 -2.20883677
                         0.06428002 -1.42756813 -0.77719381 0.55393923
 -1.7518049
             2.5817312
 -0.7049738
             0.29710192 0.16937041 -0.63759652 2.27346071 0.59148973
 -3.06137419 0.14888587 0.03693704 1.27611788 -0.9156966 -0.74853593
  0.36336082  0.84941854 -0.77134528  0.31998717 -1.15715922  0.2807826
  2.35788674 0.28047312 -1.23830583 0.55971073 2.55534557 1.7585209
 -2.31908636 -1.16643643]
 [ 1.44143401 -1.30273023 0.0982733
                                    0.61023393 -1.66404253 -0.03067419
 -3.47219136 -0.84555609 -0.42031453 -1.15138212 1.9156877 -1.51425349
 -0.44004449 0.20817502 1.63881898 -1.40745931 1.20484096 -0.35448066
 -0.9634703
             -0.24379913 0.72297561 2.18824832 -1.30149344 1.97500375 -1.61395584
 -0.10866721 0.44511048 -0.32118724 -1.07328422 0.37333917 -1.07255418
 -0.90382202 0.93334217 0.54279918 -0.54304684 0.98241633 0.3384591
             0.34414505 \ -1.37054077 \ -1.15988946 \ \ 0.62022934 \ \ 0.48661066
  0.764915
  0.02143086 0.01011749 1.11420243 -0.53548405 1.40004348 -0.39699959
 -0.6024473
             0.88528525  0.48706769  1.30353547  1.7515929
                                                          0.25028817
  0.83859257 -0.22403764 0.56114245 -0.68554737 0.04274842 1.21467233
 -1.19844121 3.55636012 -1.14787216 -1.14145097 1.70353061 1.23759415
  0.93491708 - 0.33031212 - 0.93372478 - 0.10244442  0.33170286 - 0.60153854
 -0.89814501 0.17555379 0.18705018 -0.42793725 -0.2771536
                                                          0.6545942
 -1.20497703 -1.71998109 -0.1647706
                                   0.00852352 0.60125401 0.97628025
  0.39726967 -0.85364041 -0.01901621 -1.29851232 -0.28023824 -0.29030066
  0.72460099 -0.99323958 0.52207924 1.87842071 0.87463242 -0.15788732
  0.49843531 - 0.45229578 \quad 0.32160484 \quad 0.42938905 \quad 0.51737127 - 1.11377118
 -0.04258467 0.68910059 2.07189998 1.6695096
                                               2.25143301 -0.2052847
  0.85781939 0.23552898 2.08110418 -0.88709053 -1.0037583 -1.84326642
 -2.3985645 -1.18303281]]
Hidden-output weights:
 [[ 0.77127738]
 [-1.68740343]
```

[-1.6514519]

[ 1.22359783]

[-1.38787795]

- [-0.81706864]
- [-2.07535953]
- [ 1.63401409]
- [-2.61559157]
- [ 0.62517808]
- [-1.00070499]
- [ 0.82169753]
- [-0.76325916]
- [-2.22028514]
- [-3.59788847]
- [ 0.26688467]
- [-0.2442513]
- [-1.13964814]
- [-0.48387187]
- [-3.2579528]
- [ 1.1950648 ]
- [-1.73886255]
- [-0.65183611]
- [ 0.14251482]
- [-1.68898119]
- [-0.54336465]
- [ 1.82168908]
- [ 0.08466454]
- [ 1.62217277]
- [-1.09681149]
- [-0.56478992]
- [ 2.92179914]
- [-3.33271513]
- [-1.46694335]
- [ 1.8627603 ]
- [-1.41900868]
- [ 1.93783147]
- [-2.38230332]
- [-0.05273362]
- [ 0.09860665]
- [ 1.59489029]
- [ 0.13923718]
- [ 0.32647709]
- [-0.51220271]
- [-1.87415922]
- [-0.75269223]
- [ 0.86339984]
- [-0.48237294]
- [ 0.5946124 ]
- [ 2.22733168]
- [ 0.92419535]
- [ 0.54468774]
- [ 1.83476189]

- [-0.93158284]
- [-2.8154273]
- [-0.42320537]
- [-2.73130807]
- [-1.49232127]
- [ 1.07311388]
- [ 2.57086843]
- [ 0.30180414]
- [-2.0550943]
- [ 0.69411754]
- [-2.76071315]
- [ 0.96240502]
- [ 2.05629809]
- [-2.2048558]
- [ 3.28019096]
- [ 1.13609466]
- [-1.14835529]
- [ 1.69011895]
- [ 3.81512183]
- [ 0.57057891]
- [ 1.33966561]
- [-1.42359787]
- [-0.93613482]
- [ 1.1320221 ]
- [-1.79371717]
- [ 1.10323956]
- [-0.96556529]
- [ 0.72202702]
- [ 1.34911882]
- [ 2.53479484]
- [-0.82588938]
- [-0.74821561]
- [-1.13061931]
- [-0.06684489]
- [ 0.34919607]
- [ 0.64528216]
- [-1.78478673]
- [ 0.87502155]
- [ 2.84092817]
- [ 1.25890978]
- [ 1.54281163]
- [-1.47945244]
- [-2.24153225]
- [-2.36117002]
- [ 3.57469381]
- [ 0.65436566]
- [-1.30886287]
- [-0.8685134]

```
[-1.02147502]
```

- [ 3.04993551]
- [-0.47395864]
- [-0.07916169]
- [ 1.89952411]
- [ 2.44328571]
- [ 0.68141195]
- [-2.4004152]
- [ 0.65237837]
- [ 0.00237037]
- [ 1.16312199]
- [ 2.10812542]
- [ 2.09311468]
- [ 1.96323762]
- [-0.82030826]
- [ 0.57936264]
- [-1.35913466]
- [ 2.03576393]
- 2.00070000
- [-2.03407222]
- [ 2.59587243]
- [ 2.00553461]
- [-0.79808752]
- [-1.05215708]
- [-2.12987003]
- [ 2.076819 ]
- [ 2.51687888]
- [-3.44033733]
- [-2.31646109]]

## Hidden biases:

```
[-0.83217041 0.48078105 0.60500092 1.83181869 0.84768687 -0.43361308
-0.31495565 0.50567809 -1.26771871 1.72089313 0.86846817 -0.59073022
-1.71313453 1.64559666 -0.66618967 1.31707125 -1.57484585 -1.05687093
-0.02871066 0.42936581 -0.82510897 0.48114156 -1.06762043 -0.18754183
0.29103455 -0.56827477 2.50065608 0.59289856 1.32468887 -0.00681422
2.53579178
          2.59179591 -0.3579881
                                  0.65795527  0.58878343  1.35347752
-1.14431306 0.77684639 0.83432256 -1.6018695 -1.42044452 -2.2075279
-0.27662089
           1.0778175
                       1.55049274 -0.19984998 1.42375252 -1.43011273
-2.12283033 0.09133101 0.73402503 -0.04115782 -1.95155762 0.43518948
-1.30749202 0.53624658 0.47927843 -1.36619846 -0.55337524 -1.2208661
-0.50360656 1.06612
                      -1.3447997
                                  0.44376862 -0.48235377 -0.49798926
-0.04761146 -1.03443846 -0.68489279 -1.65346779 2.10230124 0.03115642
-0.89415368 0.37019393 -0.23489184 0.09947136 0.57638062 0.81055735
-0.73941678 -0.60578679 -0.32259746 -2.35380474 -1.7089166
                                                        1.83537318
1.8712451
            0.03133935  0.55625421 -0.03510943  3.07908381  0.5831429
-0.24818138 -1.61537568 -1.60644632 0.35414385 -0.9858581 -1.39102994
-1.88373002 -1.02342941 1.68566408 0.91367684 0.16209647 1.41598164
-0.30243442 -0.94781571 1.19579611 1.18321205 -1.56529744 -0.91770512
-0.72903446 -1.45742261 1.17371663 2.23042967 -1.34104751 1.00281478
```

```
-1.04570909 -0.32285165 -1.27323543 -1.16332243 -0.85025504 -1.38503437
      -0.12145419 -0.08248742]
     Output bias:
      [-0.4295871]
[72]: # Accuracy
     from sklearn.metrics import r2_score
      y_pred = nn.predict(X_test)
      print('R2 Score: {:.3f}'.format(r2_score(y_test , y_pred)))
     R2 Score: 0.975
[73]: from sklearn.linear_model import LinearRegression
      lr = LinearRegression()
      lr.fit(X_train, y_train)
      y_pred = lr.predict(X_test)
      print('R2 Score: {:.3f}'.format(r2_score(y_test, y_pred)))
     R2 Score: 0.976
[76]: # MlpRegressor
      from sklearn.neural_network import MLPRegressor
      mlp = MLPRegressor(hidden_layer_sizes=(128,), activation='relu', solver='adam',_
       →max_iter=5000, random_state=42)
      y_train = y_train.reshape(-1)
      mlp.fit(X_train, y_train)
[76]: MLPRegressor(hidden_layer_sizes=(128,), max_iter=5000, random_state=42)
[77]: y_train.shape
[77]: (750,)
[78]: y_pred = mlp.predict(X_test)
      y_test = y_test.reshape(-1)
      print('R2 Score: {:.3f}'.format(r2_score(y_test, y_pred)))
     R2 Score: 0.976
[84]: from sklearn.datasets import make_classification
      X, y = make_classification(n_samples=1000, n_features=10, n_informative=5,__
       ⇔n_redundant=5, random_state=42)
      y = y.reshape(-1, 1)
      from sklearn.model_selection import train_test_split
      X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=42)
      from sklearn.preprocessing import StandardScaler
      scaler = StandardScaler()
```

```
X_train = scaler.fit_transform(X_train)
      X_test = scaler.transform(X_test)
[85]: nn = NeuralNetwork(X_train, y_train, n_hidden_neurons=128,__

→output_act_fn='sigmoid', error_fn='cross_entropy')
      wih, bh, who, bo = nn.train(X_train, y_train, learning_rate=0.001, epochs=5000)
     Epoch: 0, Loss: 5.929
     Epoch: 500, Loss: 4.358
     Epoch: 1000, Loss: 2.837
     Epoch: 1500, Loss: 0.827
     Epoch: 2000, Loss: 0.657
     Epoch: 2500, Loss: 0.580
     Epoch: 3000, Loss: 0.525
     Epoch: 3500, Loss: 0.486
     Epoch: 4000, Loss: 0.454
     Epoch: 4500, Loss: 0.424
     Trainig complete!
[89]: # Weights and biases
      print('Input-hidden weights:\n', wih)
      print('Hidden-output weights:\n', who)
      print('Hidden biases:\n', bh)
      print('Output bias:\n', bo)
     Input-hidden weights:
      [[ 4.91183828e-01 -1.00468647e-01 6.44106378e-01 ... 2.18844993e+00
       -9.56418207e-01 -4.09697867e-01]
      [ 9.15288413e-02 -4.90635188e-01 -1.53993291e+00 ... 1.04335534e+00
       -1.47839834e+00 -4.42668812e-01]
      [ 1.26456992e+00 -7.14245936e-01 4.96105956e-01 ... -6.09409034e-01
       -2.14842526e+00 -5.82169478e-01]
      [-1.53604276e-01 \ 1.68445391e-01 \ 1.13970488e+00 \ ... \ -5.75899104e-01
       -2.39451060e-02 2.23720422e+00]
      [ 1.72978081e+00 4.83170695e-01 -1.35860334e-03 ... -8.48581350e-01
       -9.56506553e-01 -1.96117134e+00]
      [-6.46226342e-01 -1.31219010e+00 1.66662767e+00 ... 3.94603584e-01
        2.12066826e-01 -2.76227811e-01]]
     Hidden-output weights:
      [[-2.62953365e-01]
      [-3.29512105e-01]
      [ 1.20847033e+00]
      [ 3.13681443e-01]
      [ 3.67716485e-01]
      [ 2.14878029e+00]
      [-9.06006210e-01]
      [7.28395787e-01]
```

- [ 2.49850589e-01]
- [ 1.79070340e-01]
- [ 1.62902030e+00]
- [ 1.50191246e-01]
- [ 6.30643759e-01]
- [ 7.44890741e-01]
- [-3.55293161e-01]
- [-1.16744705e+00]
- [-1.11634281e+00]
- [-2.08801813e+00]
- [ 1.94963110e+00]
- [-9.62831219e-01]
- [-2.01166385e-01]
- [-4.93545795e-01]
- [ 3.14794395e-01]
- [ 9.12193259e-01]
- [ 7.69160070e-01]
- [-5.09639344e-01]
- [-8.08503772e-02]
- [ 2.20992921e-01]
- [-3.52910269e-01]
- [ 1.29207373e+00]
- [ 1:2020/0/00:00]
- [ 4.65078921e-01]
- [ 6.73924164e-01]
- [ 3.79560790e-01]
- [ 1.02327370e+00]
- [-1.72077933e-01]
- [-1.20128929e+00]
- [ 9.38372818e-02]
- [-3.49494225e-01]
- [-5.76451809e-01]
- [ 5.41559672e-01]
- [ 1.18880083e+00]
- [-7.78755940e-02]
- [ 5.01976625e-01]
- [ 1.65749483e+00]
- [-2.86965768e-02]
- [ 2.60216650e+00]
- [ 5.59467650e-01]
- [-1.66037264e+00]
- [-1.000372040+00]
- [ 7.14018871e-01]
- [ 3.32919905e-01]
- [ 1.13311262e+00]
- [ 3.85118614e-01]
- [-3.44928082e-01]
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- [-3.00449579e-01]
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- [-4.07347343e-01]
- [ 3.95955381e-01]
- [ 1.67934345e+00]
- [ 3.11628861e-01]
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- [ 3.02528376e-01]
- [ 1.14150893e+00]
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- [-2.04818232e-01]
- [-7.91226509e-02]
- [-8.29164617e-01]
- [ 4.58560775e-01] [-2.16914496e-01]
- [ 2.98886535e-01]
- [-8.81705287e-01]
- [-4.57696367e-01] [ 6.98224350e-01]
- [-3.21102981e-01]
- 0.211020010 01
- [ 1.81459685e+00]
- [-3.97360629e-01]
- [ 1.07694466e-01]
- [ 1.04910342e+00]
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- [ 7.91287626e-01]
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- [ 2.53968644e+00]
- [-7.04036200e-01] [ 3.53896673e-02]
- [ 1.51545031e+00]
- [ 2.07393002e+00]
- [ 2.06609933e+00]
- [ 1.07752347e+00]
- [ 8.06221241e-01]
- [ 1.33097768e-01]
- [ 6.12878291e-01]
- [-5.15349945e-01]
- [-1.45379295e+00]
- [ 4.56259955e-02]
- [-1.43021254e-01]
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- [ 6.13906711e-01]
- [ 8.12328265e-01]
- [ 2.84189715e-02]
- [ 1.47680333e+00]
- [-1.36056616e+00]
- [-6.51691554e-02]

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[-8.49853393e-01]
 [-1.86319506e+00]
 [-2.65477942e-01]
 [-6.54351808e-01]
 [ 1.52254581e+00]
 [ 5.83214630e-01]
 [-1.16891117e+00]
 [ 1.61803160e+00]
 [ 8.49584949e-01]
 [ 4.63230371e-01]
 [-2.55350397e-01]
 [-1.13597249e+00]
 [-4.07339336e-01]
 [ 9.05416839e-01]
 [ 1.84922231e+00]
 [ 1.44765154e+00]
 [-5.22369209e-01]
 [-1.12947191e+00]
 [ 2.45430768e-03]
 [-1.71877497e+00]
 [ 2.15801035e-02]
 [-2.51423751e-01]
 [-3.59440498e-01]
 [-1.56587912e+00]]
Hidden biases:
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-8.20916545e-01 -1.36609613e+00 -1.77491381e+00 4.22247188e-01
-1.10081470e+00 -2.15739063e+00 3.66696126e-01 2.45745054e+00
 2.19488430e-03 8.15567871e-01 7.82697777e-02 -4.84083570e-03
 9.72036244e-01 -1.85451013e-01 1.35197124e-01 3.37483153e-01
-6.70137701e-01 1.03576276e+00 -1.69921258e-01 -7.69145973e-01
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 1.10251096e+00 3.27580979e-01 -4.74393310e-02 7.29487603e-01
 1.35969231e+00 4.57646189e-01 1.55932770e-03 -1.31511416e+00
-1.09375048e+00 -2.89926459e-01 -5.82521663e-01 -2.14709735e-01
-8.72640002e-02 5.35119892e-01 -8.91716563e-02 3.54203591e-01
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-1.23036626e+00 5.72936000e-01 1.35367806e+00 1.17017337e+00
-2.05834951e-01 1.59255845e+00 1.64583245e-01 -3.12462346e-01
-6.19756921e-01 -3.39757709e-01 -3.23994426e-01 1.38817512e-01
 2.00547472e-01 -1.61303620e-02 3.47276110e-01 1.21213906e+00
 9.76759234e-01 -1.48536395e+00 -2.51383502e+00 8.89003020e-01
-1.35378062e+00 -2.43809373e-01 -1.17243364e+00 -1.78921225e+00
 4.97402028e-01 7.71311955e-01 -5.64867749e-01 -2.59446949e+00
-5.60251648e-01 3.41921728e-01 -1.49989368e+00 2.71897591e-01
 1.85622729e-01 5.06162968e-01 1.65916671e-01 -1.08088382e+00
 1.26429254e+00 -1.66665792e-01 -4.88947958e-02 -9.80846846e-01
-4.83571012e-01 -9.59810243e-01 -2.84250508e-01 1.66395775e+00
```

```
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      -6.70189863e-01 1.07337435e+00 -6.48697237e-01 1.72914071e+00
       6.80467186e-01 -5.21072231e-01 2.27513730e+00 -6.00428720e-01
       8.64144802e-01 4.80856513e-01 1.32319774e+00 1.58728203e+00
      -1.70327417e-02 -7.70037001e-01 6.03358522e-01 -7.89145683e-01
       1.90108268e+00 9.62479779e-02 -3.66931436e-01 2.40674264e-01
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      -1.03845233e+00 5.70107233e-01 1.18231294e+00 8.29450545e-01
       9.77880161e-01 -7.50850031e-01 -1.41932568e+00 1.67609508e+00]
     Output bias:
      [0.40399672]
[86]: y_pred = nn.predict(X_test)
      from sklearn.metrics import accuracy_score
      print('Accuracy: {:.3f}'.format(accuracy_score(y_test, y_pred)))
     Accuracy: 0.908
[87]: # Logistic Regression
      from sklearn.linear_model import LogisticRegression
      lr = LogisticRegression()
      y_train = y_train.reshape(-1)
      lr.fit(X_train, y_train)
      y_pred = lr.predict(X_test)
      print('Accuracy: {:.3f}'.format(accuracy_score(y_test, y_pred)))
     Accuracy: 0.792
[88]: # MLPClassifier
      from sklearn.neural_network import MLPClassifier
      mlp = MLPClassifier(hidden_layer_sizes=(128,), activation='relu',_
       →solver='adam', max_iter=5000, random_state=42)
      mlp.fit(X train, y train)
      y pred = mlp.predict(X test)
      print('Accuracy: {:.3f}'.format(accuracy_score(y_test, y_pred)))
     Accuracy: 0.940
 []:
```