29/05/2022 17:58 Keras_First_NN_HW

Using Keras to Build and Train Neural Networks

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In this exercise we will use a neural network to predict diabetes using the Pima Diabetes Dataset. We will start by training a Random Forest to get a performance baseline. Then we will use the Keras package to quickly build and train a neural network and compare the performance. We will see how different network structures affect the performance, training time, and level of overfitting (or underfitting).

UCI Pima Diabetes Dataset

UCI ML Repositiory (http://archive.ics.uci.edu/ml/datasets/Pima+Indians+Diabetes)

Attributes: (all numeric-valued)

- 1. Number of times pregnant
- 2. Plasma glucose concentration a 2 hours in an oral glucose tolerance test
- 3. Diastolic blood pressure (mm Hg)
- 4. Triceps skin fold thickness (mm)
- 5. 2-Hour serum insulin (mu U/ml)
- 6. Body mass index (weight in kg/(height in m)^2)
- 7. Diabetes pedigree function
- 8. Age (years)
- 9. Class variable (0 or 1)

The UCI Pima Diabetes Dataset which has 8 numerical predictors and a binary outcome.

```
In [1]: #Preliminaries

from __future__ import absolute_import, division, print_function # Python 2/3 compatibility

import warnings
warnings.filterwarnings("ignore")
```

```
import numpy as np
import pandas as pd
import matplotlib.pvplot as plt
#Split arrays or matrices into random train and test subsets.
from sklearn.model selection import train test split
#Standardize features by removing the mean and scaling to unit variance.
from sklearn.preprocessing import StandardScaler
#confusion matrix
#precision recall curve: Compute precision-recall pairs for different probability thresholds.
#roc auc score:
#roc curve:sensitivity and specificity
#accuracy score:summarize the ROC curve in a single value, aggregating all ROC thresholds
from sklearn.metrics import confusion matrix, precision recall curve, roc auc score, roc curve, accuracy score
#prediction by committee is more accurate than that of any individual tree
from sklearn.ensemble import RandomForestClassifier
#Python data visualization library based on matplotlib
import seaborn as sns
%matplotlib inline
```

Install Keras:

```
In [2]:
```

```
!pip install keras
!pip install tensorflow
```

Requirement already satisfied: keras in /home/aritana/my_jupyter_notebook/my_jupyter_notebook/lib/python3.8/site-packag es (2.9.0)

WARNING: You are using pip version 21.2.4; however, version 22.1.1 is available.

You should consider upgrading via the '/home/aritana/my_jupyter_notebook/my_jupyter_notebook/bin/python -m pip install --upgrade pip' command.

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Requirement already satisfied: oauthlib>=3.0.0 in /home/aritana/my jupyter notebook/my jupyter notebook/lib/python3.8/s
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2.0)
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WARNING: You are using pip version 21.2.4; however, version 22.1.1 is available.
You should consider upgrading via the '/home/aritana/my jupyter notebook/my jupyter notebook/bin/python -m pip install
 --upgrade pip' command.
## Import Keras objects for Deep Learning
```

In [3]:

```
## Import Keras objects for Deep Learning
#sudo pip install keras

from keras.models import Sequential
#Input:is used to instantiate a Keras tensor
```

```
#Dense:implements the operation: output = activation(dot(input, kernel) + bias)
#Flatten:Flattens the input. Does not affect the batch size.
#Dropout:Dropout is a mechanism where in each training iteration (batch) we randomly remove a subset of neurons
#BatchNormalization
from keras.layers import Input, Dense, Flatten, Dropout, BatchNormalization
#Optimizers
from keras.optimizers import Adam, SGD, RMSprop
```

2022-05-29 17:13:31.027788: W tensorflow/stream_executor/platform/default/dso_loader.cc:64] Could not load dynamic libr ary 'libcudart.so.11.0'; dlerror: libcudart.so.11.0: cannot open shared object file: No such file or directory 2022-05-29 17:13:31.027844: I tensorflow/stream_executor/cuda/cudart_stub.cc:29] Ignore above cudart dlerror if you do not have a GPU set up on your machine.

Take a peek at the data -- if there are lots of "NaN" you may have internet connectivity issues print(diabetes_df.shape) diabetes_df.sample(5)

(768, 9)

Out[5]:		times_pregnant	glucose_tolerance_test	blood_pressure	skin_thickness	insulin	bmi	pedigree_function	age	has_diabetes
	737	8	65	72	23	0	32.0	0.600	42	0
	29	5	117	92	0	0	34.1	0.337	38	0
	439	6	107	88	0	0	36.8	0.727	31	0
	209	7	184	84	33	0	35.5	0.355	41	1
	206	8	196	76	29	280	37.5	0.605	57	1

```
In [6]: X = diabetes_df.iloc[:, :-1].values
y = diabetes_df["has_diabetes"].values
```

In [7]:

```
# Split the data to Train, and Test (75%, 25%)

#Because the data in the testing set already contains known values for the attribute
#that you want to predict, it is easy to determine whether the model's guesses are correct

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state=11111)
```

```
In [8]: np.mean(y), np.mean(1-y)
```

Out[8]: (0.348958333333333, 0.651041666666666)

Above, we see that about 35% of the patients in this dataset have diabetes, while 65% do not. This means we can get an accuracy of 65% without any model - just declare that no one has diabetes. We will calculate the ROC-AUC score to evaluate performance of our model, and also look at the accuracy as well to see if we improved upon the 65% accuracy.

Exercise: Get a baseline performance using Random Forest

To begin, and get a baseline for classifier performance:

- 1. Train a Random Forest model with 200 trees on the training data.
- 2. Calculate the accuracy and roc auc score of the predictions.

```
In [9]: ## Train the RF Model
    rf_model = RandomForestClassifier(n_estimators=200)
    rf_model.fit(X_train, y_train)

Out[9]: RandomForestClassifier(n_estimators=200)
```

```
In [10]:
# Make predictions on the test set - both "hard" predictions, and the scores (percent of trees voting yes)
y_pred_class_rf = rf_model.predict(X_test)
y_pred_prob_rf = rf_model.predict_proba(X_test)

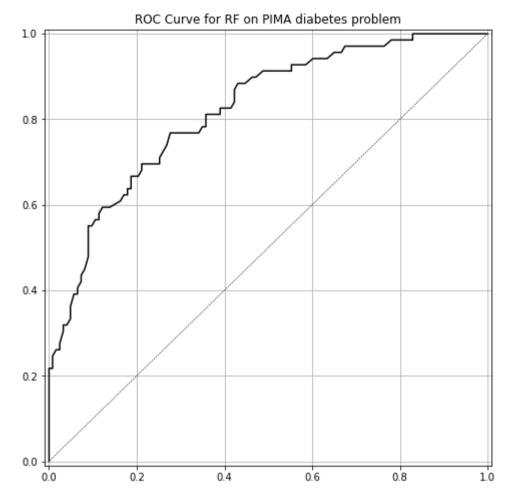
print('accuracy is {:.3f}'.format(accuracy_score(y_test,y_pred_class_rf)))
print('roc-auc is {:.3f}'.format(roc_auc_score(y_test,y_pred_prob_rf[:,1])))
```

```
accuracy is 0.755 roc-auc is 0.821
```

accuracy is 0.786 : Indica o número de previsões corretas no modelo. Neste caso 79% de acurácia, aproximadamente.

roc-auc is 0.832: Indica a media ou grau de separação, em como o modelo é capaz de distinguir entre as classes. É um valor próximo a 1, indica que o modelo está predizendo as classes de forma correta.

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Build a Single Hidden Layer Neural Network

We will use the Sequential model to quickly build a neural network. Our first network will be a single layer network. We have 8 variables, so we set the input shape to 8. Let's start by having a single hidden layer with 12 nodes.

```
## First let's normalize the data
## This aids the training of neural nets by providing numerical stability
## Random Forest does not need this as it finds a split only, as opposed to performing matrix multiplications
```

```
normalizer = StandardScaler()
          X train norm = normalizer.fit transform(X train)
          X test norm = normalizer.transform(X test)
In [13]:
          # Define the Model
          # Input size is 8-dimensional
          # 1 hidden layer, 12 hidden nodes, sigmoid activation
          # Final layer has just one node with a sigmoid activation (standard for binary classification)
          model 1 = Sequential([
              Dense(12, input shape=(8,), activation="relu"),
              Dense(1, activation="sigmoid")
          1)
         2022-05-29 17:13:39.738335: E tensorflow/stream executor/cuda/cuda driver.cc:271] failed call to cuInit: CUDA_ERROR_NO_
         DEVICE: no CUDA-capable device is detected
         2022-05-29 17:13:39.738570: I tensorflow/stream executor/cuda/cuda diagnostics.cc:156] kernel driver does not appear to
         be running on this host (aritana-Inspiron-5558): /proc/driver/nvidia/version does not exist
         2022-05-29 17:13:39.740360: I tensorflow/core/platform/cpu feature quard.cc:193] This TensorFlow binary is optimized wi
         th oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critical operation
         s: AVX2 FMA
         To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.
In [14]:
          # This is a nice tool to view the model you have created and count the parameters
          model 1.summary()
         Model: "sequential"
          Layer (type)
                                       Output Shape
          dense (Dense)
                                       (None, 12)
                                                                 108
          dense 1 (Dense)
                                       (None, 1)
                                                                 13
         Total params: 121
         Trainable params: 121
         Non-trainable params: 0
```

Comprehension guestion:

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Why do we have 121 parameters? Does that make sense?

Let's fit our model for 200 epochs.

121 parameters

- 1 hidden layer, 12 hidden nodes, sigmoid activation:
- 12 bias terms input.

Final layer has just one node with a sigmoid activation (standard for binary classification):

1 bias term input.

(12 * 8)conexões de entrada para camada oculta + 12 conexões coma camada de saida + 13 bias (12 com camada oculta e 1 com a camada de saída) = 121

```
In [15]:
     # Fit(Train) the Model
     # Compile the model with Optimizer, Loss Function and Metrics
     # Roc-Auc is not available in Keras as an off the shelf metric yet, so we will skip it here.
     model 1.compile(SGD(lr = .003), "binary crossentropy", metrics=["accuracy"])
     run hist 1 = model 1.fit(X train norm, y train, validation data=(X test norm, y test), epochs=200)
     # the fit function returns the run history.
     # It is very convenient, as it contains information about the model fit, iterations etc.
    Epoch 1/200
    cy: 0.5365
    Epoch 2/200
    cy: 0.5365
    Epoch 3/200
    y: 0.5521
    Epoch 4/200
    y: 0.5833
    Epoch 5/200
```

```
cv: 0.5833
Epoch 6/200
v: 0.5938
Epoch 7/200
v: 0.6094
Epoch 8/200
v: 0.6042
Epoch 9/200
cy: 0.6094
Epoch 10/200
v: 0.6302
Epoch 11/200
v: 0.6302
Epoch 12/200
v: 0.6302
Epoch 13/200
v: 0.6406
Epoch 14/200
cy: 0.6354
Epoch 15/200
v: 0.6406
Epoch 16/200
v: 0.6458
Epoch 17/200
v: 0.6562
Epoch 18/200
v: 0.6719
Epoch 19/200
y: 0.6719
Epoch 20/200
v: 0.6771
Epoch 21/200
```

```
v: 0.6823
Epoch 22/200
v: 0.6875
Epoch 23/200
v: 0.6875
Epoch 24/200
v: 0.6823
Epoch 25/200
v: 0.6927
Epoch 26/200
v: 0.6875
Epoch 27/200
v: 0.6927
Epoch 28/200
v: 0.6927
Epoch 29/200
v: 0.6927
Epoch 30/200
v: 0.6979
Epoch 31/200
cy: 0.7083
Epoch 32/200
v: 0.7083
Epoch 33/200
v: 0.7083
Epoch 34/200
v: 0.7083
Epoch 35/200
y: 0.7135
Epoch 36/200
y: 0.7135
```

```
Epoch 37/200
v: 0.7135
Epoch 38/200
y: 0.7135
Epoch 39/200
v: 0.7240
Epoch 40/200
v: 0.7240
Epoch 41/200
v: 0.7240
Epoch 42/200
v: 0.7240
Epoch 43/200
v: 0.7240
Epoch 44/200
v: 0.7240
Epoch 45/200
v: 0.7292
Epoch 46/200
v: 0.7292
Epoch 47/200
v: 0.7292
Epoch 48/200
cy: 0.7292
Epoch 49/200
v: 0.7292
Epoch 50/200
v: 0.7292
Epoch 51/200
y: 0.7292
Epoch 52/200
```

```
v: 0.7292
Epoch 53/200
v: 0.7292
Epoch 54/200
v: 0.7292
Epoch 55/200
v: 0.7292
Epoch 56/200
v: 0.7344
Epoch 57/200
v: 0.7396
Epoch 58/200
v: 0.7396
Epoch 59/200
v: 0.7396
Epoch 60/200
v: 0.7448
Epoch 61/200
cy: 0.7448
Epoch 62/200
v: 0.7448
Epoch 63/200
v: 0.7396
Epoch 64/200
v: 0.7396
Epoch 65/200
v: 0.7396
Epoch 66/200
y: 0.7396
Epoch 67/200
v: 0.7344
Epoch 68/200
```

```
v: 0.7344
Epoch 69/200
v: 0.7344
Epoch 70/200
v: 0.7344
Epoch 71/200
v: 0.7344
Epoch 72/200
v: 0.7344
Epoch 73/200
v: 0.7344
Epoch 74/200
v: 0.7344
Epoch 75/200
v: 0.7344
Epoch 76/200
v: 0.7344
Epoch 77/200
v: 0.7396
Epoch 78/200
v: 0.7396
Epoch 79/200
v: 0.7396
Epoch 80/200
v: 0.7448
Epoch 81/200
v: 0.7448
Epoch 82/200
y: 0.7448
Epoch 83/200
y: 0.7448
```

```
Epoch 84/200
v: 0.7448
Epoch 85/200
y: 0.7500
Epoch 86/200
v: 0.7500
Epoch 87/200
v: 0.7500
Epoch 88/200
v: 0.7500
Epoch 89/200
v: 0.7500
Epoch 90/200
v: 0.7500
Epoch 91/200
v: 0.7500
Epoch 92/200
v: 0.7500
Epoch 93/200
v: 0.7500
Epoch 94/200
v: 0.7500
Epoch 95/200
v: 0.7500
Epoch 96/200
cy: 0.7500
Epoch 97/200
v: 0.7500
Epoch 98/200
y: 0.7500
Epoch 99/200
```

```
y: 0.7500
Epoch 100/200
v: 0.7552
Epoch 101/200
v: 0.7552
Epoch 102/200
v: 0.7552
Epoch 103/200
v: 0.7552
Epoch 104/200
v: 0.7552
Epoch 105/200
v: 0.7552
Epoch 106/200
v: 0.7552
Epoch 107/200
cy: 0.7552
Epoch 108/200
v: 0.7552
Epoch 109/200
v: 0.7552
Epoch 110/200
v: 0.7552
Epoch 111/200
cv: 0.7552
Epoch 112/200
cv: 0.7552
Epoch 113/200
v: 0.7552
Epoch 114/200
v: 0.7552
Epoch 115/200
```

```
cv: 0.7552
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v: 0.7552
Epoch 117/200
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Epoch 120/200
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Epoch 122/200
v: 0.7604
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Epoch 124/200
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Epoch 125/200
v: 0.7604
Epoch 126/200
v: 0.7604
Epoch 127/200
v: 0.7552
Epoch 128/200
v: 0.7552
Epoch 129/200
y: 0.7552
Epoch 130/200
y: 0.7552
```

```
Epoch 131/200
cv: 0.7552
Epoch 132/200
cy: 0.7552
Epoch 133/200
v: 0.7552
Epoch 134/200
v: 0.7552
Epoch 135/200
v: 0.7552
Epoch 136/200
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Epoch 137/200
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Epoch 138/200
v: 0.7604
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Epoch 141/200
v: 0.7604
Epoch 142/200
v: 0.7604
Epoch 143/200
v: 0.7604
Epoch 144/200
v: 0.7604
Epoch 145/200
y: 0.7604
Epoch 146/200
```

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v: 0.7604
Epoch 147/200
v: 0.7604
Epoch 148/200
v: 0.7604
Epoch 149/200
v: 0.7656
Epoch 150/200
v: 0.7656
Epoch 151/200
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v: 0.7656
Epoch 157/200
v: 0.7656
Epoch 158/200
v: 0.7656
Epoch 159/200
v: 0.7656
Epoch 160/200
y: 0.7656
Epoch 161/200
v: 0.7604
Epoch 162/200
```

```
v: 0.7656
Epoch 163/200
cv: 0.7656
Epoch 164/200
v: 0.7656
Epoch 165/200
v: 0.7656
Epoch 166/200
cy: 0.7656
Epoch 167/200
cy: 0.7656
Epoch 168/200
cy: 0.7656
Epoch 169/200
v: 0.7656
Epoch 170/200
v: 0.7656
Epoch 171/200
v: 0.7656
Epoch 172/200
v: 0.7656
Epoch 173/200
v: 0.7656
Epoch 174/200
v: 0.7656
Epoch 175/200
v: 0.7656
Epoch 176/200
y: 0.7656
Epoch 177/200
y: 0.7656
```

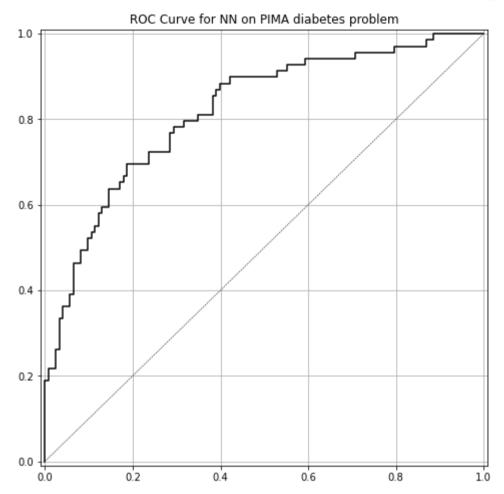
```
Epoch 178/200
v: 0.7656
Epoch 179/200
v: 0.7604
Epoch 180/200
v: 0.7604
Epoch 181/200
cy: 0.7604
Epoch 182/200
v: 0.7604
Epoch 183/200
v: 0.7604
Epoch 184/200
v: 0.7604
Epoch 185/200
cy: 0.7604
Epoch 186/200
cv: 0.7604
Epoch 187/200
cy: 0.7604
Epoch 188/200
cy: 0.7604
Epoch 189/200
cy: 0.7604
Epoch 190/200
cv: 0.7604
Epoch 191/200
v: 0.7604
Epoch 192/200
cy: 0.7604
Epoch 193/200
```

```
v: 0.7604
    Epoch 194/200
    cv: 0.7604
    Epoch 195/200
    cv: 0.7604
    Epoch 196/200
    cv: 0.7604
    Epoch 197/200
    cv: 0.7604
    Epoch 198/200
    v: 0.7604
    Epoch 199/200
    v: 0.7604
    Epoch 200/200
    v: 0.7604
In [16]:
     ## Like we did for the Random Forest, we generate two kinds of predictions
     # One is a hard decision, the other is a probabilitistic score.
     #v pred class nn 1 = model 1.predict classes(X test norm)
     y_pred_prob_nn_1 = model 1.predict(X test norm)
     predict x = model 1.predict(X test)
     y pred class nn 1 = np.argmax(predict x,axis=1)
     #0 método "predict classes" não funciona nas versões do Keras superior a 2.5
     # A indicação da correção se encontra em https://keras.rstudio.com/reference/predict proba.html#details.
     # usar:
     #y pred class nn 1 = model 1.predict(X test norm)
     #y pred class nn 1 = (y \text{ pred prob nn } 1 > 0.5).astype("int32")
    6/6 [=======] - 0s 7ms/step
    6/6 [======= ] - Os 4ms/step
```

In [17]:

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```
# Let's check out the outputs to get a feel for how keras apis work.
          y pred class nn 1[:10]
Out[17]: array([0, 0, 0, 0, 0, 0, 0, 0, 0])
In [18]:
          y pred prob nn 1[:10]
Out[18]: array([[0.53559816],
                [0.56501013],
                [0.33758485],
                 [0.2791096],
                [0.20052065],
                 [0.59508187],
                [0.04426837],
                [0.35369265],
                [0.7867891],
                [0.17987987]], dtype=float32)
In [19]:
          # Print model performance and plot the roc curve
          print('accuracy is {:.3f}'.format(accuracy score(y test,y pred class nn 1)))
          print('roc-auc is {:.3f}'.format(roc auc score(y test,y pred prob nn 1)))
          plot_roc(y_test, y_pred_prob_nn_1, 'NN')
         accuracy is 0.641
         roc-auc is 0.819
```



There may be some variation in exact numbers due to randomness, but you should get results in the same ballpark as the Random Forest - between 75% and 85% accuracy, between .8 and .9 for AUC.

Let's look at the run_hist_1 object that was created, specifically its history attribute.

```
In [20]: run_hist_1.history.keys()
```

Out[20]: dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])

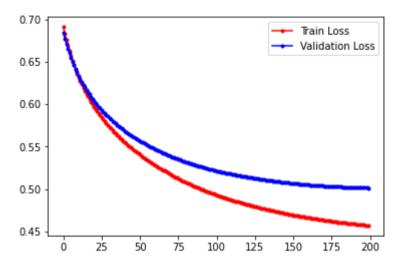
Let's plot the training loss and the validation loss over the different epochs and see how it looks.

Keras First NN HW

```
In [21]: fig, ax = plt.subplots()
    ax.plot(run_hist_1.history["loss"],'r', marker='.', label="Train Loss")
    ax.plot(run_hist_1.history["val_loss"],'b', marker='.', label="Validation Loss")
    ax.legend()
```

Out[21]: <matplotlib.legend.Legend at 0x7f931979ecd0>

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Looks like the losses are still going down on both the training set and the validation set. This suggests that the model might benefit from further training. Let's train the model a little more and see what happens. Note that it will pick up from where it left off. Train for 1000 more epochs.

```
In [22]:
   ## Note that when we call "fit" again, it picks up where it left off
   run hist 1b = model 1.fit(X train norm, y train, validation data=(X test norm, y test), epochs=1000)
   Epoch 1/1000
   cy: 0.7604
   Epoch 2/1000
   cy: 0.7604
   Epoch 3/1000
   cv: 0.7604
   Epoch 4/1000
   y: 0.7604
   Epoch 5/1000
```

```
v: 0.7604
Epoch 6/1000
cv: 0.7604
Epoch 7/1000
cv: 0.7604
Epoch 8/1000
cv: 0.7604
Epoch 9/1000
v: 0.7604
Epoch 10/1000
cy: 0.7604
Epoch 11/1000
v: 0.7604
Epoch 12/1000
v: 0.7604
Epoch 13/1000
v: 0.7604
Epoch 14/1000
cy: 0.7604
Epoch 15/1000
cy: 0.7604
Epoch 16/1000
v: 0.7604
Epoch 17/1000
v: 0.7604
Epoch 18/1000
y: 0.7604
Epoch 19/1000
y: 0.7604
Epoch 20/1000
v: 0.7604
Epoch 21/1000
```

```
v: 0.7604
Epoch 22/1000
v: 0.7604
Epoch 23/1000
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Epoch 24/1000
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Epoch 25/1000
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Epoch 27/1000
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Epoch 28/1000
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Epoch 29/1000
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Epoch 31/1000
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Epoch 32/1000
v: 0.7604
Epoch 33/1000
v: 0.7604
Epoch 34/1000
v: 0.7604
Epoch 35/1000
y: 0.7604
Epoch 36/1000
y: 0.7604
```

```
Epoch 37/1000
v: 0.7604
Epoch 38/1000
v: 0.7604
Epoch 39/1000
v: 0.7604
Epoch 40/1000
v: 0.7604
Epoch 41/1000
v: 0.7604
Epoch 42/1000
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Epoch 43/1000
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Epoch 44/1000
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Epoch 45/1000
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Epoch 46/1000
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Epoch 47/1000
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Epoch 48/1000
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Epoch 49/1000
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Epoch 50/1000
v: 0.7604
Epoch 51/1000
y: 0.7604
Epoch 52/1000
```

```
v: 0.7604
Epoch 53/1000
v: 0.7604
Epoch 54/1000
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Epoch 55/1000
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Epoch 62/1000
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Epoch 63/1000
v: 0.7604
Epoch 64/1000
v: 0.7604
Epoch 65/1000
y: 0.7604
Epoch 66/1000
v: 0.7604
Epoch 67/1000
v: 0.7604
Epoch 68/1000
```

```
v: 0.7656
Epoch 69/1000
v: 0.7656
Epoch 70/1000
v: 0.7656
Epoch 71/1000
v: 0.7656
Epoch 72/1000
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Epoch 73/1000
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Epoch 78/1000
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Epoch 79/1000
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Epoch 80/1000
v: 0.7656
Epoch 81/1000
cy: 0.7708
Epoch 82/1000
cy: 0.7708
Epoch 83/1000
cy: 0.7708
```

```
Epoch 84/1000
v: 0.7708
Epoch 85/1000
y: 0.7708
Epoch 86/1000
v: 0.7708
Epoch 87/1000
v: 0.7708
Epoch 88/1000
cy: 0.7708
Epoch 89/1000
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Epoch 90/1000
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Epoch 93/1000
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Epoch 94/1000
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Epoch 95/1000
v: 0.7760
Epoch 96/1000
v: 0.7760
Epoch 97/1000
v: 0.7760
Epoch 98/1000
y: 0.7760
Epoch 99/1000
```

```
cy: 0.7760
Epoch 100/1000
cv: 0.7760
Epoch 101/1000
v: 0.7760
Epoch 102/1000
v: 0.7760
Epoch 103/1000
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Epoch 104/1000
v: 0.7760
Epoch 105/1000
cy: 0.7760
Epoch 106/1000
cy: 0.7760
Epoch 107/1000
cy: 0.7760
Epoch 108/1000
v: 0.7760
Epoch 109/1000
cy: 0.7760
Epoch 110/1000
cy: 0.7760
Epoch 111/1000
cy: 0.7760
Epoch 112/1000
cy: 0.7760
Epoch 113/1000
cy: 0.7760
Epoch 114/1000
v: 0.7760
Epoch 115/1000
```

```
cv: 0.7760
Epoch 116/1000
v: 0.7760
Epoch 117/1000
cy: 0.7760
Epoch 118/1000
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Epoch 128/1000
v: 0.7708
Epoch 129/1000
y: 0.7708
Epoch 130/1000
cy: 0.7708
```

```
Epoch 131/1000
v: 0.7708
Epoch 132/1000
y: 0.7708
Epoch 133/1000
cy: 0.7708
Epoch 134/1000
cy: 0.7708
Epoch 135/1000
v: 0.7708
Epoch 136/1000
v: 0.7708
Epoch 137/1000
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Epoch 138/1000
y: 0.7708
Epoch 139/1000
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y: 0.7708
Epoch 146/1000
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v: 0.7708
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Epoch 161/1000
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Epoch 162/1000
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v: 0.7708
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v: 0.7656
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Epoch 176/1000
y: 0.7656
Epoch 177/1000
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Epoch 178/1000
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y: 0.7656
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Epoch 256/1000
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Epoch 272/1000
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y: 0.7552
Epoch 287/1000
```

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y: 0.7552
Epoch 288/1000
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Epoch 289/1000
v: 0.7552
Epoch 290/1000
cv: 0.7500
Epoch 291/1000
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v: 0.7500
Epoch 299/1000
cy: 0.7500
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Epoch 301/1000
y: 0.7500
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v: 0.7500
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v: 0.7500
Epoch 304/1000
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Epoch 316/1000
v: 0.7500
Epoch 317/1000
y: 0.7500
Epoch 318/1000
cy: 0.7500
```

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Epoch 319/1000
v: 0.7500
Epoch 320/1000
y: 0.7500
Epoch 321/1000
v: 0.7500
Epoch 322/1000
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Epoch 323/1000
v: 0.7500
Epoch 324/1000
v: 0.7500
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Epoch 326/1000
y: 0.7500
Epoch 327/1000
v: 0.7500
Epoch 328/1000
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Epoch 329/1000
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Epoch 330/1000
v: 0.7500
Epoch 331/1000
v: 0.7500
Epoch 332/1000
v: 0.7500
Epoch 333/1000
y: 0.7500
Epoch 334/1000
```

```
y: 0.7500
Epoch 335/1000
v: 0.7500
Epoch 336/1000
v: 0.7500
Epoch 337/1000
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Epoch 338/1000
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Epoch 339/1000
v: 0.7500
Epoch 340/1000
cy: 0.7500
Epoch 341/1000
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Epoch 342/1000
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Epoch 344/1000
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Epoch 345/1000
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y: 0.7500
Epoch 347/1000
v: 0.7500
Epoch 348/1000
cy: 0.7500
Epoch 349/1000
v: 0.7500
Epoch 350/1000
```

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v: 0.7500
Epoch 351/1000
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v: 0.7500
Epoch 353/1000
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Epoch 354/1000
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Epoch 355/1000
cy: 0.7500
Epoch 356/1000
cy: 0.7500
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Epoch 358/1000
v: 0.7500
Epoch 359/1000
cy: 0.7500
Epoch 360/1000
cy: 0.7500
Epoch 361/1000
v: 0.7500
Epoch 362/1000
v: 0.7500
Epoch 363/1000
v: 0.7552
Epoch 364/1000
y: 0.7552
Epoch 365/1000
y: 0.7552
```

```
Epoch 366/1000
v: 0.7552
Epoch 367/1000
y: 0.7552
Epoch 368/1000
v: 0.7552
Epoch 369/1000
v: 0.7552
Epoch 370/1000
v: 0.7552
Epoch 371/1000
v: 0.7552
Epoch 372/1000
v: 0.7552
Epoch 373/1000
cy: 0.7552
Epoch 374/1000
v: 0.7552
Epoch 375/1000
v: 0.7552
Epoch 376/1000
cy: 0.7552
Epoch 377/1000
cy: 0.7552
Epoch 378/1000
v: 0.7552
Epoch 379/1000
cv: 0.7552
Epoch 380/1000
cy: 0.7552
Epoch 381/1000
```

```
y: 0.7552
Epoch 382/1000
cv: 0.7552
Epoch 383/1000
v: 0.7552
Epoch 384/1000
v: 0.7552
Epoch 385/1000
v: 0.7552
Epoch 386/1000
v: 0.7552
Epoch 387/1000
cy: 0.7552
Epoch 388/1000
cy: 0.7552
Epoch 389/1000
v: 0.7552
Epoch 390/1000
cy: 0.7552
Epoch 391/1000
cy: 0.7552
Epoch 392/1000
cy: 0.7552
Epoch 393/1000
cv: 0.7552
Epoch 394/1000
v: 0.7552
Epoch 395/1000
cy: 0.7552
Epoch 396/1000
v: 0.7552
Epoch 397/1000
```

```
v: 0.7552
Epoch 398/1000
v: 0.7552
Epoch 399/1000
v: 0.7552
Epoch 400/1000
v: 0.7552
Epoch 401/1000
v: 0.7552
Epoch 402/1000
v: 0.7552
Epoch 403/1000
y: 0.7552
Epoch 404/1000
v: 0.7552
Epoch 405/1000
v: 0.7552
Epoch 406/1000
v: 0.7552
Epoch 407/1000
v: 0.7552
Epoch 408/1000
v: 0.7552
Epoch 409/1000
v: 0.7552
Epoch 410/1000
cy: 0.7552
Epoch 411/1000
cy: 0.7552
Epoch 412/1000
y: 0.7552
```

```
Epoch 413/1000
v: 0.7552
Epoch 414/1000
y: 0.7552
Epoch 415/1000
v: 0.7552
Epoch 416/1000
v: 0.7552
Epoch 417/1000
v: 0.7552
Epoch 418/1000
v: 0.7552
Epoch 419/1000
v: 0.7552
Epoch 420/1000
y: 0.7552
Epoch 421/1000
cy: 0.7552
Epoch 422/1000
cy: 0.7552
Epoch 423/1000
cy: 0.7552
Epoch 424/1000
cy: 0.7552
Epoch 425/1000
v: 0.7552
Epoch 426/1000
v: 0.7552
Epoch 427/1000
y: 0.7552
Epoch 428/1000
```

```
v: 0.7552
Epoch 429/1000
v: 0.7552
Epoch 430/1000
v: 0.7552
Epoch 431/1000
v: 0.7552
Epoch 432/1000
v: 0.7552
Epoch 433/1000
v: 0.7552
Epoch 434/1000
v: 0.7552
Epoch 435/1000
v: 0.7552
Epoch 436/1000
v: 0.7552
Epoch 437/1000
v: 0.7500
Epoch 438/1000
v: 0.7500
Epoch 439/1000
v: 0.7500
Epoch 440/1000
v: 0.7500
Epoch 441/1000
v: 0.7500
Epoch 442/1000
y: 0.7500
Epoch 443/1000
v: 0.7500
Epoch 444/1000
```

```
v: 0.7500
Epoch 445/1000
v: 0.7448
Epoch 446/1000
v: 0.7448
Epoch 447/1000
v: 0.7448
Epoch 448/1000
v: 0.7448
Epoch 449/1000
v: 0.7448
Epoch 450/1000
v: 0.7448
Epoch 451/1000
v: 0.7448
Epoch 452/1000
v: 0.7448
Epoch 453/1000
v: 0.7448
Epoch 454/1000
v: 0.7448
Epoch 455/1000
v: 0.7448
Epoch 456/1000
cy: 0.7448
Epoch 457/1000
cy: 0.7448
Epoch 458/1000
y: 0.7448
Epoch 459/1000
cy: 0.7448
```

```
Epoch 460/1000
v: 0.7448
Epoch 461/1000
y: 0.7448
Epoch 462/1000
v: 0.7448
Epoch 463/1000
v: 0.7448
Epoch 464/1000
v: 0.7448
Epoch 465/1000
v: 0.7448
Epoch 466/1000
cy: 0.7448
Epoch 467/1000
v: 0.7448
Epoch 468/1000
v: 0.7448
Epoch 469/1000
v: 0.7448
Epoch 470/1000
v: 0.7448
Epoch 471/1000
v: 0.7448
Epoch 472/1000
v: 0.7448
Epoch 473/1000
v: 0.7448
Epoch 474/1000
y: 0.7448
Epoch 475/1000
```

```
v: 0.7448
Epoch 476/1000
v: 0.7448
Epoch 477/1000
v: 0.7448
Epoch 478/1000
v: 0.7448
Epoch 479/1000
v: 0.7448
Epoch 480/1000
v: 0.7448
Epoch 481/1000
v: 0.7448
Epoch 482/1000
v: 0.7448
Epoch 483/1000
v: 0.7448
Epoch 484/1000
v: 0.7448
Epoch 485/1000
v: 0.7448
Epoch 486/1000
v: 0.7448
Epoch 487/1000
v: 0.7448
Epoch 488/1000
v: 0.7448
Epoch 489/1000
y: 0.7448
Epoch 490/1000
v: 0.7448
Epoch 491/1000
```

```
v: 0.7448
Epoch 492/1000
v: 0.7448
Epoch 493/1000
v: 0.7448
Epoch 494/1000
v: 0.7448
Epoch 495/1000
v: 0.7448
Epoch 496/1000
v: 0.7448
Epoch 497/1000
v: 0.7448
Epoch 498/1000
v: 0.7448
Epoch 499/1000
v: 0.7448
Epoch 500/1000
v: 0.7448
Epoch 501/1000
v: 0.7448
Epoch 502/1000
v: 0.7448
Epoch 503/1000
v: 0.7448
Epoch 504/1000
v: 0.7448
Epoch 505/1000
y: 0.7448
Epoch 506/1000
y: 0.7448
```

```
Epoch 507/1000
v: 0.7448
Epoch 508/1000
y: 0.7500
Epoch 509/1000
v: 0.7500
Epoch 510/1000
v: 0.7500
Epoch 511/1000
v: 0.7500
Epoch 512/1000
v: 0.7500
Epoch 513/1000
v: 0.7500
Epoch 514/1000
v: 0.7500
Epoch 515/1000
v: 0.7500
Epoch 516/1000
v: 0.7500
Epoch 517/1000
v: 0.7500
Epoch 518/1000
v: 0.7500
Epoch 519/1000
v: 0.7500
Epoch 520/1000
v: 0.7500
Epoch 521/1000
y: 0.7500
Epoch 522/1000
```

```
y: 0.7500
Epoch 523/1000
v: 0.7500
Epoch 524/1000
v: 0.7500
Epoch 525/1000
cy: 0.7500
Epoch 526/1000
cy: 0.7500
Epoch 527/1000
v: 0.7500
Epoch 528/1000
v: 0.7500
Epoch 529/1000
v: 0.7500
Epoch 530/1000
v: 0.7500
Epoch 531/1000
v: 0.7500
Epoch 532/1000
v: 0.7500
Epoch 533/1000
cy: 0.7500
Epoch 534/1000
cy: 0.7500
Epoch 535/1000
cy: 0.7500
Epoch 536/1000
cy: 0.7500
Epoch 537/1000
cv: 0.7500
Epoch 538/1000
```

```
v: 0.7500
Epoch 539/1000
v: 0.7500
Epoch 540/1000
v: 0.7500
Epoch 541/1000
v: 0.7500
Epoch 542/1000
cy: 0.7500
Epoch 543/1000
cy: 0.7500
Epoch 544/1000
cy: 0.7448
Epoch 545/1000
cy: 0.7448
Epoch 546/1000
cy: 0.7448
Epoch 547/1000
v: 0.7448
Epoch 548/1000
v: 0.7448
Epoch 549/1000
v: 0.7448
Epoch 550/1000
v: 0.7448
Epoch 551/1000
v: 0.7448
Epoch 552/1000
y: 0.7448
Epoch 553/1000
y: 0.7448
```

```
Epoch 554/1000
v: 0.7448
Epoch 555/1000
y: 0.7448
Epoch 556/1000
v: 0.7448
Epoch 557/1000
v: 0.7448
Epoch 558/1000
v: 0.7448
Epoch 559/1000
v: 0.7448
Epoch 560/1000
v: 0.7448
Epoch 561/1000
v: 0.7448
Epoch 562/1000
v: 0.7448
Epoch 563/1000
v: 0.7448
Epoch 564/1000
v: 0.7448
Epoch 565/1000
v: 0.7448
Epoch 566/1000
cv: 0.7448
Epoch 567/1000
cv: 0.7448
Epoch 568/1000
cy: 0.7448
Epoch 569/1000
```

```
v: 0.7448
Epoch 570/1000
cv: 0.7448
Epoch 571/1000
v: 0.7448
Epoch 572/1000
v: 0.7448
Epoch 573/1000
v: 0.7448
Epoch 574/1000
v: 0.7448
Epoch 575/1000
v: 0.7448
Epoch 576/1000
v: 0.7448
Epoch 577/1000
v: 0.7448
Epoch 578/1000
v: 0.7448
Epoch 579/1000
v: 0.7448
Epoch 580/1000
v: 0.7448
Epoch 581/1000
v: 0.7448
Epoch 582/1000
v: 0.7448
Epoch 583/1000
y: 0.7448
Epoch 584/1000
v: 0.7448
Epoch 585/1000
```

```
v: 0.7448
Epoch 586/1000
v: 0.7448
Epoch 587/1000
v: 0.7448
Epoch 588/1000
cy: 0.7448
Epoch 589/1000
v: 0.7448
Epoch 590/1000
v: 0.7448
Epoch 591/1000
v: 0.7448
Epoch 592/1000
cy: 0.7448
Epoch 593/1000
cy: 0.7448
Epoch 594/1000
cy: 0.7448
Epoch 595/1000
v: 0.7448
Epoch 596/1000
v: 0.7448
Epoch 597/1000
v: 0.7448
Epoch 598/1000
v: 0.7448
Epoch 599/1000
y: 0.7448
Epoch 600/1000
y: 0.7396
```

```
Epoch 601/1000
v: 0.7396
Epoch 602/1000
y: 0.7396
Epoch 603/1000
v: 0.7396
Epoch 604/1000
cy: 0.7396
Epoch 605/1000
v: 0.7396
Epoch 606/1000
v: 0.7396
Epoch 607/1000
v: 0.7396
Epoch 608/1000
v: 0.7396
Epoch 609/1000
v: 0.7396
Epoch 610/1000
v: 0.7396
Epoch 611/1000
v: 0.7396
Epoch 612/1000
v: 0.7396
Epoch 613/1000
v: 0.7396
Epoch 614/1000
v: 0.7448
Epoch 615/1000
cy: 0.7448
Epoch 616/1000
```

```
v: 0.7448
Epoch 617/1000
v: 0.7448
Epoch 618/1000
v: 0.7448
Epoch 619/1000
v: 0.7448
Epoch 620/1000
v: 0.7448
Epoch 621/1000
v: 0.7448
Epoch 622/1000
v: 0.7448
Epoch 623/1000
v: 0.7448
Epoch 624/1000
v: 0.7448
Epoch 625/1000
v: 0.7448
Epoch 626/1000
v: 0.7448
Epoch 627/1000
cy: 0.7448
Epoch 628/1000
cv: 0.7448
Epoch 629/1000
cy: 0.7448
Epoch 630/1000
y: 0.7448
Epoch 631/1000
v: 0.7448
Epoch 632/1000
```

```
v: 0.7448
Epoch 633/1000
v: 0.7448
Epoch 634/1000
v: 0.7448
Epoch 635/1000
v: 0.7448
Epoch 636/1000
v: 0.7448
Epoch 637/1000
v: 0.7448
Epoch 638/1000
v: 0.7448
Epoch 639/1000
v: 0.7448
Epoch 640/1000
v: 0.7448
Epoch 641/1000
v: 0.7448
Epoch 642/1000
v: 0.7448
Epoch 643/1000
v: 0.7448
Epoch 644/1000
v: 0.7448
Epoch 645/1000
v: 0.7448
Epoch 646/1000
y: 0.7448
Epoch 647/1000
y: 0.7448
```

```
Epoch 648/1000
v: 0.7448
Epoch 649/1000
y: 0.7448
Epoch 650/1000
v: 0.7448
Epoch 651/1000
v: 0.7448
Epoch 652/1000
cy: 0.7448
Epoch 653/1000
cy: 0.7448
Epoch 654/1000
v: 0.7448
Epoch 655/1000
cy: 0.7448
Epoch 656/1000
cy: 0.7448
Epoch 657/1000
v: 0.7448
Epoch 658/1000
cy: 0.7448
Epoch 659/1000
v: 0.7448
Epoch 660/1000
v: 0.7448
Epoch 661/1000
cy: 0.7448
Epoch 662/1000
y: 0.7448
Epoch 663/1000
```

```
v: 0.7448
Epoch 664/1000
v: 0.7448
Epoch 665/1000
cv: 0.7448
Epoch 666/1000
v: 0.7448
Epoch 667/1000
v: 0.7448
Epoch 668/1000
v: 0.7448
Epoch 669/1000
v: 0.7448
Epoch 670/1000
v: 0.7448
Epoch 671/1000
v: 0.7448
Epoch 672/1000
v: 0.7448
Epoch 673/1000
v: 0.7448
Epoch 674/1000
v: 0.7448
Epoch 675/1000
v: 0.7448
Epoch 676/1000
v: 0.7448
Epoch 677/1000
y: 0.7448
Epoch 678/1000
v: 0.7448
Epoch 679/1000
```

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```
v: 0.7448
Epoch 680/1000
v: 0.7448
Epoch 681/1000
v: 0.7448
Epoch 682/1000
v: 0.7448
Epoch 683/1000
v: 0.7448
Epoch 684/1000
v: 0.7448
Epoch 685/1000
v: 0.7448
Epoch 686/1000
v: 0.7448
Epoch 687/1000
v: 0.7448
Epoch 688/1000
v: 0.7448
Epoch 689/1000
v: 0.7448
Epoch 690/1000
v: 0.7448
Epoch 691/1000
v: 0.7448
Epoch 692/1000
v: 0.7448
Epoch 693/1000
y: 0.7448
Epoch 694/1000
y: 0.7448
```

```
Epoch 695/1000
v: 0.7448
Epoch 696/1000
y: 0.7448
Epoch 697/1000
v: 0.7448
Epoch 698/1000
v: 0.7448
Epoch 699/1000
v: 0.7448
Epoch 700/1000
v: 0.7448
Epoch 701/1000
v: 0.7448
Epoch 702/1000
v: 0.7448
Epoch 703/1000
v: 0.7448
Epoch 704/1000
v: 0.7448
Epoch 705/1000
v: 0.7448
Epoch 706/1000
v: 0.7448
Epoch 707/1000
v: 0.7448
Epoch 708/1000
v: 0.7448
Epoch 709/1000
v: 0.7448
Epoch 710/1000
```

```
v: 0.7448
Epoch 711/1000
v: 0.7448
Epoch 712/1000
v: 0.7448
Epoch 713/1000
v: 0.7448
Epoch 714/1000
v: 0.7448
Epoch 715/1000
v: 0.7448
Epoch 716/1000
v: 0.7448
Epoch 717/1000
v: 0.7448
Epoch 718/1000
v: 0.7448
Epoch 719/1000
v: 0.7448
Epoch 720/1000
v: 0.7448
Epoch 721/1000
v: 0.7448
Epoch 722/1000
v: 0.7448
Epoch 723/1000
v: 0.7448
Epoch 724/1000
y: 0.7448
Epoch 725/1000
v: 0.7448
Epoch 726/1000
```

```
v: 0.7448
Epoch 727/1000
v: 0.7448
Epoch 728/1000
v: 0.7448
Epoch 729/1000
v: 0.7448
Epoch 730/1000
v: 0.7448
Epoch 731/1000
v: 0.7448
Epoch 732/1000
v: 0.7448
Epoch 733/1000
v: 0.7448
Epoch 734/1000
v: 0.7448
Epoch 735/1000
v: 0.7448
Epoch 736/1000
v: 0.7448
Epoch 737/1000
v: 0.7448
Epoch 738/1000
v: 0.7448
Epoch 739/1000
v: 0.7448
Epoch 740/1000
y: 0.7448
Epoch 741/1000
y: 0.7448
```

```
Epoch 742/1000
v: 0.7448
Epoch 743/1000
y: 0.7448
Epoch 744/1000
v: 0.7448
Epoch 745/1000
v: 0.7448
Epoch 746/1000
v: 0.7448
Epoch 747/1000
v: 0.7448
Epoch 748/1000
v: 0.7448
Epoch 749/1000
v: 0.7448
Epoch 750/1000
v: 0.7448
Epoch 751/1000
v: 0.7448
Epoch 752/1000
v: 0.7448
Epoch 753/1000
v: 0.7448
Epoch 754/1000
v: 0.7448
Epoch 755/1000
v: 0.7448
Epoch 756/1000
v: 0.7448
Epoch 757/1000
```

```
v: 0.7448
Epoch 758/1000
v: 0.7448
Epoch 759/1000
v: 0.7448
Epoch 760/1000
v: 0.7448
Epoch 761/1000
v: 0.7448
Epoch 762/1000
v: 0.7448
Epoch 763/1000
v: 0.7448
Epoch 764/1000
v: 0.7448
Epoch 765/1000
v: 0.7448
Epoch 766/1000
v: 0.7448
Epoch 767/1000
v: 0.7448
Epoch 768/1000
v: 0.7448
Epoch 769/1000
y: 0.7448
Epoch 770/1000
v: 0.7448
Epoch 771/1000
cy: 0.7448
Epoch 772/1000
v: 0.7448
Epoch 773/1000
```

```
v: 0.7448
Epoch 774/1000
v: 0.7448
Epoch 775/1000
v: 0.7448
Epoch 776/1000
cy: 0.7448
Epoch 777/1000
v: 0.7448
Epoch 778/1000
v: 0.7448
Epoch 779/1000
v: 0.7448
Epoch 780/1000
v: 0.7448
Epoch 781/1000
v: 0.7448
Epoch 782/1000
v: 0.7448
Epoch 783/1000
v: 0.7448
Epoch 784/1000
v: 0.7448
Epoch 785/1000
v: 0.7448
Epoch 786/1000
v: 0.7448
Epoch 787/1000
y: 0.7448
Epoch 788/1000
y: 0.7448
```

```
Epoch 789/1000
v: 0.7448
Epoch 790/1000
y: 0.7448
Epoch 791/1000
v: 0.7448
Epoch 792/1000
v: 0.7448
Epoch 793/1000
v: 0.7448
Epoch 794/1000
v: 0.7448
Epoch 795/1000
v: 0.7448
Epoch 796/1000
v: 0.7448
Epoch 797/1000
v: 0.7448
Epoch 798/1000
v: 0.7448
Epoch 799/1000
v: 0.7448
Epoch 800/1000
v: 0.7448
Epoch 801/1000
v: 0.7448
Epoch 802/1000
v: 0.7448
Epoch 803/1000
y: 0.7448
Epoch 804/1000
```

```
v: 0.7448
Epoch 805/1000
v: 0.7448
Epoch 806/1000
v: 0.7448
Epoch 807/1000
v: 0.7448
Epoch 808/1000
cy: 0.7448
Epoch 809/1000
v: 0.7448
Epoch 810/1000
v: 0.7448
Epoch 811/1000
v: 0.7448
Epoch 812/1000
cy: 0.7448
Epoch 813/1000
cy: 0.7448
Epoch 814/1000
v: 0.7448
Epoch 815/1000
v: 0.7448
Epoch 816/1000
y: 0.7448
Epoch 817/1000
v: 0.7448
Epoch 818/1000
y: 0.7448
Epoch 819/1000
v: 0.7448
Epoch 820/1000
```

```
cv: 0.7448
Epoch 821/1000
cv: 0.7448
Epoch 822/1000
v: 0.7448
Epoch 823/1000
v: 0.7448
Epoch 824/1000
cy: 0.7448
Epoch 825/1000
cv: 0.7448
Epoch 826/1000
y: 0.7448
Epoch 827/1000
v: 0.7448
Epoch 828/1000
v: 0.7448
Epoch 829/1000
v: 0.7448
Epoch 830/1000
cy: 0.7448
Epoch 831/1000
cy: 0.7448
Epoch 832/1000
v: 0.7448
Epoch 833/1000
v: 0.7448
Epoch 834/1000
y: 0.7448
Epoch 835/1000
y: 0.7448
```

```
Epoch 836/1000
cv: 0.7448
Epoch 837/1000
y: 0.7448
Epoch 838/1000
v: 0.7448
Epoch 839/1000
v: 0.7448
Epoch 840/1000
cy: 0.7448
Epoch 841/1000
v: 0.7448
Epoch 842/1000
v: 0.7448
Epoch 843/1000
cy: 0.7448
Epoch 844/1000
v: 0.7448
Epoch 845/1000
v: 0.7448
Epoch 846/1000
v: 0.7448
Epoch 847/1000
v: 0.7448
Epoch 848/1000
cv: 0.7448
Epoch 849/1000
v: 0.7448
Epoch 850/1000
y: 0.7448
Epoch 851/1000
```

```
cy: 0.7448
Epoch 852/1000
v: 0.7448
Epoch 853/1000
v: 0.7448
Epoch 854/1000
cv: 0.7448
Epoch 855/1000
v: 0.7448
Epoch 856/1000
cy: 0.7448
Epoch 857/1000
cy: 0.7448
Epoch 858/1000
cy: 0.7448
Epoch 859/1000
v: 0.7448
Epoch 860/1000
v: 0.7448
Epoch 861/1000
v: 0.7448
Epoch 862/1000
cy: 0.7448
Epoch 863/1000
cv: 0.7448
Epoch 864/1000
v: 0.7448
Epoch 865/1000
v: 0.7448
Epoch 866/1000
cy: 0.7448
Epoch 867/1000
```

```
cv: 0.7448
Epoch 868/1000
v: 0.7448
Epoch 869/1000
v: 0.7448
Epoch 870/1000
cy: 0.7448
Epoch 871/1000
v: 0.7448
Epoch 872/1000
v: 0.7448
Epoch 873/1000
cy: 0.7448
Epoch 874/1000
v: 0.7448
Epoch 875/1000
v: 0.7448
Epoch 876/1000
cy: 0.7448
Epoch 877/1000
v: 0.7448
Epoch 878/1000
v: 0.7448
Epoch 879/1000
v: 0.7448
Epoch 880/1000
cy: 0.7448
Epoch 881/1000
y: 0.7448
Epoch 882/1000
y: 0.7448
```

```
Epoch 883/1000
v: 0.7448
Epoch 884/1000
y: 0.7448
Epoch 885/1000
v: 0.7448
Epoch 886/1000
cy: 0.7448
Epoch 887/1000
cy: 0.7448
Epoch 888/1000
v: 0.7448
Epoch 889/1000
cy: 0.7448
Epoch 890/1000
cy: 0.7448
Epoch 891/1000
cy: 0.7448
Epoch 892/1000
v: 0.7448
Epoch 893/1000
v: 0.7448
Epoch 894/1000
v: 0.7448
Epoch 895/1000
v: 0.7448
Epoch 896/1000
v: 0.7448
Epoch 897/1000
cy: 0.7448
Epoch 898/1000
```

```
v: 0.7448
Epoch 899/1000
v: 0.7448
Epoch 900/1000
v: 0.7448
Epoch 901/1000
v: 0.7448
Epoch 902/1000
v: 0.7448
Epoch 903/1000
v: 0.7448
Epoch 904/1000
v: 0.7500
Epoch 905/1000
v: 0.7500
Epoch 906/1000
v: 0.7500
Epoch 907/1000
v: 0.7500
Epoch 908/1000
v: 0.7500
Epoch 909/1000
v: 0.7500
Epoch 910/1000
v: 0.7500
Epoch 911/1000
v: 0.7500
Epoch 912/1000
v: 0.7500
Epoch 913/1000
v: 0.7500
Epoch 914/1000
```

```
cv: 0.7500
Epoch 915/1000
cv: 0.7500
Epoch 916/1000
cy: 0.7500
Epoch 917/1000
cy: 0.7500
Epoch 918/1000
cy: 0.7500
Epoch 919/1000
cv: 0.7500
Epoch 920/1000
v: 0.7500
Epoch 921/1000
cy: 0.7500
Epoch 922/1000
v: 0.7500
Epoch 923/1000
v: 0.7500
Epoch 924/1000
cy: 0.7500
Epoch 925/1000
v: 0.7500
Epoch 926/1000
v: 0.7500
Epoch 927/1000
v: 0.7500
Epoch 928/1000
y: 0.7500
Epoch 929/1000
y: 0.7500
```

```
Epoch 930/1000
v: 0.7500
Epoch 931/1000
y: 0.7500
Epoch 932/1000
v: 0.7500
Epoch 933/1000
v: 0.7500
Epoch 934/1000
v: 0.7500
Epoch 935/1000
cy: 0.7500
Epoch 936/1000
cy: 0.7500
Epoch 937/1000
cy: 0.7500
Epoch 938/1000
cy: 0.7500
Epoch 939/1000
cy: 0.7500
Epoch 940/1000
cy: 0.7500
Epoch 941/1000
v: 0.7500
Epoch 942/1000
v: 0.7500
Epoch 943/1000
cv: 0.7500
Epoch 944/1000
y: 0.7500
Epoch 945/1000
```

```
cy: 0.7500
Epoch 946/1000
v: 0.7500
Epoch 947/1000
v: 0.7500
Epoch 948/1000
v: 0.7500
Epoch 949/1000
v: 0.7500
Epoch 950/1000
cy: 0.7500
Epoch 951/1000
v: 0.7500
Epoch 952/1000
v: 0.7500
Epoch 953/1000
v: 0.7500
Epoch 954/1000
cy: 0.7500
Epoch 955/1000
v: 0.7500
Epoch 956/1000
v: 0.7500
Epoch 957/1000
y: 0.7500
Epoch 958/1000
cy: 0.7500
Epoch 959/1000
cy: 0.7500
Epoch 960/1000
cv: 0.7500
Epoch 961/1000
```

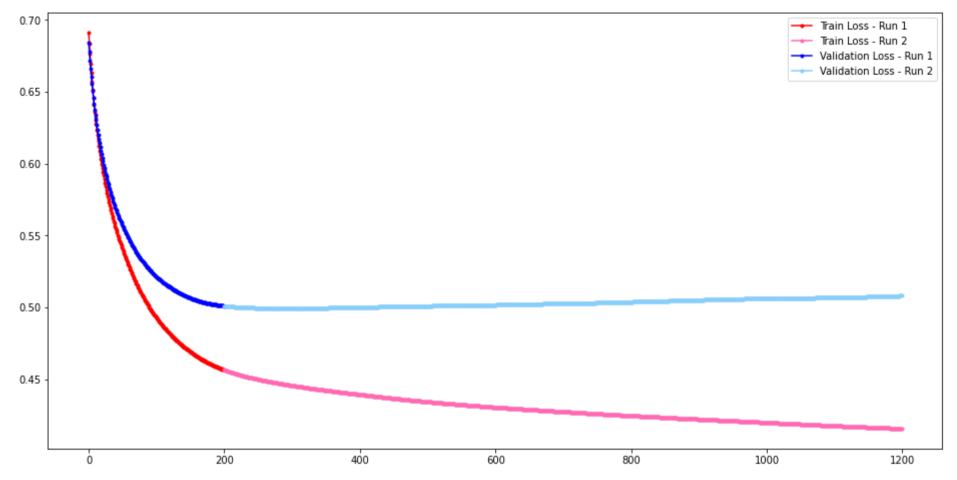
```
cv: 0.7500
Epoch 962/1000
cv: 0.7500
Epoch 963/1000
cy: 0.7500
Epoch 964/1000
cy: 0.7500
Epoch 965/1000
cy: 0.7500
Epoch 966/1000
cy: 0.7500
Epoch 967/1000
cy: 0.7500
Epoch 968/1000
cy: 0.7500
Epoch 969/1000
cy: 0.7500
Epoch 970/1000
v: 0.7500
Epoch 971/1000
v: 0.7500
Epoch 972/1000
v: 0.7500
Epoch 973/1000
v: 0.7500
Epoch 974/1000
v: 0.7500
Epoch 975/1000
cy: 0.7500
Epoch 976/1000
y: 0.7500
```

```
Epoch 977/1000
v: 0.7500
Epoch 978/1000
cy: 0.7500
Epoch 979/1000
cy: 0.7500
Epoch 980/1000
y: 0.7500
Epoch 981/1000
cy: 0.7500
Epoch 982/1000
y: 0.7500
Epoch 983/1000
v: 0.7500
Epoch 984/1000
cy: 0.7500
Epoch 985/1000
cy: 0.7500
Epoch 986/1000
v: 0.7500
Epoch 987/1000
v: 0.7500
Epoch 988/1000
cy: 0.7500
Epoch 989/1000
v: 0.7500
Epoch 990/1000
v: 0.7500
Epoch 991/1000
y: 0.7500
Epoch 992/1000
```

```
cv: 0.7500
    Epoch 993/1000
    v: 0.7500
    Epoch 994/1000
    v: 0.7552
    Epoch 995/1000
    v: 0.7552
    Epoch 996/1000
    v: 0.7552
    Epoch 997/1000
    v: 0.7552
    Epoch 998/1000
    v: 0.7552
    Epoch 999/1000
    v: 0.7552
    Epoch 1000/1000
    v: 0.7552
In [23]:
    n = len(run hist 1.history["loss"])
    m = len(run hist 1b.history['loss'])
    fig, ax = plt.subplots(figsize=(16, 8))
    ax.plot(range(n), run hist 1.history["loss"],'r', marker='.', label="Train Loss - Run 1")
    ax.plot(range(n, n+m), run hist 1b.history["loss"], 'hotpink', marker='.', label="Train Loss - Run 2")
    ax.plot(range(n), run hist 1.history["val loss"],'b', marker='.', label="Validation Loss - Run 1")
    ax.plot(range(n, n+m), run hist 1b.history["val loss"], 'LightSkyBlue', marker='.', label="Validation Loss - Run 2")
    ax.legend()
```

Out[23]: <matplotlib.legend.Legend at 0x7f9318ef9940>

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Note that this graph begins where the other left off. While the training loss is still going down, it looks like the validation loss has stabilized (or even gotten worse!). This suggests that our network will not benefit from further training. What is the appropriate number of epochs?

Training loss indica quão bom o modelo está se ajustando aos dados de treino, enquanto validation loss indica quão bem o modelo se ajusta a novos dados. Pelo gráfico acima é possível notar que para run1, que o modelo se ajusta dados de treino mas não se ajusta a novos dados. Enquanto pra run2 o modelo se ajustou de forma mais eficiente em relação à run1 a dados de treino e dados novos. O número de épocas apropriado para run seria em torno de 300 e para run2 train loss ainda está a cair em 1200 épocas e seria possível aumentar para observar uma curva a tender a uma constante..

```
In [25]: print(10)

10
```

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Exercise

Now it's your turn. Do the following in the cells below:

- Build a model with two hidden layers, each with 6 nodes
- Use the "relu" activation function for the hidden layers, and "sigmoid" for the final layer
- Use a learning rate of .003 and train for 1500 epochs
- Graph the trajectory of the loss functions, accuracy on both train and test set
- Plot the roc curve for the predictions

Experiment with different learning rates, numbers of epochs, and network structures

```
In [26]:
     # Input size is 8-dimensional
     # 1 hidden layer, 6 hidden nodes, relu activation
     # Final layer has just one node with a sigmoid activation (standard for binary classification)
     model 2 = Sequential()
     model 2.add(Dense(6, input shape=(8,), activation="relu"))
     model 2.add(Dense(6, activation="relu"))
     model 2.add(Dense(1, activation="sigmoid"))
     model 2.compile(SGD(lr = .003), "binary crossentropy", metrics=["accuracy"])
     run hist 2 = model 2.fit(X train norm, y train, validation data=(X test norm, y test), epochs=1500)
     Epoch 1/1500
    cy: 0.5365
     Epoch 2/1500
    v: 0.5521
     Epoch 3/1500
    v: 0.5625
     Epoch 4/1500
    y: 0.5677
     Epoch 5/1500
    cy: 0.5781
     Epoch 6/1500
```

```
cv: 0.5729
Epoch 7/1500
v: 0.5938
Epoch 8/1500
cy: 0.5990
Epoch 9/1500
cy: 0.6146
Epoch 10/1500
cy: 0.6146
Epoch 11/1500
cy: 0.6146
Epoch 12/1500
cy: 0.6198
Epoch 13/1500
cy: 0.6198
Epoch 14/1500
cy: 0.6302
Epoch 15/1500
v: 0.6354
Epoch 16/1500
cy: 0.6354
Epoch 17/1500
cy: 0.6302
Epoch 18/1500
cy: 0.6250
Epoch 19/1500
cy: 0.6250
Epoch 20/1500
cy: 0.6198
Epoch 21/1500
cy: 0.6302
```

```
Epoch 22/1500
cv: 0.6354
Epoch 23/1500
cy: 0.6406
Epoch 24/1500
cy: 0.6458
Epoch 25/1500
cy: 0.6458
Epoch 26/1500
cv: 0.6458
Epoch 27/1500
cy: 0.6458
Epoch 28/1500
cv: 0.6458
Epoch 29/1500
cy: 0.6458
Epoch 30/1500
cv: 0.6458
Epoch 31/1500
cy: 0.6458
Epoch 32/1500
v: 0.6458
Epoch 33/1500
v: 0.6458
Epoch 34/1500
v: 0.6458
Epoch 35/1500
v: 0.6458
Epoch 36/1500
cy: 0.6406
Epoch 37/1500
```

```
v: 0.6406
Epoch 38/1500
cv: 0.6406
Epoch 39/1500
v: 0.6406
Epoch 40/1500
cv: 0.6354
Epoch 41/1500
v: 0.6354
Epoch 42/1500
v: 0.6354
Epoch 43/1500
v: 0.6354
Epoch 44/1500
v: 0.6354
Epoch 45/1500
cy: 0.6354
Epoch 46/1500
v: 0.6354
Epoch 47/1500
v: 0.6354
Epoch 48/1500
v: 0.6354
Epoch 49/1500
cy: 0.6354
Epoch 50/1500
v: 0.6354
Epoch 51/1500
y: 0.6354
Epoch 52/1500
v: 0.6406
Epoch 53/1500
```

```
cv: 0.6406
Epoch 54/1500
v: 0.6406
Epoch 55/1500
v: 0.6406
Epoch 56/1500
cy: 0.6406
Epoch 57/1500
v: 0.6406
Epoch 58/1500
v: 0.6406
Epoch 59/1500
cy: 0.6406
Epoch 60/1500
v: 0.6406
Epoch 61/1500
v: 0.6406
Epoch 62/1500
v: 0.6406
Epoch 63/1500
cy: 0.6406
Epoch 64/1500
v: 0.6406
Epoch 65/1500
v: 0.6406
Epoch 66/1500
cy: 0.6406
Epoch 67/1500
y: 0.6406
Epoch 68/1500
y: 0.6406
```

```
Epoch 69/1500
cv: 0.6406
Epoch 70/1500
y: 0.6406
Epoch 71/1500
v: 0.6406
Epoch 72/1500
cy: 0.6406
Epoch 73/1500
v: 0.6406
Epoch 74/1500
v: 0.6406
Epoch 75/1500
cy: 0.6406
Epoch 76/1500
y: 0.6406
Epoch 77/1500
v: 0.6406
Epoch 78/1500
cy: 0.6406
Epoch 79/1500
v: 0.6406
Epoch 80/1500
v: 0.6406
Epoch 81/1500
v: 0.6406
Epoch 82/1500
cy: 0.6406
Epoch 83/1500
y: 0.6406
Epoch 84/1500
```

```
v: 0.6406
Epoch 85/1500
v: 0.6406
Epoch 86/1500
cv: 0.6406
Epoch 87/1500
v: 0.6406
Epoch 88/1500
y: 0.6406
Epoch 89/1500
cy: 0.6406
Epoch 90/1500
v: 0.6406
Epoch 91/1500
v: 0.6406
Epoch 92/1500
cy: 0.6406
Epoch 93/1500
v: 0.6406
Epoch 94/1500
v: 0.6406
Epoch 95/1500
cy: 0.6406
Epoch 96/1500
v: 0.6406
Epoch 97/1500
y: 0.6406
Epoch 98/1500
cy: 0.6406
Epoch 99/1500
v: 0.6406
Epoch 100/1500
```

```
v: 0.6406
Epoch 101/1500
cv: 0.6406
Epoch 102/1500
v: 0.6406
Epoch 103/1500
v: 0.6406
Epoch 104/1500
v: 0.6406
Epoch 105/1500
v: 0.6406
Epoch 106/1500
cy: 0.6406
Epoch 107/1500
v: 0.6406
Epoch 108/1500
v: 0.6406
Epoch 109/1500
cy: 0.6406
Epoch 110/1500
v: 0.6406
Epoch 111/1500
cy: 0.6458
Epoch 112/1500
v: 0.6458
Epoch 113/1500
v: 0.6458
Epoch 114/1500
cy: 0.6458
Epoch 115/1500
y: 0.6458
```

```
Epoch 116/1500
v: 0.6458
Epoch 117/1500
y: 0.6458
Epoch 118/1500
v: 0.6458
Epoch 119/1500
v: 0.6458
Epoch 120/1500
v: 0.6458
Epoch 121/1500
cy: 0.6458
Epoch 122/1500
cy: 0.6458
Epoch 123/1500
cv: 0.6458
Epoch 124/1500
cy: 0.6458
Epoch 125/1500
v: 0.6458
Epoch 126/1500
v: 0.6458
Epoch 127/1500
v: 0.6458
Epoch 128/1500
v: 0.6458
Epoch 129/1500
cv: 0.6458
Epoch 130/1500
y: 0.6458
Epoch 131/1500
```

```
y: 0.6458
Epoch 132/1500
cv: 0.6458
Epoch 133/1500
v: 0.6510
Epoch 134/1500
v: 0.6510
Epoch 135/1500
cy: 0.6510
Epoch 136/1500
cy: 0.6510
Epoch 137/1500
cy: 0.6510
Epoch 138/1500
cy: 0.6510
Epoch 139/1500
cy: 0.6510
Epoch 140/1500
cy: 0.6510
Epoch 141/1500
cy: 0.6510
Epoch 142/1500
v: 0.6562
Epoch 143/1500
y: 0.6562
Epoch 144/1500
v: 0.6562
Epoch 145/1500
y: 0.6562
Epoch 146/1500
v: 0.6615
Epoch 147/1500
```

```
v: 0.6615
Epoch 148/1500
v: 0.6615
Epoch 149/1500
v: 0.6615
Epoch 150/1500
v: 0.6615
Epoch 151/1500
v: 0.6615
Epoch 152/1500
v: 0.6615
Epoch 153/1500
cy: 0.6615
Epoch 154/1500
v: 0.6615
Epoch 155/1500
v: 0.6615
Epoch 156/1500
v: 0.6615
Epoch 157/1500
v: 0.6615
Epoch 158/1500
v: 0.6615
Epoch 159/1500
v: 0.6615
Epoch 160/1500
v: 0.6615
Epoch 161/1500
cy: 0.6615
Epoch 162/1500
y: 0.6615
```

```
Epoch 163/1500
cv: 0.6615
Epoch 164/1500
y: 0.6615
Epoch 165/1500
v: 0.6615
Epoch 166/1500
v: 0.6615
Epoch 167/1500
cv: 0.6615
Epoch 168/1500
v: 0.6615
Epoch 169/1500
v: 0.6615
Epoch 170/1500
y: 0.6615
Epoch 171/1500
v: 0.6615
Epoch 172/1500
v: 0.6615
Epoch 173/1500
v: 0.6615
Epoch 174/1500
v: 0.6615
Epoch 175/1500
v: 0.6615
Epoch 176/1500
v: 0.6615
Epoch 177/1500
y: 0.6615
Epoch 178/1500
```

```
v: 0.6615
Epoch 179/1500
v: 0.6615
Epoch 180/1500
v: 0.6667
Epoch 181/1500
v: 0.6667
Epoch 182/1500
y: 0.6667
Epoch 183/1500
v: 0.6667
Epoch 184/1500
cy: 0.6719
Epoch 185/1500
cy: 0.6719
Epoch 186/1500
cy: 0.6719
Epoch 187/1500
cy: 0.6719
Epoch 188/1500
cy: 0.6719
Epoch 189/1500
cy: 0.6719
Epoch 190/1500
y: 0.6771
Epoch 191/1500
v: 0.6823
Epoch 192/1500
y: 0.6823
Epoch 193/1500
v: 0.6823
Epoch 194/1500
```

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```
cv: 0.6823
Epoch 195/1500
cv: 0.6823
Epoch 196/1500
v: 0.6875
Epoch 197/1500
v: 0.6875
Epoch 198/1500
v: 0.6875
Epoch 199/1500
v: 0.6875
Epoch 200/1500
y: 0.6875
Epoch 201/1500
cy: 0.6875
Epoch 202/1500
cy: 0.6823
Epoch 203/1500
v: 0.6823
Epoch 204/1500
v: 0.6823
Epoch 205/1500
v: 0.6823
Epoch 206/1500
cy: 0.6823
Epoch 207/1500
cy: 0.6823
Epoch 208/1500
y: 0.6823
Epoch 209/1500
y: 0.6823
```

```
Epoch 210/1500
v: 0.6823
Epoch 211/1500
y: 0.6823
Epoch 212/1500
v: 0.6823
Epoch 213/1500
cy: 0.6823
Epoch 214/1500
cv: 0.6823
Epoch 215/1500
v: 0.6823
Epoch 216/1500
v: 0.6875
Epoch 217/1500
v: 0.6875
Epoch 218/1500
cv: 0.6927
Epoch 219/1500
v: 0.6927
Epoch 220/1500
v: 0.6979
Epoch 221/1500
cy: 0.6927
Epoch 222/1500
cv: 0.6927
Epoch 223/1500
v: 0.6927
Epoch 224/1500
v: 0.6927
Epoch 225/1500
```

```
cv: 0.6927
Epoch 226/1500
v: 0.6927
Epoch 227/1500
v: 0.6979
Epoch 228/1500
v: 0.6927
Epoch 229/1500
y: 0.6927
Epoch 230/1500
cy: 0.6927
Epoch 231/1500
v: 0.6927
Epoch 232/1500
v: 0.6979
Epoch 233/1500
v: 0.6979
Epoch 234/1500
v: 0.7031
Epoch 235/1500
cy: 0.7083
Epoch 236/1500
cy: 0.7083
Epoch 237/1500
y: 0.7083
Epoch 238/1500
y: 0.7083
Epoch 239/1500
y: 0.7135
Epoch 240/1500
v: 0.7135
Epoch 241/1500
```

```
cv: 0.7135
Epoch 242/1500
cv: 0.7135
Epoch 243/1500
cy: 0.7135
Epoch 244/1500
v: 0.7188
Epoch 245/1500
v: 0.7188
Epoch 246/1500
v: 0.7188
Epoch 247/1500
y: 0.7188
Epoch 248/1500
v: 0.7188
Epoch 249/1500
v: 0.7188
Epoch 250/1500
v: 0.7188
Epoch 251/1500
v: 0.7240
Epoch 252/1500
cy: 0.7240
Epoch 253/1500
cy: 0.7188
Epoch 254/1500
cy: 0.7188
Epoch 255/1500
y: 0.7188
Epoch 256/1500
cy: 0.7188
```

```
Epoch 257/1500
cv: 0.7188
Epoch 258/1500
cy: 0.7188
Epoch 259/1500
cy: 0.7188
Epoch 260/1500
v: 0.7135
Epoch 261/1500
v: 0.7135
Epoch 262/1500
v: 0.7083
Epoch 263/1500
v: 0.7135
Epoch 264/1500
v: 0.7135
Epoch 265/1500
cy: 0.7135
Epoch 266/1500
v: 0.7083
Epoch 267/1500
v: 0.7135
Epoch 268/1500
v: 0.7135
Epoch 269/1500
cv: 0.7135
Epoch 270/1500
cv: 0.7135
Epoch 271/1500
y: 0.7135
Epoch 272/1500
```

```
v: 0.7135
Epoch 273/1500
v: 0.7135
Epoch 274/1500
v: 0.7135
Epoch 275/1500
v: 0.7135
Epoch 276/1500
v: 0.7135
Epoch 277/1500
v: 0.7135
Epoch 278/1500
v: 0.7135
Epoch 279/1500
v: 0.7188
Epoch 280/1500
v: 0.7188
Epoch 281/1500
cy: 0.7240
Epoch 282/1500
v: 0.7240
Epoch 283/1500
v: 0.7240
Epoch 284/1500
y: 0.7240
Epoch 285/1500
v: 0.7240
Epoch 286/1500
y: 0.7292
Epoch 287/1500
cv: 0.7292
Epoch 288/1500
```

```
cv: 0.7344
Epoch 289/1500
cv: 0.7344
Epoch 290/1500
v: 0.7396
Epoch 291/1500
v: 0.7292
Epoch 292/1500
cy: 0.7292
Epoch 293/1500
cy: 0.7292
Epoch 294/1500
cy: 0.7240
Epoch 295/1500
v: 0.7240
Epoch 296/1500
v: 0.7240
Epoch 297/1500
v: 0.7240
Epoch 298/1500
v: 0.7240
Epoch 299/1500
v: 0.7240
Epoch 300/1500
v: 0.7292
Epoch 301/1500
v: 0.7344
Epoch 302/1500
y: 0.7396
Epoch 303/1500
y: 0.7448
```

```
Epoch 304/1500
v: 0.7448
Epoch 305/1500
y: 0.7448
Epoch 306/1500
v: 0.7500
Epoch 307/1500
cy: 0.7500
Epoch 308/1500
cy: 0.7500
Epoch 309/1500
cy: 0.7500
Epoch 310/1500
v: 0.7500
Epoch 311/1500
cy: 0.7500
Epoch 312/1500
v: 0.7552
Epoch 313/1500
v: 0.7552
Epoch 314/1500
v: 0.7552
Epoch 315/1500
v: 0.7552
Epoch 316/1500
v: 0.7552
Epoch 317/1500
v: 0.7552
Epoch 318/1500
y: 0.7604
Epoch 319/1500
```

```
v: 0.7604
Epoch 320/1500
v: 0.7604
Epoch 321/1500
v: 0.7604
Epoch 322/1500
v: 0.7604
Epoch 323/1500
v: 0.7604
Epoch 324/1500
v: 0.7552
Epoch 325/1500
v: 0.7552
Epoch 326/1500
v: 0.7552
Epoch 327/1500
v: 0.7552
Epoch 328/1500
v: 0.7552
Epoch 329/1500
v: 0.7552
Epoch 330/1500
v: 0.7552
Epoch 331/1500
v: 0.7552
Epoch 332/1500
v: 0.7552
Epoch 333/1500
y: 0.7552
Epoch 334/1500
cy: 0.7604
Epoch 335/1500
```

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```
cv: 0.7552
Epoch 336/1500
cv: 0.7500
Epoch 337/1500
v: 0.7552
Epoch 338/1500
cy: 0.7552
Epoch 339/1500
v: 0.7552
Epoch 340/1500
v: 0.7552
Epoch 341/1500
y: 0.7552
Epoch 342/1500
v: 0.7552
Epoch 343/1500
v: 0.7552
Epoch 344/1500
v: 0.7552
Epoch 345/1500
v: 0.7552
Epoch 346/1500
v: 0.7552
Epoch 347/1500
v: 0.7552
Epoch 348/1500
v: 0.7552
Epoch 349/1500
cy: 0.7552
Epoch 350/1500
cy: 0.7552
```

```
Epoch 351/1500
v: 0.7552
Epoch 352/1500
y: 0.7552
Epoch 353/1500
v: 0.7552
Epoch 354/1500
v: 0.7552
Epoch 355/1500
v: 0.7552
Epoch 356/1500
cy: 0.7552
Epoch 357/1500
v: 0.7552
Epoch 358/1500
v: 0.7552
Epoch 359/1500
v: 0.7552
Epoch 360/1500
v: 0.7552
Epoch 361/1500
v: 0.7552
Epoch 362/1500
v: 0.7552
Epoch 363/1500
v: 0.7448
Epoch 364/1500
v: 0.7396
Epoch 365/1500
y: 0.7396
Epoch 366/1500
```

```
y: 0.7396
Epoch 367/1500
v: 0.7396
Epoch 368/1500
v: 0.7396
Epoch 369/1500
v: 0.7396
Epoch 370/1500
y: 0.7396
Epoch 371/1500
cy: 0.7344
Epoch 372/1500
v: 0.7344
Epoch 373/1500
v: 0.7344
Epoch 374/1500
cy: 0.7344
Epoch 375/1500
cy: 0.7344
Epoch 376/1500
v: 0.7344
Epoch 377/1500
v: 0.7344
Epoch 378/1500
cy: 0.7344
Epoch 379/1500
cy: 0.7344
Epoch 380/1500
cy: 0.7344
Epoch 381/1500
cv: 0.7344
Epoch 382/1500
```

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```
cv: 0.7344
Epoch 383/1500
cv: 0.7344
Epoch 384/1500
v: 0.7344
Epoch 385/1500
v: 0.7344
Epoch 386/1500
v: 0.7344
Epoch 387/1500
cy: 0.7344
Epoch 388/1500
y: 0.7344
Epoch 389/1500
v: 0.7344
Epoch 390/1500
v: 0.7344
Epoch 391/1500
v: 0.7344
Epoch 392/1500
v: 0.7344
Epoch 393/1500
v: 0.7344
Epoch 394/1500
cy: 0.7344
Epoch 395/1500
cy: 0.7344
Epoch 396/1500
cy: 0.7344
Epoch 397/1500
cy: 0.7292
```

```
Epoch 398/1500
v: 0.7292
Epoch 399/1500
y: 0.7292
Epoch 400/1500
v: 0.7292
Epoch 401/1500
v: 0.7292
Epoch 402/1500
cy: 0.7292
Epoch 403/1500
v: 0.7292
Epoch 404/1500
v: 0.7292
Epoch 405/1500
y: 0.7292
Epoch 406/1500
v: 0.7292
Epoch 407/1500
v: 0.7344
Epoch 408/1500
v: 0.7344
Epoch 409/1500
cy: 0.7344
Epoch 410/1500
cy: 0.7344
Epoch 411/1500
v: 0.7396
Epoch 412/1500
y: 0.7396
Epoch 413/1500
```

```
v: 0.7396
Epoch 414/1500
v: 0.7396
Epoch 415/1500
v: 0.7396
Epoch 416/1500
v: 0.7396
Epoch 417/1500
y: 0.7396
Epoch 418/1500
v: 0.7396
Epoch 419/1500
v: 0.7396
Epoch 420/1500
v: 0.7396
Epoch 421/1500
v: 0.7396
Epoch 422/1500
cy: 0.7396
Epoch 423/1500
v: 0.7344
Epoch 424/1500
v: 0.7344
Epoch 425/1500
cy: 0.7344
Epoch 426/1500
y: 0.7344
Epoch 427/1500
y: 0.7344
Epoch 428/1500
v: 0.7344
Epoch 429/1500
```

```
v: 0.7344
Epoch 430/1500
v: 0.7344
Epoch 431/1500
cy: 0.7344
Epoch 432/1500
v: 0.7344
Epoch 433/1500
v: 0.7344
Epoch 434/1500
v: 0.7292
Epoch 435/1500
y: 0.7292
Epoch 436/1500
v: 0.7292
Epoch 437/1500
v: 0.7292
Epoch 438/1500
v: 0.7292
Epoch 439/1500
v: 0.7292
Epoch 440/1500
v: 0.7292
Epoch 441/1500
v: 0.7292
Epoch 442/1500
v: 0.7292
Epoch 443/1500
y: 0.7292
Epoch 444/1500
y: 0.7292
```

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Epoch 445/1500
v: 0.7292
Epoch 446/1500
y: 0.7292
Epoch 447/1500
v: 0.7292
Epoch 448/1500
v: 0.7292
Epoch 449/1500
v: 0.7292
Epoch 450/1500
v: 0.7344
Epoch 451/1500
v: 0.7344
Epoch 452/1500
v: 0.7292
Epoch 453/1500
v: 0.7292
Epoch 454/1500
v: 0.7344
Epoch 455/1500
cy: 0.7344
Epoch 456/1500
v: 0.7344
Epoch 457/1500
v: 0.7344
Epoch 458/1500
v: 0.7344
Epoch 459/1500
y: 0.7344
Epoch 460/1500
```

```
v: 0.7396
Epoch 461/1500
v: 0.7396
Epoch 462/1500
v: 0.7396
Epoch 463/1500
v: 0.7396
Epoch 464/1500
v: 0.7448
Epoch 465/1500
v: 0.7448
Epoch 466/1500
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Epoch 467/1500
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Epoch 468/1500
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Epoch 469/1500
v: 0.7448
Epoch 470/1500
v: 0.7448
Epoch 471/1500
v: 0.7448
Epoch 472/1500
v: 0.7448
Epoch 473/1500
v: 0.7448
Epoch 474/1500
y: 0.7448
Epoch 475/1500
v: 0.7448
Epoch 476/1500
```

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v: 0.7448
Epoch 477/1500
v: 0.7448
Epoch 478/1500
v: 0.7448
Epoch 479/1500
v: 0.7448
Epoch 480/1500
v: 0.7448
Epoch 481/1500
v: 0.7448
Epoch 482/1500
v: 0.7448
Epoch 483/1500
cy: 0.7448
Epoch 484/1500
cy: 0.7448
Epoch 485/1500
v: 0.7448
Epoch 486/1500
v: 0.7396
Epoch 487/1500
v: 0.7396
Epoch 488/1500
cy: 0.7396
Epoch 489/1500
v: 0.7396
Epoch 490/1500
y: 0.7396
Epoch 491/1500
y: 0.7396
```

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Epoch 492/1500
v: 0.7396
Epoch 493/1500
y: 0.7396
Epoch 494/1500
cy: 0.7396
Epoch 495/1500
cy: 0.7396
Epoch 496/1500
v: 0.7396
Epoch 497/1500
v: 0.7396
Epoch 498/1500
v: 0.7396
Epoch 499/1500
cy: 0.7396
Epoch 500/1500
v: 0.7396
Epoch 501/1500
v: 0.7396
Epoch 502/1500
cy: 0.7396
Epoch 503/1500
v: 0.7396
Epoch 504/1500
v: 0.7396
Epoch 505/1500
v: 0.7396
Epoch 506/1500
y: 0.7396
Epoch 507/1500
```

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v: 0.7396
Epoch 508/1500
cv: 0.7396
Epoch 509/1500
v: 0.7396
Epoch 510/1500
cv: 0.7396
Epoch 511/1500
y: 0.7396
Epoch 512/1500
v: 0.7396
Epoch 513/1500
v: 0.7396
Epoch 514/1500
v: 0.7396
Epoch 515/1500
v: 0.7396
Epoch 516/1500
v: 0.7396
Epoch 517/1500
v: 0.7344
Epoch 518/1500
v: 0.7344
Epoch 519/1500
v: 0.7344
Epoch 520/1500
v: 0.7344
Epoch 521/1500
y: 0.7344
Epoch 522/1500
v: 0.7344
Epoch 523/1500
```

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v: 0.7344
Epoch 524/1500
v: 0.7344
Epoch 525/1500
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Epoch 526/1500
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Epoch 527/1500
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Epoch 528/1500
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Epoch 529/1500
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Epoch 530/1500
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Epoch 531/1500
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Epoch 532/1500
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Epoch 533/1500
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Epoch 534/1500
v: 0.7292
Epoch 535/1500
v: 0.7292
Epoch 536/1500
v: 0.7292
Epoch 537/1500
cy: 0.7292
Epoch 538/1500
y: 0.7292
```

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Epoch 539/1500
v: 0.7292
Epoch 540/1500
y: 0.7292
Epoch 541/1500
v: 0.7292
Epoch 542/1500
v: 0.7292
Epoch 543/1500
v: 0.7344
Epoch 544/1500
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Epoch 547/1500
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Epoch 550/1500
v: 0.7344
Epoch 551/1500
v: 0.7344
Epoch 552/1500
v: 0.7344
Epoch 553/1500
y: 0.7344
Epoch 554/1500
```

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v: 0.7344
Epoch 555/1500
v: 0.7344
Epoch 556/1500
v: 0.7344
Epoch 557/1500
v: 0.7292
Epoch 558/1500
y: 0.7292
Epoch 559/1500
v: 0.7292
Epoch 560/1500
v: 0.7292
Epoch 561/1500
v: 0.7240
Epoch 562/1500
v: 0.7240
Epoch 563/1500
v: 0.7240
Epoch 564/1500
v: 0.7240
Epoch 565/1500
cy: 0.7240
Epoch 566/1500
y: 0.7240
Epoch 567/1500
y: 0.7240
Epoch 568/1500
y: 0.7240
Epoch 569/1500
v: 0.7240
Epoch 570/1500
```

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v: 0.7240
Epoch 571/1500
v: 0.7240
Epoch 572/1500
v: 0.7240
Epoch 573/1500
v: 0.7240
Epoch 574/1500
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Epoch 575/1500
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Epoch 576/1500
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Epoch 577/1500
v: 0.7240
Epoch 578/1500
v: 0.7240
Epoch 579/1500
v: 0.7240
Epoch 580/1500
v: 0.7240
Epoch 581/1500
v: 0.7240
Epoch 582/1500
v: 0.7240
Epoch 583/1500
v: 0.7240
Epoch 584/1500
y: 0.7240
Epoch 585/1500
y: 0.7240
```

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Epoch 586/1500
v: 0.7240
Epoch 587/1500
y: 0.7240
Epoch 588/1500
v: 0.7240
Epoch 589/1500
v: 0.7240
Epoch 590/1500
v: 0.7240
Epoch 591/1500
v: 0.7240
Epoch 592/1500
cy: 0.7240
Epoch 593/1500
v: 0.7240
Epoch 594/1500
v: 0.7240
Epoch 595/1500
v: 0.7240
Epoch 596/1500
v: 0.7240
Epoch 597/1500
v: 0.7240
Epoch 598/1500
v: 0.7240
Epoch 599/1500
v: 0.7240
Epoch 600/1500
y: 0.7240
Epoch 601/1500
```

```
v: 0.7240
Epoch 602/1500
v: 0.7240
Epoch 603/1500
v: 0.7240
Epoch 604/1500
v: 0.7240
Epoch 605/1500
v: 0.7240
Epoch 606/1500
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Epoch 607/1500
v: 0.7292
Epoch 608/1500
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Epoch 609/1500
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Epoch 610/1500
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Epoch 611/1500
v: 0.7292
Epoch 612/1500
v: 0.7292
Epoch 613/1500
v: 0.7292
Epoch 614/1500
v: 0.7292
Epoch 615/1500
v: 0.7292
Epoch 616/1500
v: 0.7292
Epoch 617/1500
```

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v: 0.7292
Epoch 618/1500
v: 0.7240
Epoch 619/1500
v: 0.7240
Epoch 620/1500
cy: 0.7240
Epoch 621/1500
v: 0.7188
Epoch 622/1500
v: 0.7188
Epoch 623/1500
y: 0.7188
Epoch 624/1500
v: 0.7188
Epoch 625/1500
v: 0.7188
Epoch 626/1500
v: 0.7188
Epoch 627/1500
v: 0.7188
Epoch 628/1500
v: 0.7188
Epoch 629/1500
v: 0.7188
Epoch 630/1500
v: 0.7188
Epoch 631/1500
y: 0.7188
Epoch 632/1500
y: 0.7188
```

```
Epoch 633/1500
v: 0.7188
Epoch 634/1500
y: 0.7188
Epoch 635/1500
v: 0.7188
Epoch 636/1500
v: 0.7188
Epoch 637/1500
v: 0.7188
Epoch 638/1500
v: 0.7188
Epoch 639/1500
v: 0.7188
Epoch 640/1500
y: 0.7188
Epoch 641/1500
v: 0.7188
Epoch 642/1500
v: 0.7188
Epoch 643/1500
v: 0.7188
Epoch 644/1500
v: 0.7188
Epoch 645/1500
v: 0.7188
Epoch 646/1500
v: 0.7188
Epoch 647/1500
cy: 0.7188
Epoch 648/1500
```

```
cy: 0.7188
Epoch 649/1500
v: 0.7188
Epoch 650/1500
v: 0.7188
Epoch 651/1500
v: 0.7188
Epoch 652/1500
v: 0.7188
Epoch 653/1500
v: 0.7188
Epoch 654/1500
v: 0.7188
Epoch 655/1500
v: 0.7188
Epoch 656/1500
v: 0.7188
Epoch 657/1500
v: 0.7188
Epoch 658/1500
v: 0.7188
Epoch 659/1500
v: 0.7188
Epoch 660/1500
v: 0.7188
Epoch 661/1500
v: 0.7188
Epoch 662/1500
y: 0.7188
Epoch 663/1500
v: 0.7188
Epoch 664/1500
```

```
v: 0.7188
Epoch 665/1500
v: 0.7188
Epoch 666/1500
v: 0.7188
Epoch 667/1500
v: 0.7188
Epoch 668/1500
v: 0.7188
Epoch 669/1500
v: 0.7188
Epoch 670/1500
v: 0.7188
Epoch 671/1500
v: 0.7188
Epoch 672/1500
v: 0.7188
Epoch 673/1500
cy: 0.7188
Epoch 674/1500
v: 0.7188
Epoch 675/1500
v: 0.7188
Epoch 676/1500
v: 0.7188
Epoch 677/1500
v: 0.7188
Epoch 678/1500
y: 0.7188
Epoch 679/1500
y: 0.7188
```

```
Epoch 680/1500
v: 0.7188
Epoch 681/1500
y: 0.7188
Epoch 682/1500
cy: 0.7188
Epoch 683/1500
cy: 0.7188
Epoch 684/1500
v: 0.7188
Epoch 685/1500
v: 0.7188
Epoch 686/1500
v: 0.7188
Epoch 687/1500
y: 0.7188
Epoch 688/1500
v: 0.7188
Epoch 689/1500
v: 0.7188
Epoch 690/1500
v: 0.7188
Epoch 691/1500
v: 0.7188
Epoch 692/1500
v: 0.7188
Epoch 693/1500
v: 0.7188
Epoch 694/1500
y: 0.7188
Epoch 695/1500
```

```
y: 0.7188
Epoch 696/1500
v: 0.7188
Epoch 697/1500
v: 0.7188
Epoch 698/1500
cy: 0.7188
Epoch 699/1500
v: 0.7188
Epoch 700/1500
v: 0.7188
Epoch 701/1500
v: 0.7188
Epoch 702/1500
v: 0.7188
Epoch 703/1500
v: 0.7188
Epoch 704/1500
v: 0.7188
Epoch 705/1500
cy: 0.7188
Epoch 706/1500
v: 0.7188
Epoch 707/1500
y: 0.7188
Epoch 708/1500
v: 0.7188
Epoch 709/1500
y: 0.7188
Epoch 710/1500
v: 0.7188
Epoch 711/1500
```

```
v: 0.7188
Epoch 712/1500
v: 0.7188
Epoch 713/1500
v: 0.7188
Epoch 714/1500
v: 0.7188
Epoch 715/1500
v: 0.7188
Epoch 716/1500
v: 0.7188
Epoch 717/1500
y: 0.7188
Epoch 718/1500
v: 0.7188
Epoch 719/1500
v: 0.7188
Epoch 720/1500
v: 0.7188
Epoch 721/1500
v: 0.7188
Epoch 722/1500
v: 0.7188
Epoch 723/1500
v: 0.7188
Epoch 724/1500
v: 0.7188
Epoch 725/1500
cy: 0.7188
Epoch 726/1500
y: 0.7188
```

```
Epoch 727/1500
v: 0.7188
Epoch 728/1500
y: 0.7135
Epoch 729/1500
v: 0.7135
Epoch 730/1500
v: 0.7135
Epoch 731/1500
v: 0.7135
Epoch 732/1500
v: 0.7135
Epoch 733/1500
v: 0.7135
Epoch 734/1500
v: 0.7135
Epoch 735/1500
v: 0.7135
Epoch 736/1500
v: 0.7135
Epoch 737/1500
v: 0.7135
Epoch 738/1500
v: 0.7188
Epoch 739/1500
v: 0.7188
Epoch 740/1500
v: 0.7240
Epoch 741/1500
v: 0.7240
Epoch 742/1500
```

```
v: 0.7240
Epoch 743/1500
v: 0.7240
Epoch 744/1500
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Epoch 745/1500
v: 0.7240
Epoch 746/1500
v: 0.7240
Epoch 747/1500
v: 0.7240
Epoch 748/1500
v: 0.7240
Epoch 749/1500
v: 0.7188
Epoch 750/1500
v: 0.7188
Epoch 751/1500
v: 0.7188
Epoch 752/1500
v: 0.7188
Epoch 753/1500
v: 0.7188
Epoch 754/1500
cv: 0.7188
Epoch 755/1500
v: 0.7188
Epoch 756/1500
y: 0.7240
Epoch 757/1500
v: 0.7240
Epoch 758/1500
```

```
v: 0.7240
Epoch 759/1500
v: 0.7240
Epoch 760/1500
v: 0.7240
Epoch 761/1500
v: 0.7240
Epoch 762/1500
v: 0.7240
Epoch 763/1500
v: 0.7240
Epoch 764/1500
y: 0.7240
Epoch 765/1500
v: 0.7240
Epoch 766/1500
v: 0.7240
Epoch 767/1500
v: 0.7240
Epoch 768/1500
v: 0.7240
Epoch 769/1500
v: 0.7240
Epoch 770/1500
v: 0.7240
Epoch 771/1500
v: 0.7240
Epoch 772/1500
y: 0.7240
Epoch 773/1500
y: 0.7240
```

```
Epoch 774/1500
v: 0.7240
Epoch 775/1500
y: 0.7240
Epoch 776/1500
v: 0.7240
Epoch 777/1500
v: 0.7240
Epoch 778/1500
cy: 0.7240
Epoch 779/1500
v: 0.7240
Epoch 780/1500
v: 0.7240
Epoch 781/1500
y: 0.7240
Epoch 782/1500
v: 0.7240
Epoch 783/1500
y: 0.7240
Epoch 784/1500
v: 0.7240
Epoch 785/1500
cy: 0.7240
Epoch 786/1500
cy: 0.7240
Epoch 787/1500
v: 0.7240
Epoch 788/1500
v: 0.7240
Epoch 789/1500
```

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v: 0.7240
Epoch 790/1500
v: 0.7240
Epoch 791/1500
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Epoch 792/1500
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Epoch 793/1500
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Epoch 794/1500
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Epoch 795/1500
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Epoch 796/1500
v: 0.7240
Epoch 797/1500
v: 0.7240
Epoch 798/1500
v: 0.7240
Epoch 799/1500
v: 0.7240
Epoch 800/1500
v: 0.7240
Epoch 801/1500
y: 0.7240
Epoch 802/1500
y: 0.7240
Epoch 803/1500
y: 0.7240
Epoch 804/1500
v: 0.7240
Epoch 805/1500
```

```
cv: 0.7240
Epoch 806/1500
v: 0.7240
Epoch 807/1500
v: 0.7240
Epoch 808/1500
v: 0.7240
Epoch 809/1500
v: 0.7240
Epoch 810/1500
v: 0.7240
Epoch 811/1500
cy: 0.7240
Epoch 812/1500
v: 0.7292
Epoch 813/1500
v: 0.7292
Epoch 814/1500
v: 0.7292
Epoch 815/1500
v: 0.7292
Epoch 816/1500
v: 0.7292
Epoch 817/1500
v: 0.7292
Epoch 818/1500
v: 0.7292
Epoch 819/1500
cy: 0.7292
Epoch 820/1500
cy: 0.7292
```

```
Epoch 821/1500
cv: 0.7292
Epoch 822/1500
y: 0.7292
Epoch 823/1500
v: 0.7292
Epoch 824/1500
v: 0.7292
Epoch 825/1500
v: 0.7292
Epoch 826/1500
v: 0.7292
Epoch 827/1500
v: 0.7292
Epoch 828/1500
cy: 0.7292
Epoch 829/1500
cy: 0.7292
Epoch 830/1500
v: 0.7292
Epoch 831/1500
v: 0.7292
Epoch 832/1500
v: 0.7292
Epoch 833/1500
v: 0.7292
Epoch 834/1500
v: 0.7292
Epoch 835/1500
y: 0.7292
Epoch 836/1500
```

```
v: 0.7292
Epoch 837/1500
v: 0.7292
Epoch 838/1500
v: 0.7292
Epoch 839/1500
v: 0.7292
Epoch 840/1500
y: 0.7292
Epoch 841/1500
v: 0.7292
Epoch 842/1500
v: 0.7292
Epoch 843/1500
v: 0.7292
Epoch 844/1500
v: 0.7292
Epoch 845/1500
v: 0.7292
Epoch 846/1500
v: 0.7292
Epoch 847/1500
v: 0.7292
Epoch 848/1500
y: 0.7292
Epoch 849/1500
y: 0.7292
Epoch 850/1500
y: 0.7292
Epoch 851/1500
v: 0.7292
Epoch 852/1500
```

```
v: 0.7292
Epoch 853/1500
v: 0.7292
Epoch 854/1500
v: 0.7292
Epoch 855/1500
cy: 0.7292
Epoch 856/1500
v: 0.7292
Epoch 857/1500
v: 0.7292
Epoch 858/1500
y: 0.7292
Epoch 859/1500
v: 0.7292
Epoch 860/1500
v: 0.7292
Epoch 861/1500
v: 0.7292
Epoch 862/1500
v: 0.7292
Epoch 863/1500
v: 0.7292
Epoch 864/1500
cy: 0.7292
Epoch 865/1500
v: 0.7344
Epoch 866/1500
y: 0.7344
Epoch 867/1500
y: 0.7344
```

```
Epoch 868/1500
cv: 0.7344
Epoch 869/1500
y: 0.7344
Epoch 870/1500
v: 0.7344
Epoch 871/1500
v: 0.7344
Epoch 872/1500
v: 0.7344
Epoch 873/1500
v: 0.7344
Epoch 874/1500
v: 0.7344
Epoch 875/1500
v: 0.7344
Epoch 876/1500
v: 0.7344
Epoch 877/1500
v: 0.7344
Epoch 878/1500
v: 0.7344
Epoch 879/1500
v: 0.7344
Epoch 880/1500
v: 0.7344
Epoch 881/1500
cv: 0.7344
Epoch 882/1500
cy: 0.7344
Epoch 883/1500
```

```
v: 0.7344
Epoch 884/1500
v: 0.7344
Epoch 885/1500
v: 0.7344
Epoch 886/1500
v: 0.7344
Epoch 887/1500
v: 0.7344
Epoch 888/1500
v: 0.7344
Epoch 889/1500
v: 0.7344
Epoch 890/1500
v: 0.7344
Epoch 891/1500
v: 0.7344
Epoch 892/1500
cy: 0.7344
Epoch 893/1500
v: 0.7344
Epoch 894/1500
v: 0.7344
Epoch 895/1500
y: 0.7344
Epoch 896/1500
v: 0.7344
Epoch 897/1500
y: 0.7396
Epoch 898/1500
v: 0.7396
Epoch 899/1500
```

```
v: 0.7396
Epoch 900/1500
v: 0.7396
Epoch 901/1500
v: 0.7396
Epoch 902/1500
v: 0.7396
Epoch 903/1500
v: 0.7396
Epoch 904/1500
cy: 0.7396
Epoch 905/1500
y: 0.7396
Epoch 906/1500
v: 0.7396
Epoch 907/1500
v: 0.7396
Epoch 908/1500
v: 0.7396
Epoch 909/1500
cy: 0.7396
Epoch 910/1500
v: 0.7396
Epoch 911/1500
v: 0.7396
Epoch 912/1500
v: 0.7396
Epoch 913/1500
y: 0.7396
Epoch 914/1500
y: 0.7396
```

```
Epoch 915/1500
v: 0.7396
Epoch 916/1500
y: 0.7396
Epoch 917/1500
v: 0.7396
Epoch 918/1500
v: 0.7396
Epoch 919/1500
cy: 0.7396
Epoch 920/1500
v: 0.7396
Epoch 921/1500
cy: 0.7396
Epoch 922/1500
cy: 0.7396
Epoch 923/1500
cy: 0.7396
Epoch 924/1500
v: 0.7396
Epoch 925/1500
v: 0.7396
Epoch 926/1500
v: 0.7396
Epoch 927/1500
v: 0.7396
Epoch 928/1500
v: 0.7396
Epoch 929/1500
y: 0.7396
Epoch 930/1500
```

```
v: 0.7344
Epoch 931/1500
v: 0.7344
Epoch 932/1500
cy: 0.7344
Epoch 933/1500
v: 0.7344
Epoch 934/1500
v: 0.7344
Epoch 935/1500
v: 0.7344
Epoch 936/1500
cy: 0.7344
Epoch 937/1500
v: 0.7344
Epoch 938/1500
v: 0.7344
Epoch 939/1500
v: 0.7344
Epoch 940/1500
v: 0.7344
Epoch 941/1500
v: 0.7344
Epoch 942/1500
y: 0.7344
Epoch 943/1500
v: 0.7344
Epoch 944/1500
y: 0.7344
Epoch 945/1500
v: 0.7344
Epoch 946/1500
```

```
v: 0.7344
Epoch 947/1500
v: 0.7344
Epoch 948/1500
v: 0.7344
Epoch 949/1500
v: 0.7344
Epoch 950/1500
v: 0.7344
Epoch 951/1500
v: 0.7344
Epoch 952/1500
y: 0.7344
Epoch 953/1500
v: 0.7344
Epoch 954/1500
v: 0.7344
Epoch 955/1500
cy: 0.7344
Epoch 956/1500
cy: 0.7344
Epoch 957/1500
cy: 0.7344
Epoch 958/1500
v: 0.7344
Epoch 959/1500
cv: 0.7344
Epoch 960/1500
y: 0.7344
Epoch 961/1500
y: 0.7344
```

```
Epoch 962/1500
cv: 0.7344
Epoch 963/1500
y: 0.7344
Epoch 964/1500
v: 0.7344
Epoch 965/1500
cy: 0.7344
Epoch 966/1500
v: 0.7344
Epoch 967/1500
cy: 0.7344
Epoch 968/1500
v: 0.7344
Epoch 969/1500
y: 0.7344
Epoch 970/1500
v: 0.7344
Epoch 971/1500
v: 0.7344
Epoch 972/1500
v: 0.7344
Epoch 973/1500
v: 0.7344
Epoch 974/1500
v: 0.7344
Epoch 975/1500
v: 0.7344
Epoch 976/1500
y: 0.7344
Epoch 977/1500
```

```
v: 0.7344
Epoch 978/1500
v: 0.7344
Epoch 979/1500
v: 0.7344
Epoch 980/1500
v: 0.7344
Epoch 981/1500
v: 0.7344
Epoch 982/1500
v: 0.7344
Epoch 983/1500
v: 0.7344
Epoch 984/1500
v: 0.7344
Epoch 985/1500
v: 0.7344
Epoch 986/1500
v: 0.7344
Epoch 987/1500
cy: 0.7344
Epoch 988/1500
v: 0.7344
Epoch 989/1500
cy: 0.7344
Epoch 990/1500
cy: 0.7344
Epoch 991/1500
y: 0.7344
Epoch 992/1500
v: 0.7344
Epoch 993/1500
```

```
v: 0.7344
Epoch 994/1500
cv: 0.7344
Epoch 995/1500
v: 0.7344
Epoch 996/1500
v: 0.7344
Epoch 997/1500
v: 0.7344
Epoch 998/1500
v: 0.7344
Epoch 999/1500
v: 0.7344
Epoch 1000/1500
v: 0.7344
Epoch 1001/1500
v: 0.7344
Epoch 1002/1500
v: 0.7344
Epoch 1003/1500
v: 0.7344
Epoch 1004/1500
v: 0.7344
Epoch 1005/1500
v: 0.7344
Epoch 1006/1500
v: 0.7344
Epoch 1007/1500
cy: 0.7344
Epoch 1008/1500
y: 0.7344
```

```
Epoch 1009/1500
v: 0.7344
Epoch 1010/1500
y: 0.7344
Epoch 1011/1500
v: 0.7344
Epoch 1012/1500
v: 0.7344
Epoch 1013/1500
v: 0.7344
Epoch 1014/1500
v: 0.7344
Epoch 1015/1500
v: 0.7344
Epoch 1016/1500
cy: 0.7344
Epoch 1017/1500
v: 0.7396
Epoch 1018/1500
v: 0.7396
Epoch 1019/1500
v: 0.7396
Epoch 1020/1500
v: 0.7396
Epoch 1021/1500
v: 0.7396
Epoch 1022/1500
v: 0.7396
Epoch 1023/1500
y: 0.7396
Epoch 1024/1500
```

```
y: 0.7396
Epoch 1025/1500
v: 0.7396
Epoch 1026/1500
cy: 0.7396
Epoch 1027/1500
cy: 0.7396
Epoch 1028/1500
v: 0.7396
Epoch 1029/1500
cy: 0.7396
Epoch 1030/1500
cy: 0.7396
Epoch 1031/1500
v: 0.7396
Epoch 1032/1500
v: 0.7396
Epoch 1033/1500
v: 0.7396
Epoch 1034/1500
v: 0.7396
Epoch 1035/1500
cy: 0.7396
Epoch 1036/1500
y: 0.7396
Epoch 1037/1500
cy: 0.7396
Epoch 1038/1500
cy: 0.7396
Epoch 1039/1500
cv: 0.7396
Epoch 1040/1500
```

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```
cv: 0.7396
Epoch 1041/1500
cv: 0.7396
Epoch 1042/1500
cy: 0.7396
Epoch 1043/1500
v: 0.7396
Epoch 1044/1500
v: 0.7396
Epoch 1045/1500
v: 0.7396
Epoch 1046/1500
y: 0.7396
Epoch 1047/1500
v: 0.7396
Epoch 1048/1500
v: 0.7396
Epoch 1049/1500
v: 0.7396
Epoch 1050/1500
cy: 0.7396
Epoch 1051/1500
cy: 0.7396
Epoch 1052/1500
v: 0.7396
Epoch 1053/1500
v: 0.7396
Epoch 1054/1500
y: 0.7396
Epoch 1055/1500
y: 0.7396
```

```
Epoch 1056/1500
v: 0.7396
Epoch 1057/1500
y: 0.7396
Epoch 1058/1500
v: 0.7396
Epoch 1059/1500
v: 0.7396
Epoch 1060/1500
v: 0.7396
Epoch 1061/1500
cy: 0.7396
Epoch 1062/1500
v: 0.7396
Epoch 1063/1500
cy: 0.7396
Epoch 1064/1500
cy: 0.7396
Epoch 1065/1500
v: 0.7396
Epoch 1066/1500
v: 0.7396
Epoch 1067/1500
cy: 0.7396
Epoch 1068/1500
cy: 0.7396
Epoch 1069/1500
v: 0.7396
Epoch 1070/1500
y: 0.7396
Epoch 1071/1500
```

```
y: 0.7396
Epoch 1072/1500
v: 0.7396
Epoch 1073/1500
v: 0.7396
Epoch 1074/1500
cy: 0.7396
Epoch 1075/1500
v: 0.7396
Epoch 1076/1500
v: 0.7396
Epoch 1077/1500
v: 0.7396
Epoch 1078/1500
cy: 0.7396
Epoch 1079/1500
v: 0.7396
Epoch 1080/1500
v: 0.7396
Epoch 1081/1500
v: 0.7396
Epoch 1082/1500
v: 0.7396
Epoch 1083/1500
y: 0.7396
Epoch 1084/1500
y: 0.7396
Epoch 1085/1500
y: 0.7396
Epoch 1086/1500
v: 0.7396
Epoch 1087/1500
```

```
v: 0.7396
Epoch 1088/1500
cv: 0.7396
Epoch 1089/1500
y: 0.7396
Epoch 1090/1500
v: 0.7396
Epoch 1091/1500
v: 0.7396
Epoch 1092/1500
v: 0.7396
Epoch 1093/1500
y: 0.7396
Epoch 1094/1500
cy: 0.7396
Epoch 1095/1500
cy: 0.7396
Epoch 1096/1500
v: 0.7396
Epoch 1097/1500
v: 0.7396
Epoch 1098/1500
v: 0.7396
Epoch 1099/1500
v: 0.7396
Epoch 1100/1500
v: 0.7396
Epoch 1101/1500
y: 0.7396
Epoch 1102/1500
y: 0.7396
```

```
Epoch 1103/1500
v: 0.7396
Epoch 1104/1500
y: 0.7396
Epoch 1105/1500
v: 0.7396
Epoch 1106/1500
v: 0.7396
Epoch 1107/1500
v: 0.7396
Epoch 1108/1500
v: 0.7396
Epoch 1109/1500
v: 0.7396
Epoch 1110/1500
v: 0.7396
Epoch 1111/1500
v: 0.7396
Epoch 1112/1500
v: 0.7396
Epoch 1113/1500
cy: 0.7396
Epoch 1114/1500
v: 0.7396
Epoch 1115/1500
cy: 0.7396
Epoch 1116/1500
v: 0.7396
Epoch 1117/1500
y: 0.7396
Epoch 1118/1500
```

```
y: 0.7396
Epoch 1119/1500
v: 0.7396
Epoch 1120/1500
v: 0.7396
Epoch 1121/1500
cy: 0.7396
Epoch 1122/1500
v: 0.7396
Epoch 1123/1500
v: 0.7396
Epoch 1124/1500
cy: 0.7396
Epoch 1125/1500
v: 0.7396
Epoch 1126/1500
v: 0.7396
Epoch 1127/1500
v: 0.7396
Epoch 1128/1500
v: 0.7396
Epoch 1129/1500
v: 0.7396
Epoch 1130/1500
y: 0.7396
Epoch 1131/1500
v: 0.7396
Epoch 1132/1500
y: 0.7396
Epoch 1133/1500
v: 0.7396
Epoch 1134/1500
```

```
v: 0.7396
Epoch 1135/1500
cv: 0.7396
Epoch 1136/1500
v: 0.7396
Epoch 1137/1500
v: 0.7396
Epoch 1138/1500
v: 0.7396
Epoch 1139/1500
v: 0.7396
Epoch 1140/1500
y: 0.7396
Epoch 1141/1500
v: 0.7396
Epoch 1142/1500
v: 0.7396
Epoch 1143/1500
v: 0.7396
Epoch 1144/1500
v: 0.7396
Epoch 1145/1500
v: 0.7396
Epoch 1146/1500
v: 0.7396
Epoch 1147/1500
cy: 0.7396
Epoch 1148/1500
cy: 0.7396
Epoch 1149/1500
y: 0.7396
```

```
Epoch 1150/1500
v: 0.7396
Epoch 1151/1500
y: 0.7396
Epoch 1152/1500
v: 0.7396
Epoch 1153/1500
v: 0.7396
Epoch 1154/1500
v: 0.7396
Epoch 1155/1500
v: 0.7396
Epoch 1156/1500
v: 0.7396
Epoch 1157/1500
v: 0.7448
Epoch 1158/1500
v: 0.7448
Epoch 1159/1500
v: 0.7448
Epoch 1160/1500
v: 0.7448
Epoch 1161/1500
v: 0.7396
Epoch 1162/1500
v: 0.7396
Epoch 1163/1500
v: 0.7396
Epoch 1164/1500
v: 0.7448
Epoch 1165/1500
```

```
v: 0.7396
Epoch 1166/1500
v: 0.7448
Epoch 1167/1500
cy: 0.7448
Epoch 1168/1500
v: 0.7448
Epoch 1169/1500
v: 0.7448
Epoch 1170/1500
v: 0.7448
Epoch 1171/1500
v: 0.7448
Epoch 1172/1500
cy: 0.7448
Epoch 1173/1500
v: 0.7448
Epoch 1174/1500
v: 0.7448
Epoch 1175/1500
v: 0.7448
Epoch 1176/1500
v: 0.7448
Epoch 1177/1500
y: 0.7448
Epoch 1178/1500
v: 0.7448
Epoch 1179/1500
y: 0.7448
Epoch 1180/1500
v: 0.7448
Epoch 1181/1500
```

```
v: 0.7396
Epoch 1182/1500
v: 0.7448
Epoch 1183/1500
v: 0.7396
Epoch 1184/1500
v: 0.7396
Epoch 1185/1500
v: 0.7396
Epoch 1186/1500
v: 0.7396
Epoch 1187/1500
y: 0.7396
Epoch 1188/1500
v: 0.7396
Epoch 1189/1500
v: 0.7396
Epoch 1190/1500
v: 0.7396
Epoch 1191/1500
v: 0.7396
Epoch 1192/1500
v: 0.7396
Epoch 1193/1500
cy: 0.7396
Epoch 1194/1500
v: 0.7396
Epoch 1195/1500
cy: 0.7396
Epoch 1196/1500
cy: 0.7396
```

```
Epoch 1197/1500
v: 0.7448
Epoch 1198/1500
cy: 0.7448
Epoch 1199/1500
cy: 0.7448
Epoch 1200/1500
v: 0.7448
Epoch 1201/1500
v: 0.7448
Epoch 1202/1500
v: 0.7448
Epoch 1203/1500
v: 0.7448
Epoch 1204/1500
v: 0.7448
Epoch 1205/1500
v: 0.7448
Epoch 1206/1500
v: 0.7448
Epoch 1207/1500
v: 0.7448
Epoch 1208/1500
v: 0.7448
Epoch 1209/1500
v: 0.7448
Epoch 1210/1500
v: 0.7448
Epoch 1211/1500
y: 0.7448
Epoch 1212/1500
```

```
v: 0.7448
Epoch 1213/1500
v: 0.7448
Epoch 1214/1500
v: 0.7448
Epoch 1215/1500
v: 0.7396
Epoch 1216/1500
v: 0.7396
Epoch 1217/1500
v: 0.7396
Epoch 1218/1500
v: 0.7396
Epoch 1219/1500
v: 0.7396
Epoch 1220/1500
v: 0.7396
Epoch 1221/1500
v: 0.7396
Epoch 1222/1500
v: 0.7396
Epoch 1223/1500
v: 0.7396
Epoch 1224/1500
y: 0.7396
Epoch 1225/1500
y: 0.7396
Epoch 1226/1500
y: 0.7396
Epoch 1227/1500
v: 0.7396
Epoch 1228/1500
```

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```
cv: 0.7396
Epoch 1229/1500
v: 0.7396
Epoch 1230/1500
v: 0.7396
Epoch 1231/1500
v: 0.7396
Epoch 1232/1500
v: 0.7396
Epoch 1233/1500
v: 0.7396
Epoch 1234/1500
y: 0.7396
Epoch 1235/1500
v: 0.7396
Epoch 1236/1500
v: 0.7396
Epoch 1237/1500
v: 0.7396
Epoch 1238/1500
v: 0.7344
Epoch 1239/1500
v: 0.7396
Epoch 1240/1500
cy: 0.7344
Epoch 1241/1500
v: 0.7344
Epoch 1242/1500
y: 0.7344
Epoch 1243/1500
y: 0.7344
```

```
Epoch 1244/1500
cv: 0.7344
Epoch 1245/1500
y: 0.7344
Epoch 1246/1500
v: 0.7344
Epoch 1247/1500
v: 0.7344
Epoch 1248/1500
v: 0.7344
Epoch 1249/1500
v: 0.7344
Epoch 1250/1500
v: 0.7344
Epoch 1251/1500
y: 0.7344
Epoch 1252/1500
cy: 0.7344
Epoch 1253/1500
v: 0.7344
Epoch 1254/1500
v: 0.7344
Epoch 1255/1500
v: 0.7344
Epoch 1256/1500
v: 0.7344
Epoch 1257/1500
v: 0.7344
Epoch 1258/1500
v: 0.7344
Epoch 1259/1500
```

```
v: 0.7344
Epoch 1260/1500
v: 0.7344
Epoch 1261/1500
v: 0.7344
Epoch 1262/1500
v: 0.7344
Epoch 1263/1500
v: 0.7344
Epoch 1264/1500
v: 0.7344
Epoch 1265/1500
v: 0.7344
Epoch 1266/1500
v: 0.7344
Epoch 1267/1500
v: 0.7344
Epoch 1268/1500
v: 0.7344
Epoch 1269/1500
v: 0.7344
Epoch 1270/1500
v: 0.7344
Epoch 1271/1500
y: 0.7344
Epoch 1272/1500
v: 0.7344
Epoch 1273/1500
y: 0.7344
Epoch 1274/1500
v: 0.7344
Epoch 1275/1500
```

```
v: 0.7344
Epoch 1276/1500
v: 0.7344
Epoch 1277/1500
v: 0.7344
Epoch 1278/1500
v: 0.7344
Epoch 1279/1500
v: 0.7344
Epoch 1280/1500
cv: 0.7344
Epoch 1281/1500
v: 0.7344
Epoch 1282/1500
v: 0.7344
Epoch 1283/1500
v: 0.7344
Epoch 1284/1500
v: 0.7344
Epoch 1285/1500
v: 0.7344
Epoch 1286/1500
v: 0.7344
Epoch 1287/1500
v: 0.7344
Epoch 1288/1500
v: 0.7344
Epoch 1289/1500
y: 0.7344
Epoch 1290/1500
y: 0.7344
```

```
Epoch 1291/1500
v: 0.7344
Epoch 1292/1500
v: 0.7344
Epoch 1293/1500
v: 0.7344
Epoch 1294/1500
v: 0.7344
Epoch 1295/1500
cy: 0.7344
Epoch 1296/1500
v: 0.7344
Epoch 1297/1500
v: 0.7344
Epoch 1298/1500
y: 0.7344
Epoch 1299/1500
v: 0.7344
Epoch 1300/1500
v: 0.7344
Epoch 1301/1500
v: 0.7344
Epoch 1302/1500
v: 0.7344
Epoch 1303/1500
v: 0.7344
Epoch 1304/1500
v: 0.7344
Epoch 1305/1500
y: 0.7344
Epoch 1306/1500
```

```
cy: 0.7344
Epoch 1307/1500
v: 0.7344
Epoch 1308/1500
v: 0.7344
Epoch 1309/1500
v: 0.7344
Epoch 1310/1500
v: 0.7344
Epoch 1311/1500
v: 0.7344
Epoch 1312/1500
v: 0.7344
Epoch 1313/1500
v: 0.7344
Epoch 1314/1500
v: 0.7344
Epoch 1315/1500
v: 0.7344
Epoch 1316/1500
v: 0.7344
Epoch 1317/1500
v: 0.7344
Epoch 1318/1500
y: 0.7344
Epoch 1319/1500
y: 0.7344
Epoch 1320/1500
y: 0.7344
Epoch 1321/1500
v: 0.7344
Epoch 1322/1500
```

```
v: 0.7344
Epoch 1323/1500
v: 0.7344
Epoch 1324/1500
v: 0.7344
Epoch 1325/1500
v: 0.7344
Epoch 1326/1500
v: 0.7344
Epoch 1327/1500
v: 0.7344
Epoch 1328/1500
v: 0.7344
Epoch 1329/1500
v: 0.7344
Epoch 1330/1500
v: 0.7344
Epoch 1331/1500
v: 0.7344
Epoch 1332/1500
v: 0.7344
Epoch 1333/1500
v: 0.7344
Epoch 1334/1500
v: 0.7344
Epoch 1335/1500
cy: 0.7344
Epoch 1336/1500
y: 0.7344
Epoch 1337/1500
y: 0.7292
```

```
Epoch 1338/1500
v: 0.7292
Epoch 1339/1500
y: 0.7292
Epoch 1340/1500
v: 0.7292
Epoch 1341/1500
v: 0.7292
Epoch 1342/1500
v: 0.7292
Epoch 1343/1500
v: 0.7292
Epoch 1344/1500
v: 0.7292
Epoch 1345/1500
y: 0.7292
Epoch 1346/1500
v: 0.7292
Epoch 1347/1500
v: 0.7292
Epoch 1348/1500
v: 0.7292
Epoch 1349/1500
v: 0.7292
Epoch 1350/1500
v: 0.7344
Epoch 1351/1500
v: 0.7344
Epoch 1352/1500
v: 0.7292
Epoch 1353/1500
```

```
v: 0.7344
Epoch 1354/1500
v: 0.7344
Epoch 1355/1500
v: 0.7344
Epoch 1356/1500
v: 0.7344
Epoch 1357/1500
v: 0.7344
Epoch 1358/1500
v: 0.7344
Epoch 1359/1500
v: 0.7344
Epoch 1360/1500
v: 0.7344
Epoch 1361/1500
v: 0.7344
Epoch 1362/1500
v: 0.7344
Epoch 1363/1500
v: 0.7344
Epoch 1364/1500
v: 0.7344
Epoch 1365/1500
y: 0.7344
Epoch 1366/1500
v: 0.7344
Epoch 1367/1500
y: 0.7344
Epoch 1368/1500
v: 0.7344
Epoch 1369/1500
```

```
v: 0.7344
Epoch 1370/1500
v: 0.7344
Epoch 1371/1500
v: 0.7344
Epoch 1372/1500
v: 0.7344
Epoch 1373/1500
v: 0.7344
Epoch 1374/1500
v: 0.7344
Epoch 1375/1500
y: 0.7396
Epoch 1376/1500
v: 0.7344
Epoch 1377/1500
v: 0.7344
Epoch 1378/1500
v: 0.7396
Epoch 1379/1500
v: 0.7396
Epoch 1380/1500
cy: 0.7396
Epoch 1381/1500
v: 0.7396
Epoch 1382/1500
v: 0.7396
Epoch 1383/1500
cy: 0.7396
Epoch 1384/1500
cy: 0.7396
```

```
Epoch 1385/1500
cv: 0.7396
Epoch 1386/1500
y: 0.7396
Epoch 1387/1500
v: 0.7396
Epoch 1388/1500
v: 0.7396
Epoch 1389/1500
v: 0.7396
Epoch 1390/1500
cy: 0.7396
Epoch 1391/1500
v: 0.7396
Epoch 1392/1500
v: 0.7396
Epoch 1393/1500
v: 0.7396
Epoch 1394/1500
v: 0.7396
Epoch 1395/1500
v: 0.7396
Epoch 1396/1500
v: 0.7396
Epoch 1397/1500
v: 0.7396
Epoch 1398/1500
v: 0.7396
Epoch 1399/1500
y: 0.7396
Epoch 1400/1500
```

```
v: 0.7396
Epoch 1401/1500
v: 0.7396
Epoch 1402/1500
v: 0.7396
Epoch 1403/1500
v: 0.7396
Epoch 1404/1500
v: 0.7396
Epoch 1405/1500
v: 0.7396
Epoch 1406/1500
v: 0.7396
Epoch 1407/1500
v: 0.7396
Epoch 1408/1500
v: 0.7396
Epoch 1409/1500
v: 0.7396
Epoch 1410/1500
v: 0.7396
Epoch 1411/1500
cy: 0.7396
Epoch 1412/1500
cy: 0.7396
Epoch 1413/1500
y: 0.7396
Epoch 1414/1500
y: 0.7396
Epoch 1415/1500
v: 0.7396
Epoch 1416/1500
```

```
v: 0.7396
Epoch 1417/1500
v: 0.7396
Epoch 1418/1500
cy: 0.7396
Epoch 1419/1500
v: 0.7396
Epoch 1420/1500
v: 0.7396
Epoch 1421/1500
v: 0.7396
Epoch 1422/1500
y: 0.7396
Epoch 1423/1500
v: 0.7396
Epoch 1424/1500
v: 0.7396
Epoch 1425/1500
v: 0.7396
Epoch 1426/1500
v: 0.7396
Epoch 1427/1500
v: 0.7396
Epoch 1428/1500
v: 0.7396
Epoch 1429/1500
v: 0.7396
Epoch 1430/1500
y: 0.7396
Epoch 1431/1500
y: 0.7396
```

```
Epoch 1432/1500
cv: 0.7396
Epoch 1433/1500
y: 0.7396
Epoch 1434/1500
cy: 0.7396
Epoch 1435/1500
cy: 0.7396
Epoch 1436/1500
v: 0.7396
Epoch 1437/1500
v: 0.7396
Epoch 1438/1500
v: 0.7396
Epoch 1439/1500
v: 0.7396
Epoch 1440/1500
v: 0.7396
Epoch 1441/1500
v: 0.7396
Epoch 1442/1500
v: 0.7396
Epoch 1443/1500
v: 0.7396
Epoch 1444/1500
v: 0.7396
Epoch 1445/1500
v: 0.7396
Epoch 1446/1500
y: 0.7396
Epoch 1447/1500
```

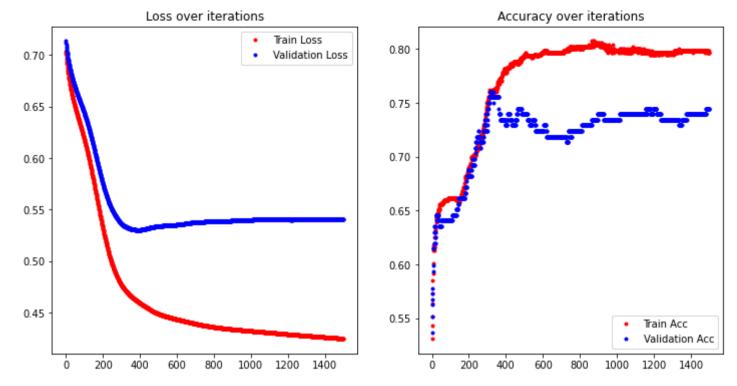
```
v: 0.7396
Epoch 1448/1500
v: 0.7396
Epoch 1449/1500
v: 0.7396
Epoch 1450/1500
v: 0.7396
Epoch 1451/1500
v: 0.7396
Epoch 1452/1500
v: 0.7396
Epoch 1453/1500
v: 0.7396
Epoch 1454/1500
v: 0.7396
Epoch 1455/1500
v: 0.7396
Epoch 1456/1500
v: 0.7396
Epoch 1457/1500
v: 0.7396
Epoch 1458/1500
v: 0.7396
Epoch 1459/1500
y: 0.7396
Epoch 1460/1500
v: 0.7396
Epoch 1461/1500
y: 0.7396
Epoch 1462/1500
v: 0.7396
Epoch 1463/1500
```

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```
cv: 0.7396
Epoch 1464/1500
v: 0.7396
Epoch 1465/1500
v: 0.7396
Epoch 1466/1500
v: 0.7396
Epoch 1467/1500
v: 0.7396
Epoch 1468/1500
v: 0.7396
Epoch 1469/1500
y: 0.7396
Epoch 1470/1500
v: 0.7396
Epoch 1471/1500
v: 0.7396
Epoch 1472/1500
v: 0.7396
Epoch 1473/1500
v: 0.7396
Epoch 1474/1500
v: 0.7396
Epoch 1475/1500
v: 0.7396
Epoch 1476/1500
v: 0.7396
Epoch 1477/1500
cy: 0.7396
Epoch 1478/1500
y: 0.7396
```

```
Epoch 1479/1500
v: 0.7396
Epoch 1480/1500
y: 0.7396
Epoch 1481/1500
v: 0.7396
Epoch 1482/1500
v: 0.7396
Epoch 1483/1500
v: 0.7448
Epoch 1484/1500
cy: 0.7448
Epoch 1485/1500
v: 0.7448
Epoch 1486/1500
v: 0.7448
Epoch 1487/1500
v: 0.7448
Epoch 1488/1500
v: 0.7448
Epoch 1489/1500
v: 0.7448
Epoch 1490/1500
v: 0.7448
Epoch 1491/1500
v: 0.7448
Epoch 1492/1500
v: 0.7448
Epoch 1493/1500
v: 0.7448
Epoch 1494/1500
```

```
v: 0.7448
     Epoch 1495/1500
     v: 0.7448
     Epoch 1496/1500
     v: 0.7448
     Epoch 1497/1500
     v: 0.7448
     Epoch 1498/1500
     v: 0.7448
     Epoch 1499/1500
     v: 0.7448
     Epoch 1500/1500
     y: 0.7448
In [28]:
     run hist 2.history.keys()
Out[28]: dict keys(['loss', 'accuracy', 'val loss', 'val accuracy'])
In [29]:
     n = len(run hist 2.history["loss"])
     fig = plt.figure(figsize=(12, 6))
     ax = fig.add subplot(1, 2, 1)
     ax.plot(range(n), (run hist 2.history["loss"]),'r.', label="Train Loss")
     ax.plot(range(n), (run hist 2.history["val loss"]), 'b.', label="Validation Loss")
     ax.legend()
     ax.set title('Loss over iterations')
     ax = fig.add subplot(1, 2, 2)
     ax.plot(range(n), (run hist 2.history["accuracy"]), 'r.', label="Train Acc")
     ax.plot(range(n), (run hist 2.history["val accuracy"]), 'b.', label="Validation Acc")
     ax.legend(loc='lower right')
     ax.set title('Accuracy over iterations')
Out[29]: Text(0.5, 1.0, 'Accuracy over iterations')
```



Loss over iterations: Pela curva de validation loss, a partir da iteração 400 houve uma piora na maneira em que o modelo se ajusta a novos dados, e o ajuste a dados de treino seque caindo, o ideal.

Accuracy over iterations: A accurácia da validação de novos dados cai em torno da iteração número 300, enquanto para dados de treino segue a melhorar.

Este modelo para dados de validação piora muito a partir da iteração 300.

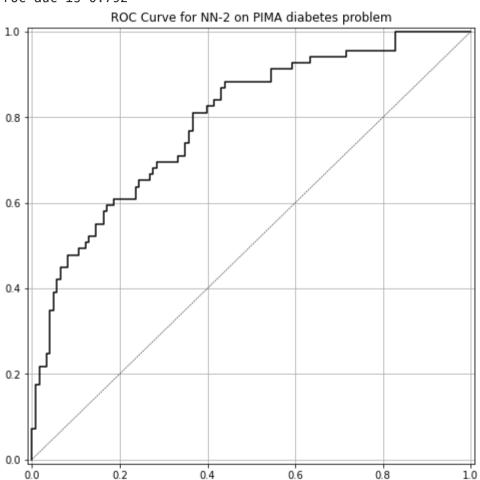
```
In [31]: #y_pred_class_nn_2 = model_2.predict_classes(X_test_norm)
    predict_x = model_1.predict(X_test)
    y_pred_class_nn_2 = np.argmax(predict_x,axis=1)

    y_pred_prob_nn_2 = model_2.predict(X_test_norm)
    print('')
    print('accuracy is {:.3f}'.format(accuracy_score(y_test,y_pred_class_nn_2)))
    print('roc-auc is {:.3f}'.format(roc_auc_score(y_test,y_pred_prob_nn_2)))
```

```
plot_roc(y_test, y_pred_prob_nn_2, 'NN-2')
```

```
6/6 [======] - 0s 4ms/step 6/6 [=========] - 0s 3ms/step
```

accuracy is 0.641 roc-auc is 0.792



A acurácia, previsão de pessoas que têm diabetes no modelo ficou em torno de 64%. A rurva ROC-AUC para avaliar a performance do modelo nos diz que nosso modelo consegue distinguir entre pacientes com e sem diabetes a uma taxa de aproximadamente 79%.

In []: