

Student Performances on Exam: Should they take a preparation course?

Use Tensorflow to predict if a student took a preparation course based on their test scores, gender, and ethnicity.

```
In [1]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
```

```
In [2]: df = pd.read_csv('StudentsPerformance.csv')
```

```
In [5]: df.head(10)
```

Out[5]:

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score	writing score
0	female	group B	bachelor's degree	standard	none	72	72	74
1	female	group C	some college	standard	completed	69	90	88
2	female	group B	master's degree	standard	none	90	95	93
3	male	group A	associate's degree	free/reduced	none	47	57	44
4	male	group C	some college	standard	none	76	78	75
5	female	group B	associate's degree	standard	none	71	83	78
6	female	group B	some college	standard	completed	88	95	92
7	male	group B	some college	free/reduced	none	40	43	39
8	male	group D	high school	free/reduced	completed	64	64	67
9	female	group B	high school	free/reduced	none	38	60	50

```
In [6]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 8 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   gender                                1000 non-null   object
1   race/ethnicity                        1000 non-null   object
2   parental level of education           1000 non-null   object
3   lunch                                 1000 non-null   object
4   test preparation course               1000 non-null   object
5   math score                           1000 non-null   int64
6   reading score                        1000 non-null   int64
7   writing score                         1000 non-null   int64
dtypes: int64(3), object(5)
memory usage: 62.6+ KB
```

```
In [7]: # What is the mean score for each exam?

df.describe()
```

Out[7]:

	math score	reading score	writing score
count	1000.00000	1000.000000	1000.000000
mean	66.08900	69.169000	68.054000
std	15.16308	14.600192	15.195657
min	0.00000	17.000000	10.000000
25%	57.00000	59.000000	57.750000
50%	66.00000	70.000000	69.000000
75%	77.00000	79.000000	79.000000
max	100.00000	100.000000	100.000000

In [8]: *# No null values!*

```
df.isnull().sum()
```

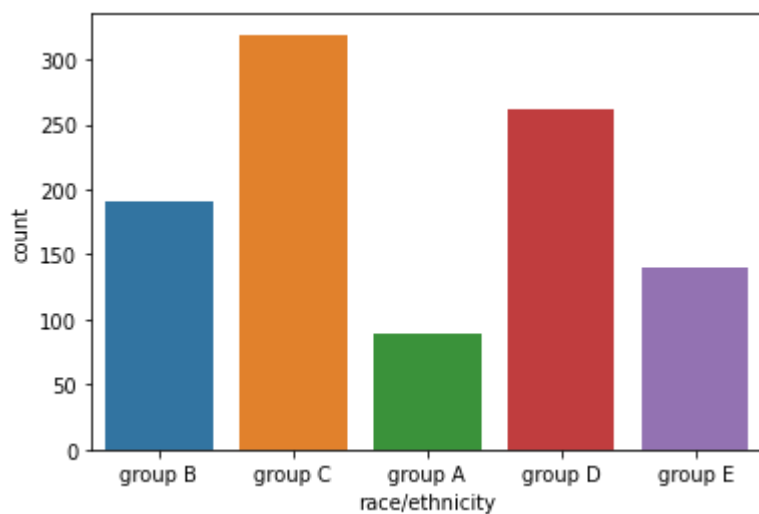
```
Out[8]: gender                0
race/ethnicity              0
parental level of education  0
lunch                      0
test preparation course     0
math score                  0
reading score               0
writing score               0
dtype: int64
```

In [13]: 

```
sns.countplot(x='race/ethnicity', data=df)
```

Out[13]: 

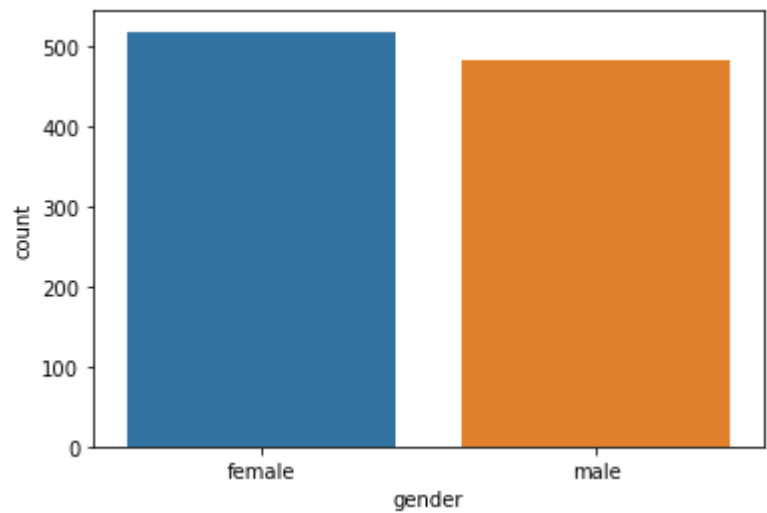
```
<AxesSubplot:xlabel='race/ethnicity', ylabel='count'>
```

In [14]: 

```
sns.countplot(x='gender', data=df)
```

Out[14]: 

```
<AxesSubplot:xlabel='gender', ylabel='count'>
```



```
In [18]: df['gender'].value_counts()
```

```
Out[18]: female    518
male        482
Name: gender, dtype: int64
```

```
In [19]: df['race/ethnicity'].value_counts()
```

```
Out[19]: group C    319
group D    262
group B    190
group E    140
group A     89
Name: race/ethnicity, dtype: int64
```

```
In [24]: type_dummies = pd.get_dummies(df['gender'],drop_first=True)
df = pd.concat([df,type_dummies],axis=1)
```

```
In [39]: df.drop(['gender'], axis=1)
```

Out[39]:

	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score	writing score	male
0	group B	bachelor's degree	standard	none	72	72	74	0
1	group C	some college	standard	completed	69	90	88	0
2	group B	master's degree	standard	none	90	95	93	0
3	group A	associate's degree	free/reduced	none	47	57	44	1
4	group C	some college	standard	none	76	78	75	1
...	...	...	...	...	...	...	...	...
995	group E	master's degree	standard	completed	88	99	95	0
996	group C	high school	free/reduced	none	62	55	55	1
997	group C	high school	free/reduced	completed	59	71	65	0
998	group D	some college	standard	completed	68	78	77	0

	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score	writing score	male
999	group D	some college	free/reduced	none	77	86	86	0

1000 rows × 8 columns

In [42]:

df.drop(['gender'], axis=1, inplace=True)

In [43]:

df.head()

Out[43]:

	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score	writing score	male
0	group B	bachelor's degree	standard	none	72	72	74	0
1	group C	some college	standard	completed	69	90	88	0
2	group B	master's degree	standard	none	90	95	93	0
3	group A	associate's degree	free/reduced	none	47	57	44	1
4	group C	some college	standard	none	76	78	75	1

In [44]:

type\_dummies = pd.get\_dummies(df['race/ethnicity'],drop\_first=True)  
df = pd.concat([df,type\_dummies],axis=1)

In [48]:

df.drop(['race/ethnicity'], axis=1, inplace=True)

In [49]:

df.head()

Out[49]:

	parental level of education	lunch	test preparation course	math score	reading score	writing score	male	group B	group C	group D	group E
0	bachelor's degree	standard	none	72	72	74	0	1	0	0	0
1	some college	standard	completed	69	90	88	0	0	1	0	0
2	master's degree	standard	none	90	95	93	0	1	0	0	0
3	associate's degree	free/reduced	none	47	57	44	1	0	0	0	0
4	some college	standard	none	76	78	75	1	0	1	0	0

In [50]:

type\_dummies = pd.get\_dummies(df['parental level of education'],drop\_first=True)  
df = pd.concat([df,type\_dummies],axis=1)

In [51]:

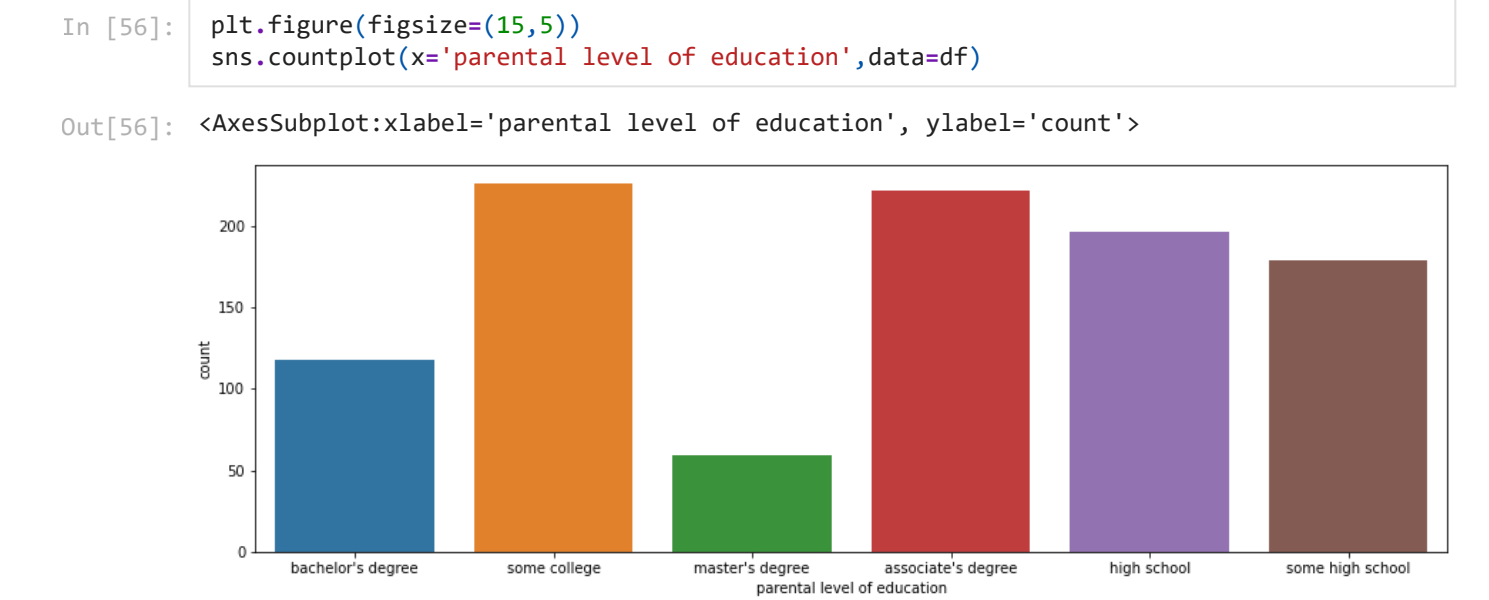
df.head()

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StudentsPerformanceExam

Out[51]:

	parental level of education	lunch	test preparation course	math score	reading score	writing score	male	group B	group C	group D	group E
0	bachelor's degree	standard	none	72	72	74	0	1	0	0	0
1	some college	standard	completed	69	90	88	0	0	1	0	0
2	master's degree	standard	none	90	95	93	0	1	0	0	0
3	associate's degree	free/reduced	none	47	57	44	1	0	0	0	0
4	some college	standard	none	76	78	75	1	0	1	0	0



In [57]:

```
df.drop(['parental level of education'],axis=1,inplace=True)
```

In [58]:

```
df.head()
```

Out[58]:

	lunch	test preparation course	math score	reading score	writing score	male	group B	group C	group D	group E	bachelor's degree
0	standard	none	72	72	74	0	1	0	0	0	1
1	standard	completed	69	90	88	0	0	1	0	0	0
2	standard	none	90	95	93	0	1	0	0	0	0
3	free/reduced	none	47	57	44	1	0	0	0	0	0
4	standard	none	76	78	75	1	0	1	0	0	0

In [59]:

```
type_dummies = pd.get_dummies(df['lunch'],drop_first=True)
```

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5/26

```
df = pd.concat([df,type_dummies],axis=1)
```

```
In [60]: df.head()
```

Out[60]:

	lunch	test preparation course	math score	reading score	writing score	male	group B	group C	group D	group E	bachelor's degree
0	standard	none	72	72	74	0	1	0	0	0	1
1	standard	completed	69	90	88	0	0	1	0	0	0
2	standard	none	90	95	93	0	1	0	0	0	0
3	free/reduced	none	47	57	44	1	0	0	0	0	0
4	standard	none	76	78	75	1	0	1	0	0	0

```
In [61]: df.drop(['lunch'],axis=1,inplace=True)
```

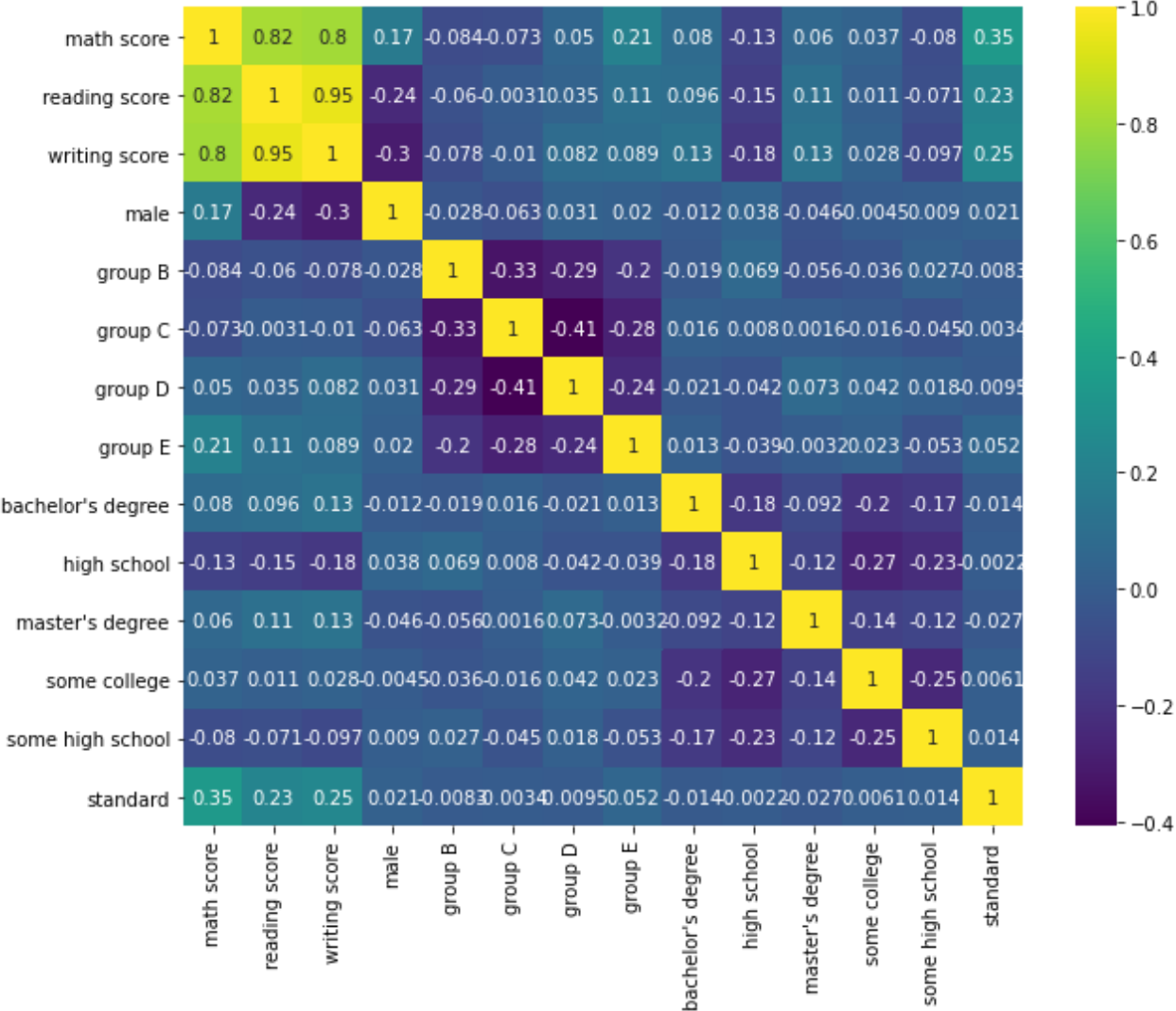
```
In [63]: df.head()
```

Out[63]:

	test preparation course	math score	reading score	writing score	male	group B	group C	group D	group E	bachelor's degree	high school	maste degr
0	none	72	72	74	0	1	0	0	0	1	0	
1	completed	69	90	88	0	0	1	0	0	0	0	
2	none	90	95	93	0	1	0	0	0	0	0	
3	none	47	57	44	1	0	0	0	0	0	0	
4	none	76	78	75	1	0	1	0	0	0	0	

```
In [80]: plt.figure(figsize=(10,8))
sns.heatmap(df.corr(), cmap='viridis',annot=True)
```

```
Out[80]: <AxesSubplot:>
```



In [98]:

df.head()

Out[98]:

	test preparation course	math score	reading score	writing score	male	group B	group C	group D	group E	bachelor's degree	high school	master's degree
0	none	72	72	74	0	1	0	0	0	1	0	
1	completed	69	90	88	0	0	1	0	0	0	0	
2	none	90	95	93	0	1	0	0	0	0	0	
3	none	47	57	44	1	0	0	0	0	0	0	
4	none	76	78	75	1	0	1	0	0	0	0	

In [104...]

type(df['math score'][1])

Out[104...]

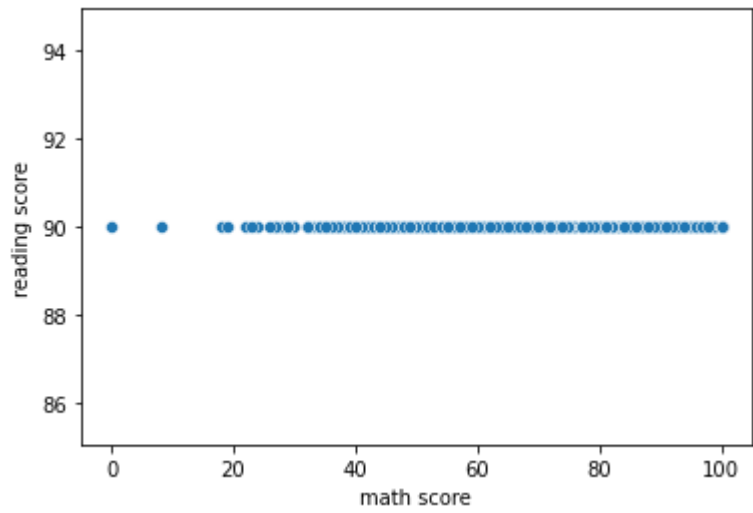
numpy.int64

In [116...]

sns.scatterplot(x='math score',y='reading score', data=df)

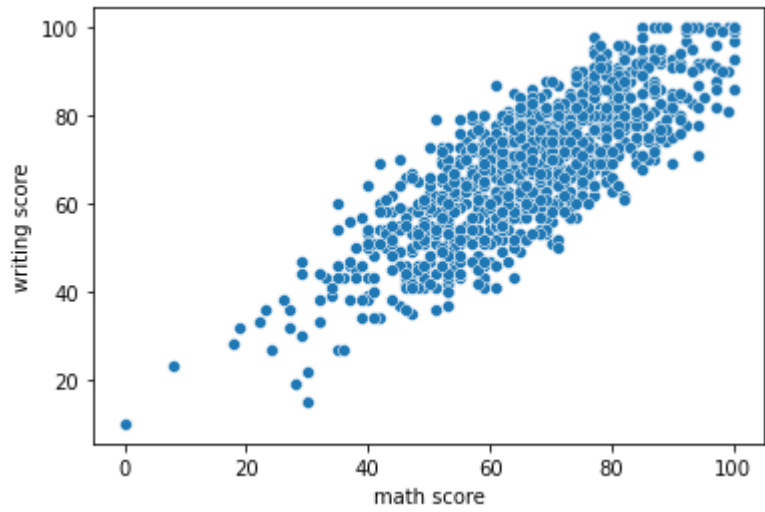
Out[116...]

<AxesSubplot:xlabel='math score', ylabel='reading score'>



```
In [117...] sns.scatterplot(x='math score',y='writing score', data=df)
```

Out[117...] <AxesSubplot:xlabel='math score', ylabel='writing score'>



```
In [130...] type_dummies = pd.get_dummies(df['test preparation course'],drop_first=True)
df = pd.concat([df,type_dummies],axis=1)
```

```
In [131...] df.head()
```

Out[131...]

	test preparation course	math score	reading score	writing score	male	group B	group C	group D	group E	bachelor's degree	high school	maste degr
0	none	72	90	74	0	1	0	0	0	1	0	
1	completed	69	90	88	0	0	1	0	0	0	0	
2	none	90	90	93	0	1	0	0	0	0	0	
3	none	47	90	44	1	0	0	0	0	0	0	
4	none	76	90	75	1	0	1	0	0	0	0	



```
In [133...] df.drop('test preparation course',axis=1,inplace=True)
```



In [134...] `df.head()`

Out[134...]

	math score	reading score	writing score	male	group B	group C	group D	group E	bachelor's degree	high school	master's degree	some college
0	72	90	74	0	1	0	0	0	1	0	0	0
1	69	90	88	0	0	1	0	0	0	0	0	1
2	90	90	93	0	1	0	0	0	0	0	1	0
3	47	90	44	1	0	0	0	0	0	0	0	0
4	76	90	75	1	0	1	0	0	0	0	0	1

In [135...] `X = df.drop('none',axis=1).values`  
`y = df['none'].values`

In [136...] `from sklearn.model_selection import train_test_split`

In [137...] `X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=4`

In [138...] `from sklearn.preprocessing import MinMaxScaler`

In [139...] `scaler = MinMaxScaler()`

In [140...] `X_train = scaler.fit_transform(X_train)`

In [141...] `X_test = scaler.transform(X_test)`

In [142...] `from tensorflow.keras.models import Sequential`  
`from tensorflow.keras.layers import Dense`

In [143...] `X_train.shape`

Out[143...] `(700, 14)`

In [144...] `model = Sequential()`  
`model.add(Dense(14,activation='relu'))`  
`model.add(Dense(7,activation='relu'))`  
`model.add(Dense(4,activation='relu'))`  
`model.add(Dense(2,activation='relu'))`  
`model.add(Dense(1,activation='sigmoid'))`  
`model.compile(optimizer='adam',loss='binary_crossentropy')`

In [145...] `model.fit(x=X_train,y=y_train,`  
`validation_data=(X_test,y_test),`  
`batch_size=128,epochs=400)`

Epoch 1/400

6/6 [=====] - 1s 28ms/step - loss: 0.6929 - val\_loss: 0.6924

```
Epoch 2/400
6/6 [=====] - 0s 5ms/step - loss: 0.6920 - val_loss: 0.6916
Epoch 3/400
6/6 [=====] - 0s 5ms/step - loss: 0.6912 - val_loss: 0.6909
Epoch 4/400
6/6 [=====] - 0s 5ms/step - loss: 0.6904 - val_loss: 0.6901
Epoch 5/400
6/6 [=====] - 0s 5ms/step - loss: 0.6895 - val_loss: 0.6894
Epoch 6/400
6/6 [=====] - 0s 5ms/step - loss: 0.6888 - val_loss: 0.6887
Epoch 7/400
6/6 [=====] - 0s 5ms/step - loss: 0.6880 - val_loss: 0.6880
Epoch 8/400
6/6 [=====] - 0s 7ms/step - loss: 0.6872 - val_loss: 0.6873
Epoch 9/400
6/6 [=====] - 0s 5ms/step - loss: 0.6864 - val_loss: 0.6866
Epoch 10/400
6/6 [=====] - 0s 5ms/step - loss: 0.6857 - val_loss: 0.6859
Epoch 11/400
6/6 [=====] - 0s 5ms/step - loss: 0.6849 - val_loss: 0.6852
Epoch 12/400
6/6 [=====] - 0s 5ms/step - loss: 0.6842 - val_loss: 0.6846
Epoch 13/400
6/6 [=====] - 0s 4ms/step - loss: 0.6835 - val_loss: 0.6839
Epoch 14/400
6/6 [=====] - 0s 5ms/step - loss: 0.6828 - val_loss: 0.6833
Epoch 15/400
6/6 [=====] - 0s 4ms/step - loss: 0.6821 - val_loss: 0.6827
Epoch 16/400
6/6 [=====] - 0s 4ms/step - loss: 0.6814 - val_loss: 0.6821
Epoch 17/400
6/6 [=====] - 0s 5ms/step - loss: 0.6807 - val_loss: 0.6815
Epoch 18/400
6/6 [=====] - 0s 5ms/step - loss: 0.6800 - val_loss: 0.6809
Epoch 19/400
6/6 [=====] - 0s 5ms/step - loss: 0.6793 - val_loss: 0.6803
Epoch 20/400
6/6 [=====] - 0s 5ms/step - loss: 0.6788 - val_loss: 0.6797
Epoch 21/400
6/6 [=====] - 0s 5ms/step - loss: 0.6781 - val_loss: 0.6792
Epoch 22/400
6/6 [=====] - 0s 5ms/step - loss: 0.6775 - val_loss: 0.6786
Epoch 23/400
6/6 [=====] - 0s 4ms/step - loss: 0.6768 - val_loss: 0.6781
Epoch 24/400
6/6 [=====] - 0s 5ms/step - loss: 0.6763 - val_loss: 0.6776
Epoch 25/400
6/6 [=====] - 0s 4ms/step - loss: 0.6757 - val_loss: 0.6771
Epoch 26/400
6/6 [=====] - 0s 5ms/step - loss: 0.6751 - val_loss: 0.6766
Epoch 27/400
6/6 [=====] - 0s 5ms/step - loss: 0.6746 - val_loss: 0.6761
Epoch 28/400
6/6 [=====] - 0s 5ms/step - loss: 0.6741 - val_loss: 0.6757
Epoch 29/400
6/6 [=====] - 0s 5ms/step - loss: 0.6735 - val_loss: 0.6752
Epoch 30/400
6/6 [=====] - 0s 5ms/step - loss: 0.6730 - val_loss: 0.6748
Epoch 31/400
6/6 [=====] - 0s 5ms/step - loss: 0.6725 - val_loss: 0.6743
Epoch 32/400
6/6 [=====] - 0s 5ms/step - loss: 0.6719 - val_loss: 0.6739
Epoch 33/400
6/6 [=====] - 0s 4ms/step - loss: 0.6715 - val_loss: 0.6735
Epoch 34/400
```

```
6/6 [=====] - 0s 5ms/step - loss: 0.6709 - val_loss: 0.6731
Epoch 35/400
6/6 [=====] - 0s 5ms/step - loss: 0.6705 - val_loss: 0.6727
Epoch 36/400
6/6 [=====] - 0s 5ms/step - loss: 0.6700 - val_loss: 0.6722
Epoch 37/400
6/6 [=====] - 0s 5ms/step - loss: 0.6695 - val_loss: 0.6718
Epoch 38/400
6/6 [=====] - 0s 5ms/step - loss: 0.6690 - val_loss: 0.6714
Epoch 39/400
6/6 [=====] - 0s 5ms/step - loss: 0.6686 - val_loss: 0.6711
Epoch 40/400
6/6 [=====] - 0s 5ms/step - loss: 0.6681 - val_loss: 0.6707
Epoch 41/400
6/6 [=====] - 0s 4ms/step - loss: 0.6676 - val_loss: 0.6702
Epoch 42/400
6/6 [=====] - 0s 5ms/step - loss: 0.6672 - val_loss: 0.6699
Epoch 43/400
6/6 [=====] - 0s 5ms/step - loss: 0.6668 - val_loss: 0.6696
Epoch 44/400
6/6 [=====] - 0s 4ms/step - loss: 0.6663 - val_loss: 0.6693
Epoch 45/400
6/6 [=====] - 0s 5ms/step - loss: 0.6658 - val_loss: 0.6689
Epoch 46/400
6/6 [=====] - 0s 5ms/step - loss: 0.6653 - val_loss: 0.6686
Epoch 47/400
6/6 [=====] - 0s 5ms/step - loss: 0.6649 - val_loss: 0.6683
Epoch 48/400
6/6 [=====] - 0s 5ms/step - loss: 0.6644 - val_loss: 0.6679
Epoch 49/400
6/6 [=====] - 0s 5ms/step - loss: 0.6640 - val_loss: 0.6676
Epoch 50/400
6/6 [=====] - 0s 4ms/step - loss: 0.6635 - val_loss: 0.6674
Epoch 51/400
6/6 [=====] - 0s 5ms/step - loss: 0.6630 - val_loss: 0.6671
Epoch 52/400
6/6 [=====] - 0s 5ms/step - loss: 0.6625 - val_loss: 0.6668
Epoch 53/400
6/6 [=====] - 0s 5ms/step - loss: 0.6620 - val_loss: 0.6665
Epoch 54/400
6/6 [=====] - 0s 5ms/step - loss: 0.6615 - val_loss: 0.6662
Epoch 55/400
6/6 [=====] - 0s 4ms/step - loss: 0.6609 - val_loss: 0.6659
Epoch 56/400
6/6 [=====] - 0s 5ms/step - loss: 0.6604 - val_loss: 0.6656
Epoch 57/400
6/6 [=====] - 0s 5ms/step - loss: 0.6599 - val_loss: 0.6655
Epoch 58/400
6/6 [=====] - 0s 5ms/step - loss: 0.6593 - val_loss: 0.6652
Epoch 59/400
6/6 [=====] - 0s 5ms/step - loss: 0.6587 - val_loss: 0.6650
Epoch 60/400
6/6 [=====] - 0s 4ms/step - loss: 0.6581 - val_loss: 0.6648
Epoch 61/400
6/6 [=====] - 0s 5ms/step - loss: 0.6574 - val_loss: 0.6646
Epoch 62/400
6/6 [=====] - 0s 5ms/step - loss: 0.6568 - val_loss: 0.6644
Epoch 63/400
6/6 [=====] - 0s 5ms/step - loss: 0.6563 - val_loss: 0.6642
Epoch 64/400
6/6 [=====] - 0s 5ms/step - loss: 0.6554 - val_loss: 0.6642
Epoch 65/400
6/6 [=====] - 0s 5ms/step - loss: 0.6549 - val_loss: 0.6638
Epoch 66/400
6/6 [=====] - 0s 6ms/step - loss: 0.6542 - val_loss: 0.6637
```

Epoch 67/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6534 - val\_loss: 0.6635  
Epoch 68/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6523 - val\_loss: 0.6632  
Epoch 69/400  
6/6 [=====] - 0s 4ms/step - loss: 0.6516 - val\_loss: 0.6631  
Epoch 70/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6506 - val\_loss: 0.6627  
Epoch 71/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6498 - val\_loss: 0.6626  
Epoch 72/400  
6/6 [=====] - 0s 4ms/step - loss: 0.6488 - val\_loss: 0.6629  
Epoch 73/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6480 - val\_loss: 0.6629  
Epoch 74/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6470 - val\_loss: 0.6627  
Epoch 75/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6461 - val\_loss: 0.6627  
Epoch 76/400  
6/6 [=====] - 0s 4ms/step - loss: 0.6453 - val\_loss: 0.6627  
Epoch 77/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6445 - val\_loss: 0.6627  
Epoch 78/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6436 - val\_loss: 0.6629  
Epoch 79/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6429 - val\_loss: 0.6629  
Epoch 80/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6417 - val\_loss: 0.6631  
Epoch 81/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6410 - val\_loss: 0.6629  
Epoch 82/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6400 - val\_loss: 0.6627  
Epoch 83/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6394 - val\_loss: 0.6629  
Epoch 84/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6383 - val\_loss: 0.6631  
Epoch 85/400  
6/6 [=====] - 0s 4ms/step - loss: 0.6374 - val\_loss: 0.6633  
Epoch 86/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6366 - val\_loss: 0.6637  
Epoch 87/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6357 - val\_loss: 0.6638  
Epoch 88/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6349 - val\_loss: 0.6641  
Epoch 89/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6342 - val\_loss: 0.6642  
Epoch 90/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6335 - val\_loss: 0.6634  
Epoch 91/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6323 - val\_loss: 0.6637  
Epoch 92/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6311 - val\_loss: 0.6635  
Epoch 93/400  
6/6 [=====] - 0s 4ms/step - loss: 0.6305 - val\_loss: 0.6636  
Epoch 94/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6294 - val\_loss: 0.6636  
Epoch 95/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6283 - val\_loss: 0.6637  
Epoch 96/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6274 - val\_loss: 0.6640  
Epoch 97/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6264 - val\_loss: 0.6641  
Epoch 98/400  
6/6 [=====] - 0s 5ms/step - loss: 0.6256 - val\_loss: 0.6644  
Epoch 99/400

```
6/6 [=====] - 0s 5ms/step - loss: 0.6247 - val_loss: 0.6644
Epoch 100/400
6/6 [=====] - 0s 5ms/step - loss: 0.6239 - val_loss: 0.6641
Epoch 101/400
6/6 [=====] - 0s 5ms/step - loss: 0.6227 - val_loss: 0.6641
Epoch 102/400
6/6 [=====] - 0s 4ms/step - loss: 0.6217 - val_loss: 0.6640
Epoch 103/400
6/6 [=====] - 0s 5ms/step - loss: 0.6211 - val_loss: 0.6641
Epoch 104/400
6/6 [=====] - 0s 5ms/step - loss: 0.6203 - val_loss: 0.6638
Epoch 105/400
6/6 [=====] - 0s 5ms/step - loss: 0.6189 - val_loss: 0.6637
Epoch 106/400
6/6 [=====] - 0s 5ms/step - loss: 0.6184 - val_loss: 0.6637
Epoch 107/400
6/6 [=====] - 0s 5ms/step - loss: 0.6174 - val_loss: 0.6632
Epoch 108/400
6/6 [=====] - 0s 5ms/step - loss: 0.6163 - val_loss: 0.6635
Epoch 109/400
6/6 [=====] - 0s 5ms/step - loss: 0.6162 - val_loss: 0.6636
Epoch 110/400
6/6 [=====] - 0s 4ms/step - loss: 0.6148 - val_loss: 0.6629
Epoch 111/400
6/6 [=====] - 0s 4ms/step - loss: 0.6146 - val_loss: 0.6639
Epoch 112/400
6/6 [=====] - 0s 5ms/step - loss: 0.6130 - val_loss: 0.6628
Epoch 113/400
6/6 [=====] - 0s 4ms/step - loss: 0.6129 - val_loss: 0.6633
Epoch 114/400
6/6 [=====] - 0s 5ms/step - loss: 0.6116 - val_loss: 0.6638
Epoch 115/400
6/6 [=====] - 0s 4ms/step - loss: 0.6108 - val_loss: 0.6632
Epoch 116/400
6/6 [=====] - 0s 5ms/step - loss: 0.6105 - val_loss: 0.6629
Epoch 117/400
6/6 [=====] - 0s 5ms/step - loss: 0.6092 - val_loss: 0.6639
Epoch 118/400
6/6 [=====] - 0s 5ms/step - loss: 0.6089 - val_loss: 0.6619
Epoch 119/400
6/6 [=====] - 0s 5ms/step - loss: 0.6079 - val_loss: 0.6638
Epoch 120/400
6/6 [=====] - 0s 5ms/step - loss: 0.6072 - val_loss: 0.6630
Epoch 121/400
6/6 [=====] - 0s 5ms/step - loss: 0.6062 - val_loss: 0.6638
Epoch 122/400
6/6 [=====] - 0s 5ms/step - loss: 0.6055 - val_loss: 0.6644
Epoch 123/400
6/6 [=====] - 0s 5ms/step - loss: 0.6047 - val_loss: 0.6651
Epoch 124/400
6/6 [=====] - 0s 5ms/step - loss: 0.6045 - val_loss: 0.6654
Epoch 125/400
6/6 [=====] - 0s 5ms/step - loss: 0.6036 - val_loss: 0.6643
Epoch 126/400
6/6 [=====] - 0s 5ms/step - loss: 0.6032 - val_loss: 0.6661
Epoch 127/400
6/6 [=====] - 0s 5ms/step - loss: 0.6027 - val_loss: 0.6659
Epoch 128/400
6/6 [=====] - 0s 5ms/step - loss: 0.6018 - val_loss: 0.6666
Epoch 129/400
6/6 [=====] - 0s 5ms/step - loss: 0.6016 - val_loss: 0.6662
Epoch 130/400
6/6 [=====] - 0s 4ms/step - loss: 0.6007 - val_loss: 0.6661
Epoch 131/400
6/6 [=====] - 0s 5ms/step - loss: 0.6004 - val_loss: 0.6665
```

Epoch 132/400  
6/6 [=====] - 0s 4ms/step - loss: 0.6001 - val\_loss: 0.6661  
Epoch 133/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5997 - val\_loss: 0.6676  
Epoch 134/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5988 - val\_loss: 0.6666  
Epoch 135/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5982 - val\_loss: 0.6665  
Epoch 136/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5977 - val\_loss: 0.6660  
Epoch 137/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5974 - val\_loss: 0.6664  
Epoch 138/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5974 - val\_loss: 0.6654  
Epoch 139/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5967 - val\_loss: 0.6683  
Epoch 140/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5957 - val\_loss: 0.6665  
Epoch 141/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5952 - val\_loss: 0.6682  
Epoch 142/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5949 - val\_loss: 0.6679  
Epoch 143/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5943 - val\_loss: 0.6684  
Epoch 144/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5937 - val\_loss: 0.6695  
Epoch 145/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5936 - val\_loss: 0.6688  
Epoch 146/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5930 - val\_loss: 0.6681  
Epoch 147/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5925 - val\_loss: 0.6697  
Epoch 148/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5926 - val\_loss: 0.6672  
Epoch 149/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5915 - val\_loss: 0.6681  
Epoch 150/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5915 - val\_loss: 0.6687  
Epoch 151/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5911 - val\_loss: 0.6670  
Epoch 152/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5904 - val\_loss: 0.6680  
Epoch 153/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5899 - val\_loss: 0.6686  
Epoch 154/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5894 - val\_loss: 0.6680  
Epoch 155/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5890 - val\_loss: 0.6682  
Epoch 156/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5885 - val\_loss: 0.6689  
Epoch 157/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5888 - val\_loss: 0.6692  
Epoch 158/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5881 - val\_loss: 0.6669  
Epoch 159/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5884 - val\_loss: 0.6679  
Epoch 160/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5880 - val\_loss: 0.6701  
Epoch 161/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5874 - val\_loss: 0.6656  
Epoch 162/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5867 - val\_loss: 0.6687  
Epoch 163/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5865 - val\_loss: 0.6682  
Epoch 164/400

```
6/6 [=====] - 0s 5ms/step - loss: 0.5861 - val_loss: 0.6657
Epoch 165/400
6/6 [=====] - 0s 5ms/step - loss: 0.5855 - val_loss: 0.6687
Epoch 166/400
6/6 [=====] - 0s 5ms/step - loss: 0.5855 - val_loss: 0.6685
Epoch 167/400
6/6 [=====] - 0s 5ms/step - loss: 0.5848 - val_loss: 0.6674
Epoch 168/400
6/6 [=====] - 0s 5ms/step - loss: 0.5847 - val_loss: 0.6670
Epoch 169/400
6/6 [=====] - 0s 4ms/step - loss: 0.5843 - val_loss: 0.6689
Epoch 170/400
6/6 [=====] - 0s 5ms/step - loss: 0.5842 - val_loss: 0.6691
Epoch 171/400
6/6 [=====] - 0s 5ms/step - loss: 0.5839 - val_loss: 0.6681
Epoch 172/400
6/6 [=====] - 0s 4ms/step - loss: 0.5835 - val_loss: 0.6687
Epoch 173/400
6/6 [=====] - 0s 5ms/step - loss: 0.5835 - val_loss: 0.6688
Epoch 174/400
6/6 [=====] - 0s 4ms/step - loss: 0.5825 - val_loss: 0.6691
Epoch 175/400
6/6 [=====] - 0s 5ms/step - loss: 0.5823 - val_loss: 0.6678
Epoch 176/400
6/6 [=====] - 0s 5ms/step - loss: 0.5821 - val_loss: 0.6693
Epoch 177/400
6/6 [=====] - 0s 5ms/step - loss: 0.5815 - val_loss: 0.6675
Epoch 178/400
6/6 [=====] - 0s 5ms/step - loss: 0.5813 - val_loss: 0.6681
Epoch 179/400
6/6 [=====] - 0s 5ms/step - loss: 0.5812 - val_loss: 0.6688
Epoch 180/400
6/6 [=====] - 0s 5ms/step - loss: 0.5807 - val_loss: 0.6696
Epoch 181/400
6/6 [=====] - 0s 5ms/step - loss: 0.5802 - val_loss: 0.6690
Epoch 182/400
6/6 [=====] - 0s 5ms/step - loss: 0.5800 - val_loss: 0.6699
Epoch 183/400
6/6 [=====] - 0s 5ms/step - loss: 0.5798 - val_loss: 0.6687
Epoch 184/400
6/6 [=====] - 0s 4ms/step - loss: 0.5800 - val_loss: 0.6686
Epoch 185/400
6/6 [=====] - 0s 5ms/step - loss: 0.5799 - val_loss: 0.6716
Epoch 186/400
6/6 [=====] - 0s 4ms/step - loss: 0.5787 - val_loss: 0.6677
Epoch 187/400
6/6 [=====] - 0s 4ms/step - loss: 0.5788 - val_loss: 0.6691
Epoch 188/400
6/6 [=====] - 0s 5ms/step - loss: 0.5778 - val_loss: 0.6718
Epoch 189/400
6/6 [=====] - 0s 5ms/step - loss: 0.5780 - val_loss: 0.6705
Epoch 190/400
6/6 [=====] - 0s 4ms/step - loss: 0.5772 - val_loss: 0.6698
Epoch 191/400
6/6 [=====] - 0s 5ms/step - loss: 0.5771 - val_loss: 0.6704
Epoch 192/400
6/6 [=====] - 0s 4ms/step - loss: 0.5763 - val_loss: 0.6698
Epoch 193/400
6/6 [=====] - 0s 5ms/step - loss: 0.5758 - val_loss: 0.6699
Epoch 194/400
6/6 [=====] - 0s 5ms/step - loss: 0.5753 - val_loss: 0.6686
Epoch 195/400
6/6 [=====] - 0s 5ms/step - loss: 0.5753 - val_loss: 0.6682
Epoch 196/400
6/6 [=====] - 0s 5ms/step - loss: 0.5764 - val_loss: 0.6662
```

Epoch 197/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5743 - val\_loss: 0.6692  
Epoch 198/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5742 - val\_loss: 0.6683  
Epoch 199/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5738 - val\_loss: 0.6676  
Epoch 200/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5730 - val\_loss: 0.6691  
Epoch 201/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5727 - val\_loss: 0.6671  
Epoch 202/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5722 - val\_loss: 0.6685  
Epoch 203/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5720 - val\_loss: 0.6680  
Epoch 204/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5719 - val\_loss: 0.6678  
Epoch 205/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5716 - val\_loss: 0.6663  
Epoch 206/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5705 - val\_loss: 0.6687  
Epoch 207/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5706 - val\_loss: 0.6697  
Epoch 208/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5706 - val\_loss: 0.6683  
Epoch 209/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5700 - val\_loss: 0.6653  
Epoch 210/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5705 - val\_loss: 0.6695  
Epoch 211/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5688 - val\_loss: 0.6679  
Epoch 212/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5686 - val\_loss: 0.6682  
Epoch 213/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5686 - val\_loss: 0.6683  
Epoch 214/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5679 - val\_loss: 0.6650  
Epoch 215/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5688 - val\_loss: 0.6672  
Epoch 216/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5671 - val\_loss: 0.6672  
Epoch 217/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5677 - val\_loss: 0.6660  
Epoch 218/400  
6/6 [=====] - 0s 6ms/step - loss: 0.5673 - val\_loss: 0.6703  
Epoch 219/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5668 - val\_loss: 0.6679  
Epoch 220/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5662 - val\_loss: 0.6693  
Epoch 221/400  
6/6 [=====] - 0s 7ms/step - loss: 0.5656 - val\_loss: 0.6693  
Epoch 222/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5660 - val\_loss: 0.6693  
Epoch 223/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5661 - val\_loss: 0.6722  
Epoch 224/400  
6/6 [=====] - 0s 6ms/step - loss: 0.5679 - val\_loss: 0.6666  
Epoch 225/400  
6/6 [=====] - 0s 6ms/step - loss: 0.5657 - val\_loss: 0.6688  
Epoch 226/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5670 - val\_loss: 0.6723  
Epoch 227/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5632 - val\_loss: 0.6674  
Epoch 228/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5654 - val\_loss: 0.6670  
Epoch 229/400



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6/6 [=====] - 0s 4ms/step - loss: 0.5655 - val_loss: 0.6718
Epoch 230/400
6/6 [=====] - 0s 5ms/step - loss: 0.5641 - val_loss: 0.6683
Epoch 231/400
6/6 [=====] - 0s 5ms/step - loss: 0.5628 - val_loss: 0.6694
Epoch 232/400
6/6 [=====] - 0s 5ms/step - loss: 0.5628 - val_loss: 0.6703
Epoch 233/400
6/6 [=====] - 0s 4ms/step - loss: 0.5623 - val_loss: 0.6692
Epoch 234/400
6/6 [=====] - 0s 4ms/step - loss: 0.5627 - val_loss: 0.6707
Epoch 235/400
6/6 [=====] - 0s 5ms/step - loss: 0.5617 - val_loss: 0.6701
Epoch 236/400
6/6 [=====] - 0s 5ms/step - loss: 0.5615 - val_loss: 0.6699
Epoch 237/400
6/6 [=====] - 0s 5ms/step - loss: 0.5615 - val_loss: 0.6694
Epoch 238/400
6/6 [=====] - 0s 5ms/step - loss: 0.5609 - val_loss: 0.6701
Epoch 239/400
6/6 [=====] - 0s 5ms/step - loss: 0.5608 - val_loss: 0.6710
Epoch 240/400
6/6 [=====] - 0s 5ms/step - loss: 0.5612 - val_loss: 0.6716
Epoch 241/400
6/6 [=====] - 0s 5ms/step - loss: 0.5603 - val_loss: 0.6718
Epoch 242/400
6/6 [=====] - 0s 6ms/step - loss: 0.5595 - val_loss: 0.6709
Epoch 243/400
6/6 [=====] - 0s 5ms/step - loss: 0.5602 - val_loss: 0.6692
Epoch 244/400
6/6 [=====] - 0s 5ms/step - loss: 0.5598 - val_loss: 0.6698
Epoch 245/400
6/6 [=====] - 0s 4ms/step - loss: 0.5595 - val_loss: 0.6716
Epoch 246/400
6/6 [=====] - 0s 5ms/step - loss: 0.5604 - val_loss: 0.6732
Epoch 247/400
6/6 [=====] - 0s 4ms/step - loss: 0.5588 - val_loss: 0.6697
Epoch 248/400
6/6 [=====] - 0s 5ms/step - loss: 0.5596 - val_loss: 0.6685
Epoch 249/400
6/6 [=====] - 0s 4ms/step - loss: 0.5594 - val_loss: 0.6710
Epoch 250/400
6/6 [=====] - 0s 5ms/step - loss: 0.5605 - val_loss: 0.6713
Epoch 251/400
6/6 [=====] - 0s 5ms/step - loss: 0.5600 - val_loss: 0.6683
Epoch 252/400
6/6 [=====] - 0s 5ms/step - loss: 0.5582 - val_loss: 0.6717
Epoch 253/400
6/6 [=====] - 0s 5ms/step - loss: 0.5580 - val_loss: 0.6710
Epoch 254/400
6/6 [=====] - 0s 5ms/step - loss: 0.5572 - val_loss: 0.6683
Epoch 255/400
6/6 [=====] - 0s 5ms/step - loss: 0.5579 - val_loss: 0.6706
Epoch 256/400
6/6 [=====] - 0s 5ms/step - loss: 0.5567 - val_loss: 0.6689
Epoch 257/400
6/6 [=====] - 0s 5ms/step - loss: 0.5563 - val_loss: 0.6683
Epoch 258/400
6/6 [=====] - 0s 5ms/step - loss: 0.5565 - val_loss: 0.6683
Epoch 259/400
6/6 [=====] - 0s 5ms/step - loss: 0.5558 - val_loss: 0.6676
Epoch 260/400
6/6 [=====] - 0s 5ms/step - loss: 0.5554 - val_loss: 0.6694
Epoch 261/400
6/6 [=====] - 0s 5ms/step - loss: 0.5556 - val_loss: 0.6684
```

Epoch 262/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5550 - val\_loss: 0.6670  
Epoch 263/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5547 - val\_loss: 0.6666  
Epoch 264/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5545 - val\_loss: 0.6667  
Epoch 265/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5552 - val\_loss: 0.6642  
Epoch 266/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5541 - val\_loss: 0.6691  
Epoch 267/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5534 - val\_loss: 0.6653  
Epoch 268/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5531 - val\_loss: 0.6625  
Epoch 269/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5526 - val\_loss: 0.6653  
Epoch 270/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5520 - val\_loss: 0.6679  
Epoch 271/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5513 - val\_loss: 0.6634  
Epoch 272/400  
6/6 [=====] - 0s 7ms/step - loss: 0.5529 - val\_loss: 0.6621  
Epoch 273/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5537 - val\_loss: 0.6681  
Epoch 274/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5523 - val\_loss: 0.6615  
Epoch 275/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5511 - val\_loss: 0.6651  
Epoch 276/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5512 - val\_loss: 0.6669  
Epoch 277/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5505 - val\_loss: 0.6618  
Epoch 278/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5499 - val\_loss: 0.6640  
Epoch 279/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5493 - val\_loss: 0.6636  
Epoch 280/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5496 - val\_loss: 0.6619  
Epoch 281/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5500 - val\_loss: 0.6647  
Epoch 282/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5477 - val\_loss: 0.6592  
Epoch 283/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5474 - val\_loss: 0.6601  
Epoch 284/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5472 - val\_loss: 0.6609  
Epoch 285/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5474 - val\_loss: 0.6594  
Epoch 286/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5462 - val\_loss: 0.6572  
Epoch 287/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5462 - val\_loss: 0.6601  
Epoch 288/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5459 - val\_loss: 0.6580  
Epoch 289/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5450 - val\_loss: 0.6590  
Epoch 290/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5452 - val\_loss: 0.6588  
Epoch 291/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5451 - val\_loss: 0.6604  
Epoch 292/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5442 - val\_loss: 0.6585  
Epoch 293/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5446 - val\_loss: 0.6596  
Epoch 294/400

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6/6 [=====] - 0s 4ms/step - loss: 0.5436 - val_loss: 0.6594
Epoch 295/400
6/6 [=====] - 0s 5ms/step - loss: 0.5452 - val_loss: 0.6605
Epoch 296/400
6/6 [=====] - 0s 5ms/step - loss: 0.5438 - val_loss: 0.6555
Epoch 297/400
6/6 [=====] - 0s 5ms/step - loss: 0.5440 - val_loss: 0.6571
Epoch 298/400
6/6 [=====] - 0s 5ms/step - loss: 0.5425 - val_loss: 0.6584
Epoch 299/400
6/6 [=====] - 0s 5ms/step - loss: 0.5419 - val_loss: 0.6556
Epoch 300/400
6/6 [=====] - 0s 5ms/step - loss: 0.5415 - val_loss: 0.6560
Epoch 301/400
6/6 [=====] - 0s 5ms/step - loss: 0.5411 - val_loss: 0.6551
Epoch 302/400
6/6 [=====] - 0s 5ms/step - loss: 0.5401 - val_loss: 0.6553
Epoch 303/400
6/6 [=====] - 0s 5ms/step - loss: 0.5397 - val_loss: 0.6562
Epoch 304/400
6/6 [=====] - 0s 5ms/step - loss: 0.5399 - val_loss: 0.6552
Epoch 305/400
6/6 [=====] - 0s 5ms/step - loss: 0.5381 - val_loss: 0.6533
Epoch 306/400
6/6 [=====] - 0s 5ms/step - loss: 0.5397 - val_loss: 0.6553
Epoch 307/400
6/6 [=====] - 0s 5ms/step - loss: 0.5387 - val_loss: 0.6593
Epoch 308/400
6/6 [=====] - 0s 4ms/step - loss: 0.5376 - val_loss: 0.6531
Epoch 309/400
6/6 [=====] - 0s 4ms/step - loss: 0.5393 - val_loss: 0.6550
Epoch 310/400
6/6 [=====] - 0s 5ms/step - loss: 0.5368 - val_loss: 0.6530
Epoch 311/400
6/6 [=====] - 0s 5ms/step - loss: 0.5375 - val_loss: 0.6544
Epoch 312/400
6/6 [=====] - 0s 4ms/step - loss: 0.5356 - val_loss: 0.6540
Epoch 313/400
6/6 [=====] - 0s 5ms/step - loss: 0.5349 - val_loss: 0.6529
Epoch 314/400
6/6 [=====] - 0s 5ms/step - loss: 0.5353 - val_loss: 0.6541
Epoch 315/400
6/6 [=====] - 0s 5ms/step - loss: 0.5353 - val_loss: 0.6567
Epoch 316/400
6/6 [=====] - 0s 4ms/step - loss: 0.5353 - val_loss: 0.6525
Epoch 317/400
6/6 [=====] - 0s 5ms/step - loss: 0.5338 - val_loss: 0.6554
Epoch 318/400
6/6 [=====] - 0s 5ms/step - loss: 0.5337 - val_loss: 0.6545
Epoch 319/400
6/6 [=====] - 0s 5ms/step - loss: 0.5334 - val_loss: 0.6536
Epoch 320/400
6/6 [=====] - 0s 5ms/step - loss: 0.5333 - val_loss: 0.6557
Epoch 321/400
6/6 [=====] - 0s 4ms/step - loss: 0.5328 - val_loss: 0.6574
Epoch 322/400
6/6 [=====] - 0s 5ms/step - loss: 0.5338 - val_loss: 0.6544
Epoch 323/400
6/6 [=====] - 0s 5ms/step - loss: 0.5339 - val_loss: 0.6525
Epoch 324/400
6/6 [=====] - 0s 7ms/step - loss: 0.5321 - val_loss: 0.6554
Epoch 325/400
6/6 [=====] - 0s 5ms/step - loss: 0.5329 - val_loss: 0.6541
Epoch 326/400
6/6 [=====] - 0s 5ms/step - loss: 0.5318 - val_loss: 0.6533
```

Epoch 327/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5323 - val\_loss: 0.6532  
Epoch 328/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5310 - val\_loss: 0.6542  
Epoch 329/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5310 - val\_loss: 0.6540  
Epoch 330/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5311 - val\_loss: 0.6533  
Epoch 331/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5311 - val\_loss: 0.6559  
Epoch 332/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5302 - val\_loss: 0.6542  
Epoch 333/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5304 - val\_loss: 0.6558  
Epoch 334/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5294 - val\_loss: 0.6537  
Epoch 335/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5301 - val\_loss: 0.6526  
Epoch 336/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5299 - val\_loss: 0.6528  
Epoch 337/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5289 - val\_loss: 0.6509  
Epoch 338/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5288 - val\_loss: 0.6541  
Epoch 339/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5292 - val\_loss: 0.6529  
Epoch 340/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5282 - val\_loss: 0.6548  
Epoch 341/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5283 - val\_loss: 0.6556  
Epoch 342/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5282 - val\_loss: 0.6560  
Epoch 343/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5272 - val\_loss: 0.6527  
Epoch 344/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5285 - val\_loss: 0.6551  
Epoch 345/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5278 - val\_loss: 0.6543  
Epoch 346/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5278 - val\_loss: 0.6526  
Epoch 347/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5272 - val\_loss: 0.6569  
Epoch 348/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5274 - val\_loss: 0.6544  
Epoch 349/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5265 - val\_loss: 0.6557  
Epoch 350/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5270 - val\_loss: 0.6548  
Epoch 351/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5266 - val\_loss: 0.6534  
Epoch 352/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5265 - val\_loss: 0.6567  
Epoch 353/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5263 - val\_loss: 0.6561  
Epoch 354/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5267 - val\_loss: 0.6553  
Epoch 355/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5261 - val\_loss: 0.6588  
Epoch 356/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5265 - val\_loss: 0.6559  
Epoch 357/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5244 - val\_loss: 0.6588  
Epoch 358/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5255 - val\_loss: 0.6568  
Epoch 359/400

6/6 [=====] - 0s 5ms/step - loss: 0.5251 - val\_loss: 0.6558  
Epoch 360/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5246 - val\_loss: 0.6578  
Epoch 361/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5264 - val\_loss: 0.6604  
Epoch 362/400  
6/6 [=====] - 0s 2ms/step - loss: 0.5251 - val\_loss: 0.6547  
Epoch 363/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5261 - val\_loss: 0.6567  
Epoch 364/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5253 - val\_loss: 0.6581  
Epoch 365/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5239 - val\_loss: 0.6547  
Epoch 366/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5239 - val\_loss: 0.6575  
Epoch 367/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5238 - val\_loss: 0.6544  
Epoch 368/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5233 - val\_loss: 0.6558  
Epoch 369/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5229 - val\_loss: 0.6566  
Epoch 370/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5226 - val\_loss: 0.6583  
Epoch 371/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5246 - val\_loss: 0.6590  
Epoch 372/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5238 - val\_loss: 0.6553  
Epoch 373/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5240 - val\_loss: 0.6596  
Epoch 374/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5231 - val\_loss: 0.6593  
Epoch 375/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5222 - val\_loss: 0.6551  
Epoch 376/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5233 - val\_loss: 0.6576  
Epoch 377/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5242 - val\_loss: 0.6609  
Epoch 378/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5237 - val\_loss: 0.6561  
Epoch 379/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5205 - val\_loss: 0.6625  
Epoch 380/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5221 - val\_loss: 0.6577  
Epoch 381/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5207 - val\_loss: 0.6570  
Epoch 382/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5209 - val\_loss: 0.6586  
Epoch 383/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5205 - val\_loss: 0.6593  
Epoch 384/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5201 - val\_loss: 0.6599  
Epoch 385/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5198 - val\_loss: 0.6600  
Epoch 386/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5197 - val\_loss: 0.6606  
Epoch 387/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5209 - val\_loss: 0.6626  
Epoch 388/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5201 - val\_loss: 0.6583  
Epoch 389/400  
6/6 [=====] - 0s 5ms/step - loss: 0.5193 - val\_loss: 0.6618  
Epoch 390/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5200 - val\_loss: 0.6588  
Epoch 391/400  
6/6 [=====] - 0s 4ms/step - loss: 0.5185 - val\_loss: 0.6606

```

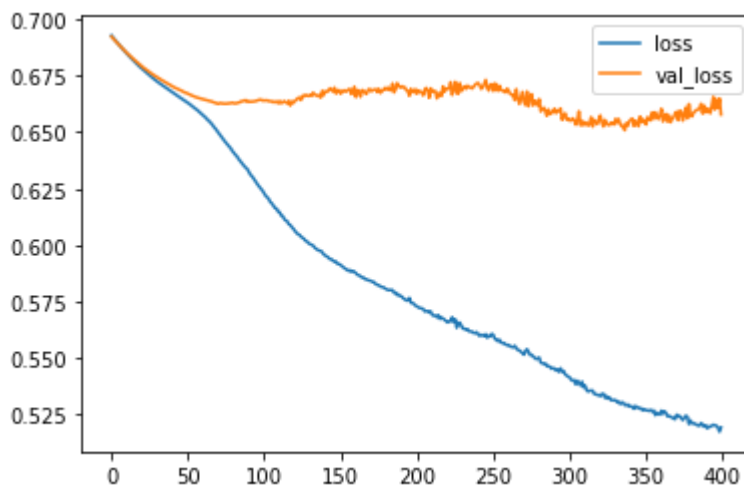
Epoch 392/400
6/6 [=====] - 0s 5ms/step - loss: 0.5193 - val_loss: 0.6598
Epoch 393/400
6/6 [=====] - 0s 5ms/step - loss: 0.5193 - val_loss: 0.6584
Epoch 394/400
6/6 [=====] - 0s 5ms/step - loss: 0.5197 - val_loss: 0.6618
Epoch 395/400
6/6 [=====] - 0s 5ms/step - loss: 0.5201 - val_loss: 0.6657
Epoch 396/400
6/6 [=====] - 0s 5ms/step - loss: 0.5199 - val_loss: 0.6601
Epoch 397/400
6/6 [=====] - 0s 4ms/step - loss: 0.5197 - val_loss: 0.6644
Epoch 398/400
6/6 [=====] - 0s 4ms/step - loss: 0.5189 - val_loss: 0.6608
Epoch 399/400
6/6 [=====] - 0s 5ms/step - loss: 0.5171 - val_loss: 0.6649
Epoch 400/400
6/6 [=====] - 0s 5ms/step - loss: 0.5189 - val_loss: 0.6579

```

Out[145... <keras.callbacks.History at 0x2213583d820>

```
In [146... model_loss = pd.DataFrame(model.history.history)
model_loss.plot()
```

Out[146... <AxesSubplot:>



```
In [147... from tensorflow.keras.callbacks import EarlyStopping
```

```
In [151... from tensorflow.keras.layers import Dropout
```

```
In [152... model = Sequential()

model.add(Dense(14,activation='relu'))
model.add(Dropout(0.5))

model.add(Dense(7,activation='relu'))
model.add(Dropout(0.5))

model.add(Dense(4,activation='relu'))
model.add(Dropout(0.5))

model.add(Dense(2,activation='relu'))
model.add(Dropout(0.5))

model.add(Dense(1,activation='sigmoid'))
```

```
model.compile(loss='binary_crossentropy',optimizer='adam')
```

In [153...

```
model.fit(x=X_train,y=y_train,epochs=600,validation_data=(X_test,y_test),  
         callbacks=[early_stop])
```

```
Epoch 1/600  
22/22 [=====] - 1s 5ms/step - loss: 1.0873 - val_loss: 0.7530  
Epoch 2/600  
22/22 [=====] - 0s 2ms/step - loss: 0.9492 - val_loss: 0.7087  
Epoch 3/600  
22/22 [=====] - 0s 2ms/step - loss: 0.8315 - val_loss: 0.6923  
Epoch 4/600  
22/22 [=====] - 0s 2ms/step - loss: 0.8065 - val_loss: 0.6843  
Epoch 5/600  
22/22 [=====] - 0s 2ms/step - loss: 0.7259 - val_loss: 0.6816  
Epoch 6/600  
22/22 [=====] - 0s 2ms/step - loss: 0.7245 - val_loss: 0.6801  
Epoch 7/600  
22/22 [=====] - 0s 3ms/step - loss: 0.7213 - val_loss: 0.6787  
Epoch 8/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6949 - val_loss: 0.6771  
Epoch 9/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6846 - val_loss: 0.6756  
Epoch 10/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6872 - val_loss: 0.6744  
Epoch 11/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6794 - val_loss: 0.6730  
Epoch 12/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6826 - val_loss: 0.6719  
Epoch 13/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6792 - val_loss: 0.6708  
Epoch 14/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6769 - val_loss: 0.6697  
Epoch 15/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6610 - val_loss: 0.6687  
Epoch 16/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6743 - val_loss: 0.6678  
Epoch 17/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6629 - val_loss: 0.6669  
Epoch 18/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6713 - val_loss: 0.6662  
Epoch 19/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6659 - val_loss: 0.6654  
Epoch 20/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6623 - val_loss: 0.6647  
Epoch 21/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6627 - val_loss: 0.6641  
Epoch 22/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6633 - val_loss: 0.6635  
Epoch 23/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6614 - val_loss: 0.6628  
Epoch 24/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6605 - val_loss: 0.6624  
Epoch 25/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6591 - val_loss: 0.6619  
Epoch 26/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6562 - val_loss: 0.6614  
Epoch 27/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6568 - val_loss: 0.6610  
Epoch 28/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6586 - val_loss: 0.6606  
Epoch 29/600  
22/22 [=====] - 0s 2ms/step - loss: 0.6568 - val_loss: 0.6603
```

```
Epoch 30/600
22/22 [=====] - 0s 2ms/step - loss: 0.6552 - val_loss: 0.6600
Epoch 31/600
22/22 [=====] - 0s 2ms/step - loss: 0.6567 - val_loss: 0.6597
Epoch 32/600
22/22 [=====] - 0s 2ms/step - loss: 0.6544 - val_loss: 0.6594
Epoch 33/600
22/22 [=====] - 0s 2ms/step - loss: 0.6553 - val_loss: 0.6592
Epoch 34/600
22/22 [=====] - 0s 2ms/step - loss: 0.6570 - val_loss: 0.6589
Epoch 35/600
22/22 [=====] - 0s 2ms/step - loss: 0.6571 - val_loss: 0.6587
Epoch 36/600
22/22 [=====] - 0s 2ms/step - loss: 0.6536 - val_loss: 0.6586
Epoch 37/600
22/22 [=====] - 0s 2ms/step - loss: 0.6544 - val_loss: 0.6584
Epoch 38/600
22/22 [=====] - 0s 2ms/step - loss: 0.6528 - val_loss: 0.6583
Epoch 39/600
22/22 [=====] - 0s 2ms/step - loss: 0.6517 - val_loss: 0.6581
Epoch 40/600
22/22 [=====] - 0s 2ms/step - loss: 0.6571 - val_loss: 0.6580
Epoch 41/600
22/22 [=====] - 0s 2ms/step - loss: 0.6519 - val_loss: 0.6579
Epoch 42/600
22/22 [=====] - 0s 2ms/step - loss: 0.6526 - val_loss: 0.6578
Epoch 43/600
22/22 [=====] - 0s 2ms/step - loss: 0.6549 - val_loss: 0.6577
Epoch 44/600
22/22 [=====] - 0s 2ms/step - loss: 0.6509 - val_loss: 0.6576
Epoch 45/600
22/22 [=====] - 0s 2ms/step - loss: 0.6513 - val_loss: 0.6575
Epoch 46/600
22/22 [=====] - 0s 2ms/step - loss: 0.6513 - val_loss: 0.6575
Epoch 47/600
22/22 [=====] - 0s 2ms/step - loss: 0.6506 - val_loss: 0.6574
Epoch 48/600
22/22 [=====] - 0s 2ms/step - loss: 0.6513 - val_loss: 0.6573
Epoch 49/600
22/22 [=====] - 0s 2ms/step - loss: 0.6509 - val_loss: 0.6573
Epoch 50/600
22/22 [=====] - 0s 2ms/step - loss: 0.6532 - val_loss: 0.6573
Epoch 51/600
22/22 [=====] - 0s 2ms/step - loss: 0.6504 - val_loss: 0.6573
Epoch 52/600
22/22 [=====] - 0s 2ms/step - loss: 0.6518 - val_loss: 0.6572
Epoch 53/600
22/22 [=====] - 0s 2ms/step - loss: 0.6519 - val_loss: 0.6572
Epoch 54/600
22/22 [=====] - 0s 2ms/step - loss: 0.6505 - val_loss: 0.6572
Epoch 55/600
22/22 [=====] - 0s 2ms/step - loss: 0.6490 - val_loss: 0.6572
Epoch 56/600
22/22 [=====] - 0s 2ms/step - loss: 0.6501 - val_loss: 0.6572
Epoch 57/600
22/22 [=====] - 0s 2ms/step - loss: 0.6574 - val_loss: 0.6572
Epoch 58/600
22/22 [=====] - 0s 2ms/step - loss: 0.6505 - val_loss: 0.6572
Epoch 59/600
22/22 [=====] - 0s 2ms/step - loss: 0.6508 - val_loss: 0.6572
Epoch 60/600
22/22 [=====] - 0s 2ms/step - loss: 0.6506 - val_loss: 0.6572
Epoch 61/600
22/22 [=====] - 0s 2ms/step - loss: 0.6518 - val_loss: 0.6572
Epoch 62/600
```



```

22/22 [=====] - 0s 2ms/step - loss: 0.6509 - val_loss: 0.6572
Epoch 63/600
22/22 [=====] - 0s 2ms/step - loss: 0.6499 - val_loss: 0.6572
Epoch 64/600
22/22 [=====] - 0s 2ms/step - loss: 0.6487 - val_loss: 0.6572
Epoch 65/600
22/22 [=====] - 0s 2ms/step - loss: 0.6502 - val_loss: 0.6572
Epoch 66/600
22/22 [=====] - 0s 2ms/step - loss: 0.6491 - val_loss: 0.6572
Epoch 67/600
22/22 [=====] - 0s 2ms/step - loss: 0.6490 - val_loss: 0.6572
Epoch 68/600
22/22 [=====] - 0s 2ms/step - loss: 0.6508 - val_loss: 0.6572
Epoch 69/600
22/22 [=====] - 0s 2ms/step - loss: 0.6495 - val_loss: 0.6572
Epoch 70/600
22/22 [=====] - 0s 2ms/step - loss: 0.6490 - val_loss: 0.6572
Epoch 71/600
22/22 [=====] - 0s 2ms/step - loss: 0.6520 - val_loss: 0.6572
Epoch 72/600
22/22 [=====] - 0s 2ms/step - loss: 0.6497 - val_loss: 0.6572
Epoch 73/600
22/22 [=====] - 0s 2ms/step - loss: 0.6486 - val_loss: 0.6572
Epoch 74/600
22/22 [=====] - 0s 2ms/step - loss: 0.6502 - val_loss: 0.6573
Epoch 75/600
22/22 [=====] - 0s 2ms/step - loss: 0.6503 - val_loss: 0.6573
Epoch 76/600
22/22 [=====] - 0s 2ms/step - loss: 0.6495 - val_loss: 0.6573
Epoch 77/600
22/22 [=====] - 0s 2ms/step - loss: 0.6504 - val_loss: 0.6573
Epoch 78/600
22/22 [=====] - 0s 2ms/step - loss: 0.6512 - val_loss: 0.6573
Epoch 79/600
22/22 [=====] - 0s 2ms/step - loss: 0.6501 - val_loss: 0.6573
Epoch 80/600
22/22 [=====] - 0s 2ms/step - loss: 0.6488 - val_loss: 0.6573
Epoch 81/600
22/22 [=====] - 0s 2ms/step - loss: 0.6496 - val_loss: 0.6573
Epoch 82/600
22/22 [=====] - 0s 2ms/step - loss: 0.6500 - val_loss: 0.6573
Epoch 83/600
22/22 [=====] - 0s 2ms/step - loss: 0.6500 - val_loss: 0.6574
Epoch 84/600
22/22 [=====] - 0s 2ms/step - loss: 0.6482 - val_loss: 0.6573
Epoch 00084: early stopping

```

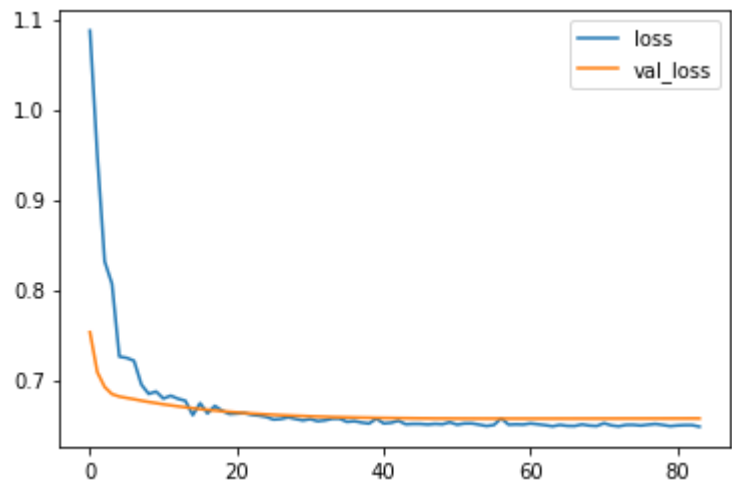
Out[153... <keras.callbacks.History at 0x22137a23730>

```

In [154... model_loss = pd.DataFrame(model.history.history)
          model_loss.plot()

```

Out[154... <AxesSubplot:>



In [ ]: