```
In [58]:
           import pandas as pd
           import numpy as np
           import seaborn as sns
           df = pd.read_csv('heart_failure_clinical_records_dataset.csv')
In [59]:
           df.head()
In [60]:
Out[60]:
                  anaemia
                            creatinine_phosphokinase diabetes ejection_fraction high_blood_pressure
                                                                                                    platelets
              75.0
                         0
                                                582
                                                            0
                                                                           20
                                                                                                   265000.00
              55.0
                         0
                                               7861
                                                            0
                                                                           38
                                                                                                   263358.03
           1
                         0
                                                146
                                                            0
                                                                           20
                                                                                                   162000.00
           2
              65.0
              50.0
                         1
                                                111
                                                            0
                                                                           20
                                                                                                   210000.00
              65.0
                         1
                                                160
                                                            1
                                                                           20
                                                                                                   327000.00
           sns.scatterplot(x='age',y='platelets', data=df)
In [61]:
          <AxesSubplot:xlabel='age', ylabel='platelets'>
             800000
             600000
             400000
             200000
                  0
                              50
                                                70
                                       60
                                                        80
                                             age
In [62]:
           df['age'].value_counts()
          60.000
                      33
Out[62]:
          50.000
                      27
          65.000
                      26
          70.000
                      25
          45.000
                      19
                      17
          55.000
          75.000
                      11
          58.000
                      10
          53.000
                      10
                       8
          63.000
          80.000
                       7
          72.000
                       7
          42.000
                       7
                       7
          40.000
          85.000
                       6
```

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62.000 5 68.000 5 52.000 5 49.000 4 61.000 4 51.000 4 4 73.000 59.000 4 46.000 3 3 64.000 3 82.000 3 90.000 69.000 3 95.000 2 2 2 2 48.000 60.667 67.000 44.000 2 2 66.000 2 77.000 78.000 2 54.000 2 2 57.000 1 94.000 56.000 1 41.000 1 87.000 1 79.000 1 86.000 1 43.000 1 81.000 47.000 1

Name: age, dtype: int64

In [63]:

df.corr()

| \cap | 111 | \vdash | Г | L | 2 | 1 | 0 |
|--------|-----|----------|---|---|-----------|---|---|
| U | и | _ | L | U | $_{\sim}$ | J | ۰ |

| | age | anaemia | $creatinine_phosphokinase$ | diabetes | ejection_fraction hi |
|--------------------------|-----------|-----------|-----------------------------|-----------|----------------------|
| age | 1.000000 | 0.088006 | -0.081584 | -0.101012 | 0.060098 |
| anaemia | 0.088006 | 1.000000 | -0.190741 | -0.012729 | 0.031557 |
| creatinine_phosphokinase | -0.081584 | -0.190741 | 1.000000 | -0.009639 | -0.044080 |
| diabetes | -0.101012 | -0.012729 | -0.009639 | 1.000000 | -0.004850 |
| ejection_fraction | 0.060098 | 0.031557 | -0.044080 | -0.004850 | 1.000000 |
| high_blood_pressure | 0.093289 | 0.038182 | -0.070590 | -0.012732 | 0.024445 |
| platelets | -0.052354 | -0.043786 | 0.024463 | 0.092193 | 0.072177 |
| serum_creatinine | 0.159187 | 0.052174 | -0.016408 | -0.046975 | -0.011302 |
| serum_sodium | -0.045966 | 0.041882 | 0.059550 | -0.089551 | 0.175902 |
| sex | 0.065430 | -0.094769 | 0.079791 | -0.157730 | -0.148386 |
| smoking | 0.018668 | -0.107290 | 0.002421 | -0.147173 | -0.067315 |
| time | -0.224068 | -0.141414 | -0.009346 | 0.033726 | 0.041729 |
| DEATH_EVENT | 0.253729 | 0.066270 | 0.062728 | -0.001943 | -0.268603 |
| 4 | | | | | |

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```
#High correlation between age and creatinine serum
In [64]:
In [65]:
          #Age, high blood pressure, and serum creatinine correlate to chances of death the most.
          df.corr()['age'].sort_values()
Out[65]: time
                                     -0.224068
         diabetes
                                     -0.101012
         creatinine_phosphokinase
                                     -0.081584
                                     -0.052354
         platelets
         serum_sodium
                                     -0.045966
         smoking
                                      0.018668
         ejection_fraction
                                      0.060098
                                      0.065430
         sex
                                      0.088006
         anaemia
         high blood pressure
                                      0.093289
         serum creatinine
                                      0.159187
         DEATH_EVENT
                                      0.253729
                                      1.000000
         age
         Name: age, dtype: float64
In [66]: | #Within the data set, less than half the people died from heart failure.
          df['DEATH_EVENT'].value_counts()
               203
Out[66]:
               96
         Name: DEATH EVENT, dtype: int64
          df.drop('time',axis=1,inplace=True)
In [67]:
In [68]:
          from sklearn.model_selection import train_test_split
In [69]:
          X = df.drop('DEATH_EVENT',axis=1).values
          y = df['DEATH EVENT'].values
          X_train, X_test, y_train, y_test = train_test_split(X, y,
          test_size=0.25, random_state=101)
          from sklearn.preprocessing import MinMaxScaler
In [70]:
          scaler = MinMaxScaler()
In [71]:
          X_train = scaler.fit_transform(X_train)
In [72]:
In [73]:
          X_test = scaler.transform(X_test)
In [74]:
          from tensorflow.keras.models import Sequential
          from tensorflow.keras.layers import Dense,Dropout
In [75]:
          X_train.shape
In [76]:
Out[76]: (224, 11)
In [77]:
          model = Sequential()
```

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```
model.add(Dense(11,activation='relu'))
model.add(Dense(5,activation='relu'))
# BINARY CLASSIFICATION
model.add(Dense(1,activation='sigmoid'))
model.compile(loss='binary_crossentropy',optimizer='adam')
```

```
In [78]: model.fit(x=X_train,y=y_train,epochs=600,validation_data=(X_test,y_test))
```

```
Epoch 1/600
7/7 [============ ] - 1s 43ms/step - loss: 0.6935 - val loss: 0.6888
Epoch 2/600
7/7 [==========] - 0s 10ms/step - loss: 0.6916 - val loss: 0.6854
Epoch 4/600
Epoch 5/600
Epoch 6/600
Epoch 8/600
Epoch 9/600
Epoch 10/600
Epoch 11/600
Epoch 12/600
Epoch 13/600
Epoch 14/600
Epoch 15/600
Epoch 16/600
Epoch 17/600
Epoch 18/600
Epoch 19/600
Epoch 20/600
Epoch 21/600
Epoch 22/600
Epoch 23/600
Epoch 24/600
Epoch 25/600
Epoch 27/600
```

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```
Epoch 28/600
Epoch 29/600
Epoch 30/600
Epoch 31/600
Epoch 32/600
Epoch 33/600
Epoch 34/600
Epoch 35/600
Epoch 36/600
Epoch 37/600
Epoch 38/600
Epoch 39/600
Epoch 40/600
Epoch 41/600
Epoch 42/600
Epoch 43/600
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Epoch 45/600
Epoch 46/600
Epoch 47/600
Epoch 48/600
Epoch 49/600
Epoch 50/600
Epoch 51/600
Epoch 52/600
Epoch 53/600
Epoch 54/600
Epoch 55/600
Epoch 56/600
Epoch 57/600
Epoch 58/600
Epoch 59/600
```

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Epoch 60/600
Epoch 61/600
Epoch 62/600
Epoch 63/600
Epoch 64/600
Epoch 65/600
Epoch 66/600
Epoch 67/600
Epoch 68/600
Epoch 69/600
Epoch 70/600
Epoch 71/600
Epoch 72/600
Epoch 73/600
Epoch 74/600
Epoch 75/600
7/7 [===========] - ETA: 0s - loss: 0.691 - 0s 5ms/step - loss: 0.594
6 - val loss: 0.5469
Epoch 76/600
Epoch 77/600
Epoch 78/600
Epoch 79/600
Epoch 80/600
7/7 [============] - 0s 6ms/step - loss: 0.5858 - val_loss: 0.5414
Epoch 81/600
Epoch 82/600
Epoch 83/600
Epoch 84/600
Epoch 85/600
Epoch 86/600
Epoch 87/600
Epoch 88/600
Epoch 89/600
Epoch 90/600
Epoch 91/600
```

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```
Epoch 92/600
Epoch 93/600
Epoch 94/600
7/7 [==========] - 0s 6ms/step - loss: 0.5520 - val_loss: 0.5162
Epoch 95/600
Epoch 96/600
Epoch 97/600
Epoch 98/600
Epoch 99/600
Epoch 100/600
Epoch 101/600
Epoch 102/600
Epoch 103/600
Epoch 104/600
Epoch 105/600
Epoch 106/600
Epoch 107/600
Epoch 108/600
Epoch 109/600
Epoch 110/600
7/7 [===========] - 0s 6ms/step - loss: 0.5104 - val_loss: 0.5278
Epoch 111/600
Epoch 112/600
Epoch 113/600
Epoch 114/600
Epoch 115/600
Epoch 116/600
Epoch 117/600
Epoch 118/600
Epoch 119/600
Epoch 120/600
Epoch 121/600
Epoch 122/600
Epoch 123/600
Epoch 124/600
```

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Epoch 125/600
Epoch 126/600
Epoch 127/600
Epoch 128/600
Epoch 129/600
Epoch 130/600
Epoch 131/600
Epoch 132/600
Epoch 133/600
Epoch 134/600
Epoch 135/600
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Epoch 148/600
Epoch 149/600
Epoch 150/600
Epoch 151/600
Epoch 152/600
Epoch 153/600
Epoch 154/600
Epoch 155/600
Epoch 156/600
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Epoch 157/600
Epoch 158/600
Epoch 159/600
Epoch 160/600
Epoch 161/600
Epoch 162/600
Epoch 163/600
Epoch 164/600
Epoch 165/600
Epoch 166/600
Epoch 167/600
Epoch 168/600
Epoch 169/600
Epoch 170/600
Epoch 171/600
Epoch 172/600
Epoch 173/600
Epoch 174/600
Epoch 175/600
7/7 [==========] - 0s 5ms/step - loss: 0.4532 - val_loss: 0.5444
Epoch 176/600
Epoch 177/600
Epoch 178/600
Epoch 179/600
Epoch 180/600
Epoch 181/600
Epoch 182/600
Epoch 183/600
Epoch 184/600
Epoch 185/600
Epoch 186/600
Epoch 187/600
Epoch 188/600
Epoch 189/600
```

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Epoch 190/600
Epoch 191/600
Epoch 192/600
Epoch 193/600
Epoch 194/600
Epoch 195/600
Epoch 196/600
Epoch 197/600
Epoch 198/600
Epoch 199/600
Epoch 200/600
Epoch 201/600
7/7 [==========] - 0s 5ms/step - loss: 0.4466 - val loss: 0.5518
Epoch 202/600
Epoch 203/600
Epoch 204/600
Epoch 205/600
Epoch 206/600
Epoch 207/600
Epoch 208/600
Epoch 209/600
Epoch 210/600
Epoch 211/600
Epoch 212/600
Epoch 213/600
Epoch 214/600
Epoch 215/600
Epoch 216/600
Epoch 217/600
Epoch 218/600
Epoch 219/600
Epoch 220/600
Epoch 221/600
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Epoch 222/600
Epoch 223/600
Epoch 224/600
Epoch 225/600
Epoch 226/600
Epoch 227/600
Epoch 228/600
Epoch 229/600
Epoch 230/600
Epoch 231/600
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Epoch 235/600
Epoch 236/600
Epoch 237/600
Epoch 238/600
Epoch 239/600
Epoch 240/600
Epoch 241/600
Epoch 242/600
7/7 [===========] - 0s 4ms/step - loss: 0.4381 - val_loss: 0.5689
Epoch 243/600
Epoch 244/600
Epoch 245/600
7/7 [==========] - 0s 4ms/step - loss: 0.4368 - val loss: 0.5655
Epoch 246/600
Epoch 247/600
Epoch 248/600
Epoch 249/600
Epoch 250/600
Epoch 251/600
Epoch 252/600
Epoch 253/600
7/7 [============ ] - 0s 5ms/step - loss: 0.4362 - val loss: 0.5707
Epoch 254/600
```

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Epoch 255/600
Epoch 256/600
Epoch 257/600
Epoch 258/600
Epoch 259/600
Epoch 260/600
Epoch 261/600
Epoch 262/600
Epoch 263/600
Epoch 264/600
Epoch 265/600
7/7 [==========] - 0s 5ms/step - loss: 0.4344 - val_loss: 0.5663
Epoch 266/600
Epoch 267/600
Epoch 268/600
Epoch 269/600
Epoch 270/600
Epoch 271/600
Epoch 272/600
Epoch 273/600
Epoch 274/600
Epoch 275/600
Epoch 276/600
Epoch 277/600
Epoch 278/600
Epoch 279/600
Epoch 280/600
Epoch 281/600
7/7 [==========] - 0s 4ms/step - loss: 0.4324 - val_loss: 0.5694
Epoch 282/600
Epoch 283/600
Epoch 284/600
Epoch 285/600
Epoch 286/600
```

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Epoch 287/600
Epoch 288/600
Epoch 289/600
Epoch 290/600
Epoch 291/600
Epoch 292/600
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Epoch 296/600
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Epoch 301/600
Epoch 302/600
Epoch 303/600
Epoch 304/600
Epoch 305/600
Epoch 306/600
Epoch 307/600
Epoch 308/600
7/7 [==========] - 0s 4ms/step - loss: 0.4266 - val loss: 0.5837
Epoch 309/600
Epoch 310/600
Epoch 311/600
Epoch 312/600
Epoch 313/600
Epoch 314/600
Epoch 315/600
Epoch 316/600
Epoch 317/600
Epoch 318/600
7/7 [============ ] - 0s 4ms/step - loss: 0.4280 - val loss: 0.5869
Epoch 319/600
```

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Epoch 320/600
Epoch 321/600
Epoch 322/600
Epoch 323/600
Epoch 324/600
Epoch 325/600
Epoch 326/600
Epoch 327/600
Epoch 328/600
Epoch 329/600
Epoch 330/600
Epoch 331/600
Epoch 332/600
Epoch 333/600
Epoch 334/600
Epoch 335/600
7/7 [==========] - 0s 4ms/step - loss: 0.4236 - val loss: 0.5847
Epoch 336/600
Epoch 337/600
Epoch 338/600
Epoch 339/600
Epoch 340/600
Epoch 341/600
Epoch 342/600
Epoch 343/600
Epoch 344/600
Epoch 345/600
Epoch 346/600
7/7 [==========] - 0s 4ms/step - loss: 0.4218 - val_loss: 0.5894
Epoch 347/600
Epoch 348/600
Epoch 349/600
Epoch 350/600
Epoch 351/600
```

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Epoch 352/600
Epoch 353/600
Epoch 354/600
Epoch 355/600
Epoch 356/600
Epoch 357/600
Epoch 358/600
Epoch 359/600
Epoch 360/600
7/7 [==========] - 0s 4ms/step - loss: 0.4204 - val loss: 0.5897
Epoch 361/600
Epoch 362/600
7/7 [==========] - 0s 4ms/step - loss: 0.4192 - val loss: 0.5935
Epoch 363/600
Epoch 364/600
Epoch 365/600
Epoch 366/600
Epoch 367/600
7/7 [==========] - 0s 4ms/step - loss: 0.4208 - val_loss: 0.5988
Epoch 368/600
Epoch 369/600
Epoch 370/600
Epoch 371/600
Epoch 372/600
Epoch 373/600
Epoch 374/600
Epoch 375/600
Epoch 376/600
Epoch 377/600
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Epoch 382/600
Epoch 383/600
Epoch 384/600
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Epoch 385/600
Epoch 386/600
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Epoch 388/600
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Epoch 405/600
Epoch 406/600
Epoch 407/600
Epoch 408/600
Epoch 409/600
Epoch 410/600
Epoch 411/600
7/7 [==========] - 0s 4ms/step - loss: 0.4129 - val_loss: 0.5981
Epoch 412/600
Epoch 413/600
Epoch 414/600
Epoch 415/600
Epoch 416/600
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Epoch 417/600
Epoch 418/600
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Epoch 449/600
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Epoch 450/600
Epoch 451/600
Epoch 452/600
Epoch 453/600
Epoch 454/600
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Epoch 456/600
Epoch 457/600
Epoch 458/600
Epoch 459/600
Epoch 460/600
Epoch 461/600
Epoch 462/600
Epoch 463/600
7/7 [==========] - 0s 4ms/step - loss: 0.4069 - val loss: 0.6054
Epoch 464/600
Epoch 465/600
Epoch 466/600
Epoch 467/600
Epoch 468/600
Epoch 469/600
Epoch 470/600
Epoch 471/600
Epoch 472/600
7/7 [==========] - 0s 4ms/step - loss: 0.4065 - val loss: 0.6069
Epoch 473/600
Epoch 474/600
Epoch 475/600
Epoch 476/600
Epoch 477/600
Epoch 478/600
Epoch 479/600
Epoch 480/600
Epoch 481/600
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Epoch 482/600
Epoch 483/600
Epoch 484/600
Epoch 485/600
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Epoch 496/600
Epoch 497/600
Epoch 498/600
Epoch 499/600
Epoch 500/600
Epoch 501/600
Epoch 502/600
7/7 [==========] - 0s 4ms/step - loss: 0.4022 - val_loss: 0.6053
Epoch 503/600
Epoch 504/600
Epoch 505/600
Epoch 506/600
Epoch 507/600
Epoch 508/600
Epoch 509/600
Epoch 510/600
Epoch 511/600
Epoch 512/600
Epoch 513/600
Epoch 514/600
```

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```
Epoch 515/600
Epoch 516/600
7/7 [==========] - 0s 4ms/step - loss: 0.4009 - val loss: 0.6095
Epoch 517/600
Epoch 518/600
Epoch 519/600
Epoch 520/600
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Epoch 545/600
Epoch 546/600
7/7 [==========] - 0s 4ms/step - loss: 0.3970 - val_loss: 0.6120
```

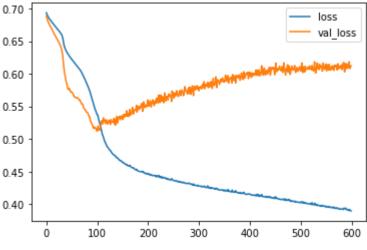
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```
Epoch 547/600
Epoch 548/600
Epoch 549/600
Epoch 550/600
Epoch 551/600
Epoch 552/600
Epoch 553/600
Epoch 554/600
Epoch 555/600
Epoch 556/600
7/7 [==========] - 0s 4ms/step - loss: 0.3960 - val loss: 0.6055
Epoch 557/600
7/7 [==========] - 0s 4ms/step - loss: 0.3965 - val loss: 0.6144
Epoch 558/600
Epoch 559/600
Epoch 560/600
7/7 [==========] - 0s 4ms/step - loss: 0.3950 - val loss: 0.6057
Epoch 561/600
Epoch 562/600
Epoch 563/600
7/7 [==========] - 0s 5ms/step - loss: 0.3952 - val loss: 0.6070
Epoch 564/600
Epoch 565/600
Epoch 566/600
Epoch 567/600
Epoch 568/600
Epoch 569/600
Epoch 570/600
Epoch 571/600
4 - val loss: 0.6117
Epoch 572/600
Epoch 573/600
Epoch 574/600
Epoch 575/600
Epoch 576/600
Epoch 577/600
Epoch 578/600
7/7 [==========] - 0s 4ms/step - loss: 0.3925 - val_loss: 0.6088
```

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```
Epoch 579/600
  Epoch 580/600
  Epoch 581/600
  Epoch 582/600
  Epoch 583/600
  7/7 [==========] - 0s 4ms/step - loss: 0.3920 - val loss: 0.6068
  Epoch 584/600
  Epoch 585/600
  7/7 [===========] - 0s 4ms/step - loss: 0.3923 - val_loss: 0.6173
  Epoch 587/600
  Epoch 588/600
  Epoch 589/600
  7/7 [==========] - 0s 4ms/step - loss: 0.3936 - val loss: 0.6155
  Epoch 590/600
  Epoch 591/600
  Epoch 592/600
  Epoch 593/600
  Epoch 594/600
  Epoch 595/600
  Epoch 596/600
  7/7 [===========] - 0s 4ms/step - loss: 0.3920 - val_loss: 0.6185
  Epoch 597/600
  Epoch 598/600
  Epoch 599/600
  Epoch 600/600
  Out[78]: <keras.callbacks.History at 0x1f5c53adc70>
   losses = pd.DataFrame(model.history.history)
In [79]:
   losses.plot()
In [80]:
Out[80]: <AxesSubplot:>
```

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```
In [81]:
     #Data was way too overfit, needed to use callbacks
     model = Sequential()
In [82]:
     model.add(Dense(11,activation='relu'))
     model.add(Dense(5,activation='relu'))
     model.add(Dense(1,activation='sigmoid'))
     model.compile(loss='binary crossentropy',optimizer='adam')
     from tensorflow.keras.callbacks import EarlyStopping
In [83]:
In [84]:
     early stop = EarlyStopping(monitor='val loss', mode='min',
     verbose=1,patience=25)
     model.fit(x=X_train,y=y_train,epochs=600,validation_data=(X_test,y_test),
In [85]:
           callbacks=[early stop])
     Epoch 1/600
     7/7 [============= ] - 0s 21ms/step - loss: 0.6860 - val loss: 0.6722
     Epoch 2/600
     Epoch 3/600
     Epoch 4/600
     7/7 [============== ] - 0s 4ms/step - loss: 0.6665 - val_loss: 0.6485
     Epoch 5/600
     Epoch 6/600
     7/7 [============== ] - 0s 5ms/step - loss: 0.6581 - val_loss: 0.6389
     Epoch 7/600
     7/7 [============== ] - 0s 5ms/step - loss: 0.6549 - val_loss: 0.6345
     Epoch 8/600
     Epoch 9/600
     Epoch 10/600
     Epoch 11/600
     Epoch 12/600
```

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```
Epoch 13/600
Epoch 14/600
Epoch 15/600
Epoch 16/600
Epoch 17/600
Epoch 18/600
Epoch 19/600
Epoch 20/600
Epoch 21/600
Epoch 22/600
Epoch 23/600
7/7 [==========] - 0s 4ms/step - loss: 0.6270 - val loss: 0.5947
Epoch 24/600
Epoch 25/600
Epoch 26/600
7/7 [==========] - 0s 4ms/step - loss: 0.6246 - val_loss: 0.5930
Epoch 27/600
Epoch 28/600
Epoch 29/600
Epoch 30/600
Epoch 31/600
Epoch 32/600
Epoch 33/600
Epoch 34/600
Epoch 35/600
Epoch 36/600
Epoch 37/600
Epoch 38/600
Epoch 39/600
Epoch 40/600
Epoch 41/600
Epoch 42/600
Epoch 43/600
Epoch 44/600
Epoch 45/600
```

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```
Epoch 46/600
Epoch 47/600
Epoch 48/600
Epoch 49/600
7/7 [==========] - 0s 4ms/step - loss: 0.6069 - val loss: 0.5827
Epoch 50/600
Epoch 51/600
Epoch 52/600
Epoch 53/600
Epoch 54/600
Epoch 55/600
Epoch 56/600
Epoch 57/600
Epoch 58/600
Epoch 59/600
Epoch 60/600
Epoch 61/600
Epoch 62/600
Epoch 63/600
Epoch 64/600
Epoch 65/600
Epoch 66/600
Epoch 67/600
Epoch 68/600
Epoch 69/600
Epoch 70/600
Epoch 71/600
Epoch 72/600
Epoch 73/600
Epoch 74/600
Epoch 75/600
Epoch 76/600
Epoch 77/600
```

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```
Epoch 78/600
Epoch 79/600
Epoch 80/600
Epoch 81/600
Epoch 82/600
Epoch 83/600
Epoch 84/600
Epoch 85/600
Epoch 86/600
Epoch 87/600
Epoch 88/600
Epoch 89/600
Epoch 90/600
Epoch 91/600
Epoch 92/600
Epoch 93/600
Epoch 94/600
Epoch 95/600
Epoch 96/600
Epoch 97/600
Epoch 98/600
Epoch 99/600
Epoch 100/600
Epoch 101/600
7/7 [==========] - 0s 4ms/step - loss: 0.5203 - val loss: 0.5537
Epoch 102/600
Epoch 103/600
Epoch 104/600
Epoch 105/600
Epoch 106/600
Epoch 107/600
Epoch 108/600
Epoch 109/600
Epoch 110/600
```

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```
Epoch 111/600
Epoch 112/600
Epoch 113/600
Epoch 114/600
Epoch 115/600
Epoch 116/600
Epoch 117/600
Epoch 118/600
Epoch 119/600
Epoch 120/600
Epoch 121/600
Epoch 122/600
Epoch 123/600
Epoch 124/600
Epoch 125/600
Epoch 126/600
Epoch 127/600
Epoch 128/600
Epoch 129/600
Epoch 130/600
Epoch 131/600
Epoch 132/600
Epoch 133/600
Epoch 134/600
Epoch 135/600
Epoch 136/600
Epoch 137/600
Epoch 138/600
Epoch 139/600
Epoch 140/600
Epoch 141/600
Epoch 142/600
```

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In [86]:

```
Epoch 143/600
 Epoch 144/600
 Epoch 145/600
 Epoch 146/600
 Epoch 147/600
 Epoch 148/600
 Epoch 149/600
 Epoch 151/600
 Epoch 152/600
 Epoch 153/600
 Epoch 154/600
 Epoch 155/600
 Epoch 156/600
 7/7 [==========] - 0s 4ms/step - loss: 0.4808 - val_loss: 0.5364
 Epoch 157/600
 Epoch 158/600
 7/7 [==========] - 0s 4ms/step - loss: 0.4798 - val_loss: 0.5400
 Epoch 159/600
 Epoch 160/600
 Epoch 161/600
 Epoch 162/600
 Epoch 163/600
 Epoch 164/600
 Epoch 165/600
 Epoch 166/600
 Epoch 167/600
 Epoch 168/600
 Epoch 169/600
 Epoch 170/600
 Epoch 171/600
 Epoch 00171: early stopping
Out[85]: <keras.callbacks.History at 0x1f5c553b910>
```

model_loss.plot()

model loss = pd.DataFrame(model.history.history)

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In [87]:

Epoch 9/600

Epoch 10/600

Epoch 11/600

Epoch 12/600

```
Out[86]: <AxesSubplot:>
```

```
0.65 - loss val_loss - val_loss - val_loss - 0.55 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 - 0.50 -
```

#Training got better, will now add dropout layers

```
In [88]:
      from tensorflow.keras.layers import Dropout
In [89]:
      model = Sequential()
      model.add(Dense(11,activation='relu'))
      model.add(Dropout(0.5))
      model.add(Dense(5,activation='relu'))
      model.add(Dropout(0.5))
      model.add(Dense(1,activation='sigmoid'))
      model.compile(loss='binary_crossentropy',optimizer='adam')
In [90]:
      model.fit(x=X_train,y=y_train,epochs=600,validation_data=(X_test,y_test),
           callbacks=[early stop])
     Epoch 1/600
     7/7 [============ ] - 0s 21ms/step - loss: 0.7794 - val loss: 0.6063
     Epoch 2/600
     Epoch 3/600
     Epoch 4/600
     Epoch 5/600
     Epoch 6/600
     7/7 [============ ] - 0s 4ms/step - loss: 0.7026 - val loss: 0.6037
     Epoch 7/600
     Epoch 8/600
```

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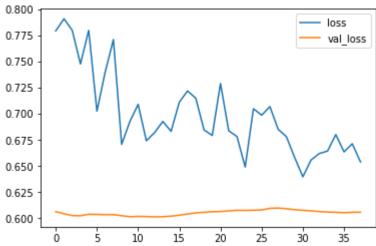
7/7 [============] - 0s 4ms/step - loss: 0.6927 - val loss: 0.6015

7/7 [==========] - 0s 4ms/step - loss: 0.7090 - val_loss: 0.6018

Epoch 13/600

```
Epoch 14/600
   Epoch 15/600
   7/7 [==========] - 0s 4ms/step - loss: 0.6833 - val loss: 0.6020
   Epoch 16/600
   Epoch 17/600
   Epoch 18/600
   Epoch 19/600
   Epoch 20/600
   Epoch 21/600
   Epoch 22/600
   Epoch 23/600
   7/7 [==========] - 0s 4ms/step - loss: 0.6783 - val loss: 0.6077
   Epoch 24/600
   Epoch 25/600
   Epoch 26/600
   Epoch 27/600
   7/7 [==========] - 0s 5ms/step - loss: 0.7069 - val loss: 0.6095
   Epoch 28/600
   Epoch 29/600
   Epoch 30/600
   7/7 [===========] - 0s 4ms/step - loss: 0.6583 - val_loss: 0.6083
   Epoch 31/600
   Epoch 32/600
   Epoch 33/600
   Epoch 34/600
   7/7 [==========] - 0s 4ms/step - loss: 0.6643 - val loss: 0.6059
   Epoch 35/600
   Epoch 36/600
   7/7 [==========] - 0s 4ms/step - loss: 0.6636 - val loss: 0.6053
   Epoch 37/600
   Epoch 38/600
   Epoch 00038: early stopping
Out[90]: <keras.callbacks.History at 0x1f5c6692d00>
   model loss = pd.DataFrame(model.history.history)
In [91]:
   model_loss.plot()
Out[91]: <AxesSubplot:>
```

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```
#The neuro network is too small to use 0.5 dropouts.
In [92]:
          df.corr()['DEATH_EVENT'].sort_values()
In [93]:
         ejection fraction
                                     -0.268603
Out[93]:
          serum sodium
                                      -0.195204
         platelets
                                      -0.049139
          smoking
                                      -0.012623
                                      -0.004316
          sex
         diabetes
                                      -0.001943
          creatinine_phosphokinase
                                       0.062728
          anaemia
                                       0.066270
         high_blood_pressure
                                       0.079351
          age
                                       0.253729
          serum_creatinine
                                      0.294278
         DEATH EVENT
                                       1.000000
         Name: DEATH_EVENT, dtype: float64
          df.drop('ejection_fraction',axis=1,inplace=True)
In [94]:
          X = df.drop('DEATH_EVENT',axis=1).values
In [95]:
          y = df['DEATH EVENT'].values
          X_train, X_test, y_train, y_test = train_test_split(X, y,
          test_size=0.25, random_state=101)
In [96]:
          from sklearn.preprocessing import MinMaxScaler
          scaler = MinMaxScaler()
In [97]:
In [98]:
          X_train = scaler.fit_transform(X_train)
In [99]:
          X test = scaler.transform(X test)
In [100...
          from tensorflow.keras.models import Sequential
In [101...
          from tensorflow.keras.layers import Dense,Dropout
          model = Sequential()
In [102...
          model.add(Dense(11,activation='relu'))
```

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```
model.add(Dense(5,activation='relu'))
   # BINARY CLASSIFICATION
   model.add(Dense(1,activation='sigmoid'))
   model.compile(loss='binary crossentropy',optimizer='adam')
   from tensorflow.keras.callbacks import EarlyStopping
In [103...
In [104...
   early_stop = EarlyStopping(monitor='val_loss',mode='min',
   verbose=1,patience=25)
   model.fit(x=X train,y=y train,epochs=600,validation data=(X test,y test),
In [105...
      callbacks=[early stop])
   Epoch 1/600
   7/7 [=========] - 0s 22ms/step - loss: 0.8354 - val loss: 0.8569
   Epoch 2/600
   Epoch 3/600
   Epoch 4/600
   Epoch 5/600
   Epoch 6/600
   Epoch 7/600
   Epoch 8/600
   Epoch 9/600
   Epoch 10/600
   7/7 [==========] - 0s 4ms/step - loss: 0.7036 - val loss: 0.7018
   Epoch 11/600
   Epoch 12/600
   Epoch 13/600
   Epoch 14/600
   Epoch 15/600
   Epoch 16/600
   Epoch 17/600
   Epoch 18/600
   Epoch 19/600
   Epoch 20/600
   Epoch 21/600
   Epoch 22/600
   Epoch 23/600
   Epoch 24/600
```

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```
Epoch 25/600
Epoch 26/600
Epoch 27/600
Epoch 28/600
Epoch 29/600
Epoch 30/600
Epoch 31/600
Epoch 32/600
Epoch 33/600
Epoch 34/600
Epoch 35/600
Epoch 36/600
Epoch 37/600
Epoch 38/600
7/7 [==========] - 0s 4ms/step - loss: 0.6224 - val loss: 0.5825
Epoch 39/600
Epoch 40/600
Epoch 41/600
Epoch 42/600
Epoch 44/600
Epoch 45/600
Epoch 46/600
Epoch 47/600
Epoch 48/600
7/7 [==========] - 0s 4ms/step - loss: 0.6096 - val loss: 0.5720
Epoch 49/600
Epoch 50/600
Epoch 51/600
Epoch 52/600
Epoch 53/600
Epoch 54/600
Epoch 55/600
Epoch 56/600
7/7 [=========] - 0s 5ms/step - loss: 0.5989 - val_loss: 0.5655
```

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```
Epoch 57/600
Epoch 58/600
Epoch 59/600
Epoch 60/600
Epoch 61/600
Epoch 62/600
Epoch 63/600
Epoch 64/600
Epoch 65/600
Epoch 66/600
Epoch 67/600
Epoch 68/600
Epoch 69/600
Epoch 70/600
7/7 [==========] - 0s 4ms/step - loss: 0.5777 - val_loss: 0.5562
Epoch 71/600
Epoch 72/600
Epoch 73/600
Epoch 74/600
Epoch 75/600
Epoch 76/600
Epoch 77/600
Epoch 78/600
Epoch 79/600
Epoch 80/600
Epoch 81/600
Epoch 82/600
Epoch 83/600
Epoch 84/600
Epoch 85/600
Epoch 86/600
Epoch 87/600
Epoch 88/600
Epoch 89/600
```

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```
Epoch 90/600
 Epoch 91/600
 7/7 [==========] - 0s 4ms/step - loss: 0.5492 - val loss: 0.5493
 Epoch 92/600
 Epoch 93/600
 Epoch 94/600
 Epoch 95/600
 Epoch 96/600
 Epoch 97/600
 Epoch 98/600
 Epoch 99/600
 Epoch 100/600
 Epoch 101/600
 Epoch 102/600
 Epoch 103/600
 Epoch 104/600
 Epoch 105/600
 Epoch 106/600
 Epoch 107/600
 Epoch 108/600
 Epoch 109/600
 Epoch 110/600
 Epoch 111/600
 Epoch 112/600
 Epoch 113/600
 Epoch 114/600
 Epoch 115/600
 Epoch 116/600
 Epoch 117/600
 Epoch 118/600
 Epoch 00118: early stopping
Out[105... <keras.callbacks.History at 0x1f5c6844be0>
```

In [106... model_loss = pd.DataFrame(model.history.history)
 model_loss.plot()

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```
Out[106... <AxesSubplot:>
        0.85
                                               loss
                                               val loss
        0.80
        0.75
        0.70
        0.65
        0.60
        0.55
                          40
                                60
                                      80
                                            100
                   20
                                                   120
In [107...
         from tensorflow.keras.layers import Dropout
         X train.shape
In [127...
Out[127... (224, 10)
In [122...
         model = Sequential()
         model.add(Dense(11,activation='relu'))
         model.add(Dropout(0.6))
         model.add(Dense(5,activation='relu'))
         model.add(Dropout(0.6))
         model.add(Dense(1,activation='sigmoid'))
         model.compile(loss='binary crossentropy',optimizer='adam')
         model.fit(x=X train,y=y train,epochs=600,validation data=(X test,y test),
In [123...
                 callbacks=[early stop])
        Epoch 1/600
        7/7 [============ ] - 1s 49ms/step - loss: 0.8210 - val loss: 0.8158
        Epoch 2/600
        7/7 [============ ] - 0s 10ms/step - loss: 0.8776 - val loss: 0.7944
        Epoch 3/600
        Epoch 4/600
        7/7 [==========] - 0s 10ms/step - loss: 0.8473 - val_loss: 0.7579
        Epoch 5/600
        7/7 [============ ] - 0s 10ms/step - loss: 0.7872 - val loss: 0.7438
        Epoch 6/600
        7/7 [==========] - 0s 10ms/step - loss: 0.7646 - val_loss: 0.7313
        Epoch 7/600
        7/7 [============= ] - 0s 11ms/step - loss: 0.7939 - val_loss: 0.7199
        Epoch 8/600
        7/7 [==========] - 0s 10ms/step - loss: 0.7357 - val_loss: 0.7113
        Epoch 9/600
        7/7 [============= ] - 0s 10ms/step - loss: 0.7170 - val_loss: 0.7043
        Epoch 10/600
```

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Epoch 11/600

7/7 [==========] - 0s 10ms/step - loss: 0.7622 - val_loss: 0.6977

```
7/7 [==========] - 0s 8ms/step - loss: 0.7305 - val loss: 0.6920
Epoch 12/600
Epoch 13/600
Epoch 14/600
Epoch 15/600
Epoch 16/600
Epoch 17/600
Epoch 18/600
Epoch 19/600
Epoch 20/600
Epoch 21/600
Epoch 22/600
Epoch 23/600
7/7 [==========] - 0s 8ms/step - loss: 0.6971 - val loss: 0.6573
Epoch 24/600
Epoch 25/600
Epoch 26/600
Epoch 27/600
Epoch 28/600
Epoch 29/600
Epoch 31/600
Epoch 32/600
Epoch 33/600
Epoch 34/600
Epoch 35/600
7/7 [==========] - 0s 8ms/step - loss: 0.6672 - val loss: 0.6401
Epoch 36/600
Epoch 37/600
Epoch 38/600
Epoch 39/600
Epoch 40/600
Epoch 41/600
Epoch 42/600
Epoch 43/600
```

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```
Epoch 44/600
Epoch 45/600
2 - val loss: 0.6286
Epoch 46/600
Epoch 47/600
Epoch 48/600
7/7 [============] - 0s 10ms/step - loss: 0.6554 - val loss: 0.6255
Epoch 49/600
7/7 [==========] - 0s 10ms/step - loss: 0.6696 - val_loss: 0.6246
Epoch 50/600
Epoch 51/600
Epoch 52/600
Epoch 53/600
7/7 [==========] - 0s 9ms/step - loss: 0.6469 - val loss: 0.6207
Epoch 54/600
Epoch 55/600
7/7 [==========] - 0s 8ms/step - loss: 0.6658 - val loss: 0.6197
Epoch 56/600
Epoch 57/600
7/7 [==========] - 0s 10ms/step - loss: 0.6596 - val loss: 0.6190
Epoch 58/600
Epoch 59/600
Epoch 60/600
Epoch 61/600
Epoch 62/600
Epoch 63/600
Epoch 64/600
Epoch 65/600
7/7 [==========] - 0s 10ms/step - loss: 0.6577 - val loss: 0.6164
Epoch 66/600
2 - val loss: 0.6159
Epoch 67/600
Epoch 68/600
Epoch 69/600
Epoch 70/600
Epoch 71/600
Epoch 72/600
Epoch 73/600
Epoch 74/600
Epoch 75/600
```

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```
Epoch 76/600
Epoch 77/600
Epoch 78/600
Epoch 79/600
Epoch 80/600
Epoch 81/600
Epoch 82/600
Epoch 83/600
Epoch 84/600
Epoch 85/600
Epoch 86/600
Epoch 87/600
Epoch 88/600
Epoch 89/600
Epoch 90/600
Epoch 91/600
Epoch 92/600
Epoch 93/600
7/7 [==========] - 0s 10ms/step - loss: 0.6418 - val_loss: 0.6060
Epoch 94/600
Epoch 95/600
Epoch 96/600
Epoch 97/600
Epoch 98/600
Epoch 99/600
Epoch 100/600
Epoch 101/600
Epoch 102/600
Epoch 103/600
Epoch 104/600
7/7 [==========] - 0s 10ms/step - loss: 0.6376 - val_loss: 0.6046
Epoch 105/600
Epoch 106/600
Epoch 107/600
```

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```
Epoch 108/600
Epoch 109/600
Epoch 110/600
Epoch 111/600
Epoch 112/600
Epoch 113/600
7/7 [============ ] - 0s 10ms/step - loss: 0.6392 - val loss: 0.6030
Epoch 114/600
Epoch 115/600
Epoch 116/600
Epoch 117/600
7/7 [==========] - 0s 9ms/step - loss: 0.6482 - val loss: 0.6024
Epoch 118/600
Epoch 119/600
7/7 [==========] - 0s 8ms/step - loss: 0.6484 - val loss: 0.6027
Epoch 120/600
Epoch 121/600
7/7 [==========] - 0s 8ms/step - loss: 0.6448 - val loss: 0.6027
Epoch 122/600
2 - val loss: 0.6025
Epoch 123/600
Epoch 124/600
Epoch 125/600
Epoch 126/600
Epoch 127/600
Epoch 128/600
Epoch 129/600
Epoch 130/600
Epoch 131/600
Epoch 132/600
Epoch 133/600
Epoch 134/600
Epoch 135/600
Epoch 136/600
Epoch 137/600
Epoch 138/600
Epoch 139/600
```

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```
Epoch 140/600
Epoch 141/600
Epoch 142/600
Epoch 143/600
Epoch 144/600
Epoch 145/600
Epoch 146/600
Epoch 147/600
Epoch 148/600
7/7 [==========] - 0s 8ms/step - loss: 0.6462 - val loss: 0.5985
Epoch 149/600
7/7 [=========] - 0s 10ms/step - loss: 0.6390 - val loss: 0.5984
Epoch 150/600
Epoch 151/600
Epoch 152/600
Epoch 153/600
Epoch 154/600
Epoch 155/600
Epoch 156/600
Epoch 157/600
Epoch 158/600
Epoch 159/600
Epoch 160/600
Epoch 161/600
Epoch 162/600
Epoch 163/600
Epoch 164/600
7/7 [==========] - 0s 11ms/step - loss: 0.6389 - val loss: 0.5963
Epoch 165/600
Epoch 166/600
Epoch 167/600
Epoch 168/600
Epoch 169/600
Epoch 170/600
Epoch 171/600
Epoch 172/600
```

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```
Epoch 173/600
Epoch 174/600
Epoch 175/600
Epoch 176/600
7/7 [=========] - 0s 10ms/step - loss: 0.6392 - val loss: 0.5953
Epoch 177/600
Epoch 178/600
Epoch 179/600
Epoch 180/600
Epoch 181/600
Epoch 182/600
Epoch 183/600
7/7 [==========] - 0s 11ms/step - loss: 0.6442 - val loss: 0.5948
Epoch 184/600
Epoch 185/600
Epoch 186/600
Epoch 187/600
Epoch 188/600
Epoch 189/600
Epoch 190/600
Epoch 191/600
Epoch 192/600
Epoch 193/600
Epoch 194/600
7/7 [============ ] - 0s 10ms/step - loss: 0.6377 - val loss: 0.5944
Epoch 195/600
Epoch 196/600
Epoch 197/600
7/7 [==========] - 0s 9ms/step - loss: 0.6366 - val loss: 0.5944
Epoch 198/600
Epoch 199/600
7/7 [==========] - 0s 8ms/step - loss: 0.6382 - val loss: 0.5942
Epoch 200/600
Epoch 201/600
Epoch 202/600
Epoch 203/600
Epoch 204/600
```

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```
Epoch 205/600
Epoch 206/600
Epoch 207/600
Epoch 208/600
Epoch 209/600
Epoch 210/600
Epoch 211/600
Epoch 212/600
Epoch 213/600
Epoch 214/600
7/7 [==========] - 0s 8ms/step - loss: 0.6372 - val loss: 0.5934
Epoch 215/600
7/7 [==========] - 0s 10ms/step - loss: 0.6462 - val loss: 0.5933
Epoch 216/600
Epoch 217/600
Epoch 218/600
Epoch 219/600
Epoch 220/600
Epoch 221/600
Epoch 222/600
7/7 [==========] - 0s 11ms/step - loss: 0.6374 - val_loss: 0.5930
Epoch 223/600
7/7 [===========] - 0s 9ms/step - loss: 0.6449 - val_loss: 0.5930
Epoch 224/600
Epoch 225/600
Epoch 226/600
Epoch 227/600
Epoch 228/600
7/7 [=========] - 0s 10ms/step - loss: 0.6437 - val loss: 0.5927
Epoch 229/600
Epoch 230/600
Epoch 231/600
Epoch 232/600
Epoch 233/600
Epoch 234/600
Epoch 235/600
7/7 [==========] - 0s 8ms/step - loss: 0.6381 - val_loss: 0.5928
Epoch 236/600
7/7 [============ ] - 0s 11ms/step - loss: 0.6350 - val loss: 0.5929
Epoch 237/600
```

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```
Epoch 238/600
5 - val loss: 0.5925
Epoch 239/600
Epoch 240/600
Epoch 241/600
Epoch 242/600
Epoch 243/600
Epoch 245/600
Epoch 246/600
Epoch 247/600
Epoch 248/600
Epoch 249/600
Epoch 250/600
7/7 [==========] - 0s 9ms/step - loss: 0.6378 - val_loss: 0.5920
Epoch 251/600
Epoch 252/600
Epoch 253/600
Epoch 254/600
Epoch 255/600
Epoch 256/600
Epoch 257/600
Epoch 258/600
7/7 [==========] - 0s 8ms/step - loss: 0.6361 - val loss: 0.5918
Epoch 259/600
Epoch 260/600
7/7 [==========] - 0s 9ms/step - loss: 0.6365 - val loss: 0.5918
Epoch 261/600
Epoch 262/600
Epoch 263/600
Epoch 264/600
Epoch 265/600
Epoch 266/600
Epoch 268/600
Epoch 269/600
```

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```
Epoch 270/600
Epoch 271/600
Epoch 272/600
Epoch 273/600
7/7 [==========] - 0s 10ms/step - loss: 0.6337 - val loss: 0.5914
Epoch 274/600
7/7 [============ ] - 0s 10ms/step - loss: 0.6430 - val loss: 0.5913
Epoch 275/600
7/7 [==========] - 0s 10ms/step - loss: 0.6388 - val_loss: 0.5913
Epoch 276/600
Epoch 277/600
Epoch 278/600
7/7 [==========] - 0s 10ms/step - loss: 0.6363 - val loss: 0.5911
Epoch 279/600
Epoch 280/600
Epoch 281/600
7/7 [==========] - 0s 8ms/step - loss: 0.6429 - val loss: 0.5910
Epoch 282/600
Epoch 283/600
Epoch 284/600
Epoch 285/600
Epoch 286/600
Epoch 287/600
Epoch 288/600
Epoch 289/600
Epoch 290/600
Epoch 291/600
Epoch 292/600
7/7 [==========] - 0s 8ms/step - loss: 0.6432 - val loss: 0.5905
Epoch 293/600
Epoch 294/600
Epoch 295/600
Epoch 296/600
Epoch 297/600
Epoch 298/600
Epoch 299/600
Epoch 300/600
7/7 [==========] - 0s 8ms/step - loss: 0.6368 - val loss: 0.5908
Epoch 301/600
```

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```
Epoch 302/600
Epoch 303/600
Epoch 304/600
Epoch 305/600
Epoch 306/600
Epoch 307/600
Epoch 308/600
Epoch 309/600
Epoch 310/600
Epoch 311/600
Epoch 312/600
Epoch 313/600
Epoch 314/600
Epoch 315/600
Epoch 316/600
Epoch 317/600
Epoch 318/600
Epoch 319/600
Epoch 320/600
Epoch 321/600
Epoch 322/600
Epoch 323/600
7/7 [==========] - 0s 10ms/step - loss: 0.6400 - val loss: 0.5898
Epoch 324/600
7/7 [==========] - 0s 10ms/step - loss: 0.6351 - val loss: 0.5898
Epoch 325/600
Epoch 326/600
Epoch 327/600
Epoch 328/600
Epoch 329/600
Epoch 330/600
Epoch 331/600
Epoch 332/600
Epoch 333/600
Epoch 334/600
```

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```
Epoch 335/600
Epoch 336/600
Epoch 337/600
Epoch 338/600
Epoch 339/600
Epoch 340/600
Epoch 341/600
Epoch 342/600
7/7 [==========] - 0s 10ms/step - loss: 0.6302 - val_loss: 0.5870
Epoch 343/600
7/7 [============ ] - 0s 10ms/step - loss: 0.6297 - val loss: 0.5868
Epoch 344/600
7/7 [==========] - 0s 10ms/step - loss: 0.6340 - val loss: 0.5867
Epoch 345/600
Epoch 346/600
Epoch 347/600
Epoch 348/600
Epoch 349/600
7/7 [==========] - 0s 11ms/step - loss: 0.6385 - val loss: 0.5866
Epoch 350/600
Epoch 351/600
Epoch 352/600
Epoch 353/600
Epoch 354/600
Epoch 355/600
Epoch 356/600
Epoch 357/600
Epoch 358/600
Epoch 359/600
Epoch 360/600
Epoch 361/600
Epoch 362/600
Epoch 363/600
Epoch 364/600
Epoch 365/600
7/7 [==========] - 0s 10ms/step - loss: 0.6323 - val loss: 0.5850
Epoch 366/600
```

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```
Epoch 367/600
Epoch 368/600
7/7 [============= ] - 0s 12ms/step - loss: 0.6315 - val loss: 0.5852
Epoch 369/600
7/7 [==========] - 0s 9ms/step - loss: 0.6289 - val_loss: 0.5849
Epoch 370/600
Epoch 371/600
Epoch 372/600
Epoch 373/600
7/7 [==========] - 0s 10ms/step - loss: 0.6292 - val_loss: 0.5843
Epoch 374/600
7/7 [==========] - 0s 11ms/step - loss: 0.6366 - val_loss: 0.5840
Epoch 375/600
Epoch 376/600
Epoch 377/600
Epoch 378/600
Epoch 379/600
Epoch 380/600
Epoch 381/600
Epoch 382/600
Epoch 383/600
7/7 [============= ] - 0s 10ms/step - loss: 0.6313 - val_loss: 0.5831
Epoch 384/600
7/7 [==========] - 0s 11ms/step - loss: 0.6307 - val_loss: 0.5829
Epoch 385/600
Epoch 386/600
7/7 [==========] - 0s 10ms/step - loss: 0.6343 - val_loss: 0.5829
Epoch 387/600
Epoch 388/600
7/7 [==========] - 0s 10ms/step - loss: 0.6423 - val loss: 0.5829
Epoch 389/600
Epoch 390/600
7/7 [==========] - 0s 9ms/step - loss: 0.6385 - val_loss: 0.5835
Epoch 391/600
Epoch 392/600
Epoch 393/600
Epoch 394/600
Epoch 395/600
Epoch 396/600
Epoch 397/600
7/7 [==========] - 0s 8ms/step - loss: 0.6273 - val_loss: 0.5828
Epoch 398/600
7/7 [============ ] - 0s 10ms/step - loss: 0.6293 - val loss: 0.5825
Epoch 399/600
```

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```
9 - val loss: 0.5822
Epoch 400/600
Epoch 401/600
Epoch 402/600
Epoch 403/600
Epoch 404/600
Epoch 405/600
Epoch 406/600
7/7 [==========] - 0s 10ms/step - loss: 0.6242 - val_loss: 0.5819
Epoch 407/600
Epoch 408/600
7/7 [==========] - 0s 8ms/step - loss: 0.6354 - val loss: 0.5817
Epoch 409/600
Epoch 410/600
7/7 [==========] - 0s 10ms/step - loss: 0.6289 - val loss: 0.5818
Epoch 411/600
7/7 [============= ] - 0s 10ms/step - loss: 0.6265 - val loss: 0.5815
Epoch 412/600
Epoch 413/600
Epoch 414/600
Epoch 415/600
Epoch 416/600
Epoch 417/600
7/7 [===========] - 0s 8ms/step - loss: 0.6287 - val_loss: 0.5796
Epoch 418/600
7/7 [==========] - 0s 10ms/step - loss: 0.6307 - val_loss: 0.5795
Epoch 419/600
Epoch 420/600
7/7 [==========] - 0s 10ms/step - loss: 0.6260 - val loss: 0.5793
Epoch 421/600
Epoch 422/600
Epoch 423/600
Epoch 424/600
Epoch 425/600
Epoch 426/600
Epoch 427/600
Epoch 428/600
Epoch 429/600
Epoch 430/600
Epoch 431/600
```

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```
Epoch 432/600
Epoch 433/600
Epoch 434/600
Epoch 435/600
Epoch 436/600
7/7 [============ ] - 0s 10ms/step - loss: 0.6253 - val loss: 0.5790
Epoch 437/600
Epoch 438/600
Epoch 439/600
Epoch 440/600
7/7 [============ ] - 0s 10ms/step - loss: 0.6252 - val loss: 0.5784
Epoch 441/600
Epoch 442/600
Epoch 443/600
Epoch 444/600
Epoch 445/600
7/7 [==========] - 0s 10ms/step - loss: 0.6239 - val loss: 0.5785
Epoch 446/600
9 - val loss: 0.5779
Epoch 447/600
7/7 [============= ] - 0s 10ms/step - loss: 0.6284 - val loss: 0.5776
Epoch 448/600
Epoch 449/600
Epoch 450/600
Epoch 451/600
Epoch 452/600
Epoch 453/600
Epoch 454/600
Epoch 455/600
Epoch 456/600
Epoch 457/600
Epoch 458/600
Epoch 459/600
7/7 [============== ] - 0s 10ms/step - loss: 0.6245 - val_loss: 0.5758
Epoch 460/600
Epoch 461/600
Epoch 462/600
Epoch 463/600
```

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```
7/7 [==========] - ETA: 0s - loss: 0.775 - 0s 9ms/step - loss: 0.623
8 - val loss: 0.5761
Epoch 464/600
Epoch 465/600
Epoch 466/600
Epoch 467/600
Epoch 468/600
Epoch 469/600
7/7 [==========] - 0s 10ms/step - loss: 0.6296 - val_loss: 0.5753
7/7 [==========] - 0s 10ms/step - loss: 0.6172 - val_loss: 0.5753
Epoch 471/600
Epoch 472/600
Epoch 473/600
Epoch 474/600
7/7 [==========] - 0s 10ms/step - loss: 0.6149 - val loss: 0.5729
Epoch 475/600
Epoch 476/600
Epoch 477/600
Epoch 478/600
Epoch 479/600
Epoch 480/600
Epoch 481/600
Epoch 482/600
Epoch 483/600
Epoch 484/600
Epoch 485/600
Epoch 486/600
7/7 [==========] - 0s 11ms/step - loss: 0.6387 - val loss: 0.5753
Epoch 487/600
7/7 [==========] - 0s 10ms/step - loss: 0.6222 - val loss: 0.5758
Epoch 488/600
Epoch 489/600
Epoch 490/600
Epoch 491/600
Epoch 492/600
Epoch 493/600
Epoch 494/600
Epoch 495/600
```

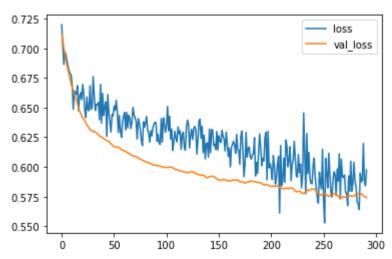
localhost:8891/lab 51/59

```
Epoch 496/600
           =====] - 0s 9ms/step - loss: 0.6243 - val loss: 0.5739
7/7 [========
Epoch 497/600
        Epoch 498/600
       ==========] - 0s 9ms/step - loss: 0.6486 - val loss: 0.5746
7/7 [=======
Epoch 500/600
Epoch 00500: early stopping
```

Out[123... <keras.callbacks.History at 0x1f5c9e20550>

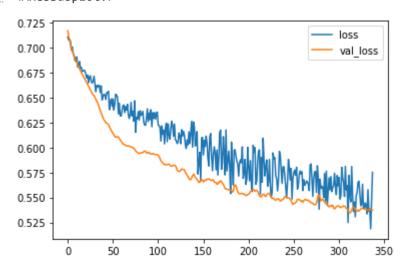
```
In [110...
          \#dropout = 0.5
          model loss = pd.DataFrame(model.history.history)
          model loss.plot()
```

Out[110... <AxesSubplot:>



```
In [115...
          \#dropout = 0.2
          model loss = pd.DataFrame(model.history.history)
          model loss.plot()
```

Out[115... <AxesSubplot:>

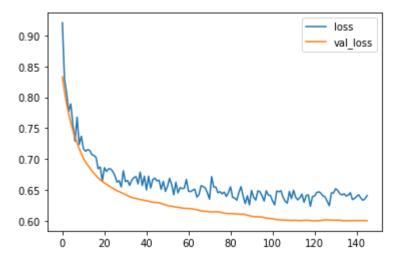


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```
In [118... #dropout = 0.6

model_loss = pd.DataFrame(model.history.history)
model_loss.plot()
```

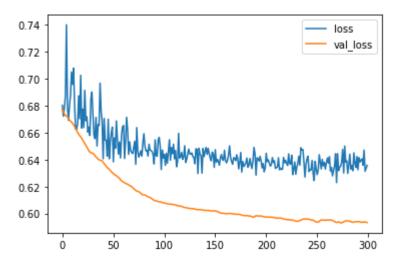
```
Out[118... <AxesSubplot:>
```



```
In [121... #dropout = 0.7

model_loss = pd.DataFrame(model.history.history)
model_loss.plot()
```

Out[121... <AxesSubplot:>

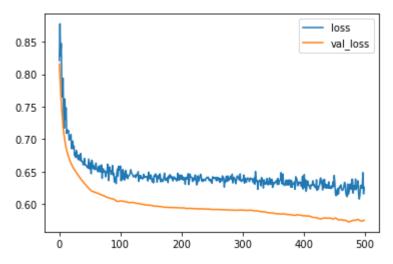


```
In [124... #best dropout = 0.6

model_loss = pd.DataFrame(model.history.history)
model_loss.plot()
```

Out[124... <AxesSubplot:>

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```
In [130... model = Sequential()
    model.add(Dense(10,activation='relu'))
    model.add(Dense(5,activation='relu'))
    model.add(Dense(1,activation='sigmoid'))
    model.compile(loss='binary_crossentropy',optimizer='adam')
    model.fit(x=X_train,y=y_train,epochs=600,validation_data=(X_test,y_test),callbacks=[early_stop])
```

```
Epoch 1/600
7/7 [=========== ] - 1s 45ms/step - loss: 0.7231 - val loss: 0.7235
Epoch 2/600
Epoch 3/600
Epoch 4/600
Epoch 5/600
Epoch 6/600
Epoch 7/600
5 - val loss: 0.6844
Epoch 8/600
Epoch 9/600
Epoch 10/600
Epoch 11/600
Epoch 12/600
7/7 [============= ] - 0s 10ms/step - loss: 0.6786 - val loss: 0.6677
Epoch 13/600
Epoch 14/600
Epoch 15/600
Epoch 16/600
```

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```
Epoch 17/600
Epoch 18/600
7/7 [============] - 0s 7ms/step - loss: 0.6672 - val_loss: 0.6486
Epoch 19/600
7/7 [==========] - 0s 8ms/step - loss: 0.6652 - val loss: 0.6455
Epoch 20/600
Epoch 21/600
Epoch 22/600
Epoch 23/600
Epoch 24/600
Epoch 25/600
Epoch 26/600
Epoch 27/600
Epoch 28/600
Epoch 29/600
Epoch 30/600
Epoch 31/600
Epoch 32/600
Epoch 33/600
7/7 [==========] - 0s 10ms/step - loss: 0.6474 - val loss: 0.6086
Epoch 34/600
Epoch 35/600
Epoch 36/600
Epoch 37/600
Epoch 38/600
Epoch 39/600
Epoch 40/600
Epoch 41/600
Epoch 42/600
Epoch 43/600
Epoch 44/600
5 - val loss: 0.5968
Epoch 45/600
Epoch 46/600
Epoch 47/600
Epoch 48/600
```

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```
Epoch 49/600
Epoch 50/600
Epoch 51/600
Epoch 52/600
Epoch 53/600
Epoch 54/600
7/7 [===========] - 0s 10ms/step - loss: 0.6393 - val_loss: 0.5913
Epoch 55/600
Epoch 56/600
Epoch 57/600
Epoch 58/600
7/7 [==========] - 0s 9ms/step - loss: 0.6380 - val_loss: 0.5896
Epoch 59/600
Epoch 60/600
Epoch 61/600
Epoch 62/600
Epoch 63/600
Epoch 64/600
Epoch 65/600
Epoch 66/600
Epoch 67/600
Epoch 68/600
Epoch 69/600
Epoch 70/600
Epoch 71/600
Epoch 72/600
Epoch 73/600
Epoch 74/600
Epoch 75/600
Epoch 76/600
7/7 [==========] - 0s 10ms/step - loss: 0.6321 - val_loss: 0.5823
Epoch 77/600
7/7 [===========] - 0s 8ms/step - loss: 0.6319 - val_loss: 0.5826
Epoch 78/600
Epoch 79/600
Epoch 80/600
Epoch 81/600
```

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```
Epoch 82/600
7/7 [============ - 0s 10ms/step - loss: 0.6299 - val loss: 0.5803
Epoch 83/600
Epoch 84/600
Epoch 85/600
Epoch 86/600
Epoch 87/600
Epoch 88/600
Epoch 89/600
Epoch 90/600
Epoch 91/600
7/7 [==========] - 0s 9ms/step - loss: 0.6262 - val loss: 0.5773
Epoch 92/600
Epoch 93/600
Epoch 94/600
Epoch 95/600
Epoch 96/600
Epoch 97/600
7/7 [==========] - 0s 8ms/step - loss: 0.6236 - val loss: 0.5753
Epoch 98/600
Epoch 99/600
Epoch 100/600
Epoch 101/600
Epoch 102/600
Epoch 103/600
Epoch 104/600
Epoch 105/600
Epoch 106/600
Epoch 107/600
Epoch 108/600
Epoch 109/600
Epoch 110/600
Epoch 111/600
Epoch 112/600
Epoch 113/600
```

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```
Epoch 114/600
Epoch 115/600
Epoch 116/600
Epoch 117/600
Epoch 118/600
Epoch 119/600
Epoch 120/600
Epoch 121/600
Epoch 122/600
Epoch 123/600
Epoch 124/600
Epoch 125/600
Epoch 126/600
Epoch 127/600
Epoch 128/600
7/7 [===========] - 0s 8ms/step - loss: 0.5911 - val_loss: 0.5701
Epoch 129/600
Epoch 130/600
Epoch 131/600
Epoch 132/600
Epoch 133/600
Epoch 134/600
Epoch 135/600
7/7 [==========] - 0s 9ms/step - loss: 0.5817 - val_loss: 0.5725
Epoch 136/600
Epoch 137/600
Epoch 138/600
Epoch 139/600
Epoch 140/600
Epoch 141/600
Epoch 142/600
7/7 [===========] - 0s 7ms/step - loss: 0.5725 - val_loss: 0.5719
Epoch 143/600
Epoch 144/600
Epoch 145/600
Epoch 146/600
```

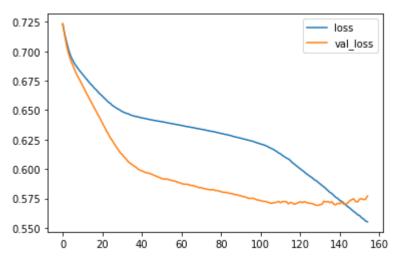
localhost:8891/lab 58/59

```
Epoch 147/600
Epoch 148/600
Epoch 149/600
Epoch 150/600
Epoch 151/600
Epoch 152/600
Epoch 153/600
Epoch 154/600
7/7 [===========] - 0s 8ms/step - loss: 0.5559 - val_loss: 0.5743
Epoch 155/600
Epoch 00155: early stopping
```

Out[130... <keras.callbacks.History at 0x1f5cb286400>

```
In [131... model_loss = pd.DataFrame(model.history.history)
model_loss.plot()
```

Out[131... <AxesSubplot:>



```
In [ ]:
```

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