```
In [2]:
          import numpy as np
          import pandas as pd
          import matplotlib.pyplot as plt
 In [4]:
          dataFrame = pd.read_excel("websites.xlsx")
          dataFrame
               Type URL_LENGTH NUMBER_SPECIAL_CHARACTERS TCP_CONVERSATION_EXCHANGE DIST_REI
 Out[4]:
            0
                       23.303047
                                                     13.445560
                                                                                  159.066933
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                                                     23.018073
                                                                                  172.149800
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                                                     27.525833
                                                                                  168.393333
                        14.792707
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                                                                                  174.999533
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                                                     37.992127
                                                                                  181.476200
                       10.051787
          547
                  0
                                                     31.787480
                                                                                  62.072373
         548 rows × 31 columns
 In [5]:
          y = dataFrame["Type"].values
          x = dataFrame.drop("Type",axis=1).values
 In [6]:
          from sklearn.model_selection import train_test_split
          x_train, x_test, y_train, y_test = train_test_split(x,y,test_size=0.28)
 In [7]:
          from sklearn.preprocessing import MinMaxScaler
          scaler = MinMaxScaler()
          x_train = scaler.fit_transform(x_train)
          x_test = scaler.transform(x_test)
In [21]:
          import tensorflow as tf
          from tensorflow.keras import models
          from tensorflow.keras import layers
          model1 = models.Sequential()
          #kaç tane sütun (x özelliği) varsa ilk katmanda o kadar nörön olması önerilir
          model1.add(layers.Dense(30,activation="relu"))
          #çıkış katmanı nöron sayısı ile ilk katmandaki nöron sayısı arasında olması önerilir
          model1.add(layers.Dense(15,activation="relu"))
          model1.add(layers.Dense(15,activation="relu"))
```

```
#binary classification ve 0-1 arasi çıktılar için sigmoid fonksiyonu
model1.add(layers.Dense(1,activation="sigmoid"))
model1.compile(optimizer="adam",loss="binary_crossentropy")
```

In [22]:

```
model1.fit(x_train,y_train,validation_data=(x_test,y_test),epochs=700)
```

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Epoch 1/700
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13/13 [=================== ] - 0s 5ms/step - loss: 0.1492 - val_loss: 0.19
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13/13 [============] - 0s 5ms/step - loss: 0.0623 - val_loss: 0.19
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13/13 [============] - 0s 4ms/step - loss: 0.0182 - val_loss: 0.22
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Epoch 458/700
13/13 [========================] - 0s 5ms/step - loss: 9.8519e-04 - val_loss:
0.3830
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Epoch 462/700
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Epoch 467/700
0.3882
Epoch 468/700
0.3878
Epoch 469/700
0.3885
Epoch 470/700
13/13 [========================] - 0s 5ms/step - loss: 8.2408e-04 - val_loss:
0.3881
Epoch 471/700
0.3873
Epoch 472/700
0.3891
Epoch 473/700
Epoch 474/700
0.3887
Epoch 475/700
0.3905
Epoch 476/700
0.3925
Epoch 477/700
0.3944
Epoch 478/700
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Epoch 479/700
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Epoch 480/700
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0.3976
Epoch 481/700
13/13 [========================] - 0s 5ms/step - loss: 7.2023e-04 - val_loss:
0.3967
Epoch 482/700
0.3962
Epoch 483/700
0.3977
Epoch 484/700
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Epoch 485/700
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Epoch 486/700
0.4008
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0.4026
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Epoch 489/700
0.4013
Epoch 490/700
0.4039
Epoch 491/700
72
Epoch 492/700
0.4062
Epoch 493/700
13/13 [========================] - 0s 6ms/step - loss: 8.7087e-04 - val_loss:
0.4049
Epoch 494/700
0.4058
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0.4105
Epoch 496/700
Epoch 497/700
0.4106
Epoch 498/700
0.4083
Epoch 499/700
0.4110
Epoch 500/700
0.4139
Epoch 501/700
0.4109
Epoch 502/700
0.4142
Epoch 503/700
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0.4125
Epoch 504/700
13/13 [========================] - 0s 5ms/step - loss: 6.4356e-04 - val_loss:
0.4161
Epoch 505/700
0.4151
Epoch 506/700
0.4158
Epoch 507/700
0.4174
Epoch 508/700
0.4198
Epoch 509/700
0.4208
Epoch 510/700
0.4193
Epoch 511/700
0.4204
Epoch 512/700
0.4211
Epoch 513/700
0.4213
Epoch 514/700
13/13 [========================] - 0s 5ms/step - loss: 6.6831e-04 - val_loss:
0.4286
Epoch 515/700
0.4244
Epoch 516/700
13/13 [========================] - 0s 4ms/step - loss: 5.5343e-04 - val_loss:
0.4229
Epoch 517/700
0.4234
Epoch 518/700
0.4255
Epoch 519/700
0.4225
Epoch 520/700
0.4243
Epoch 521/700
0.4236
Epoch 522/700
0.4291
Epoch 523/700
0.4265
Epoch 524/700
0.4311
Epoch 525/700
0.4315
Epoch 526/700
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0.4331
Epoch 527/700
13/13 [========================] - 0s 5ms/step - loss: 4.5940e-04 - val_loss:
0.4312
Epoch 528/700
0.4323
Epoch 529/700
0.4364
Epoch 530/700
0.4361
Epoch 531/700
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Epoch 532/700
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Epoch 533/700
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Epoch 534/700
0.4386
Epoch 535/700
0.4397
Epoch 536/700
0.4392
Epoch 537/700
13/13 [========================] - 0s 4ms/step - loss: 4.2029e-04 - val_loss:
0.4419
Epoch 538/700
0.4411
Epoch 539/700
13/13 [========================] - 0s 4ms/step - loss: 4.1929e-04 - val_loss:
0.4420
Epoch 540/700
0.4420
Epoch 541/700
0.4428
Epoch 542/700
Epoch 543/700
0.4422
Epoch 544/700
0.4493
Epoch 545/700
0.4484
Epoch 546/700
0.4430
Epoch 547/700
0.4474
Epoch 548/700
0.4525
Epoch 549/700
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0.4490
Epoch 550/700
13/13 [========================] - 0s 5ms/step - loss: 3.8822e-04 - val_loss:
0.4473
Epoch 551/700
0.4467
Epoch 552/700
0.4503
Epoch 553/700
0.4509
Epoch 554/700
0.4548
Epoch 555/700
0.4537
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0.4534
Epoch 558/700
0.4530
Epoch 559/700
0.4532
Epoch 560/700
13/13 [========================] - 0s 6ms/step - loss: 3.4184e-04 - val_loss:
0.4558
Epoch 561/700
0.4611
Epoch 562/700
13/13 [========================] - 0s 5ms/step - loss: 3.5377e-04 - val_loss:
0.4591
Epoch 563/700
0.4627
Epoch 564/700
0.4641
Epoch 565/700
0.4675
Epoch 566/700
0.4610
Epoch 567/700
0.4662
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Epoch 569/700
0.4661
Epoch 570/700
0.4683
Epoch 571/700
0.4676
Epoch 572/700
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0.4695
Epoch 573/700
13/13 [========================] - 0s 5ms/step - loss: 3.0065e-04 - val_loss:
0.4714
Epoch 574/700
0.4701
Epoch 575/700
0.4734
Epoch 576/700
0.4749
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0.4752
Epoch 578/700
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0.4742
Epoch 580/700
0.4793
Epoch 581/700
0.4788
Epoch 582/700
0.4912
Epoch 583/700
13/13 [========================] - 0s 5ms/step - loss: 2.9728e-04 - val_loss:
0.4783
Epoch 584/700
0.4809
Epoch 585/700
13/13 [========================] - 0s 5ms/step - loss: 2.8663e-04 - val_loss:
0.4846
Epoch 586/700
0.4813
Epoch 587/700
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Epoch 588/700
0.4812
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Epoch 590/700
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Epoch 591/700
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Epoch 592/700
0.4941
Epoch 593/700
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Epoch 594/700
0.4788
Epoch 595/700
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0.4797
Epoch 596/700
13/13 [========================] - 0s 5ms/step - loss: 2.4235e-04 - val_loss:
0.5014
Epoch 597/700
0.4903
Epoch 598/700
0.4949
Epoch 599/700
0.4985
Epoch 600/700
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Epoch 601/700
0.4947
Epoch 602/700
0.5004
Epoch 603/700
0.4958
Epoch 604/700
0.5020
Epoch 605/700
0.5017
Epoch 606/700
13/13 [========================] - 0s 5ms/step - loss: 2.7085e-04 - val_loss:
0.4988
Epoch 607/700
0.5039
Epoch 608/700
13/13 [========================] - 0s 5ms/step - loss: 2.1971e-04 - val_loss:
0.4946
Epoch 609/700
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Epoch 613/700
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Epoch 615/700
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Epoch 618/700
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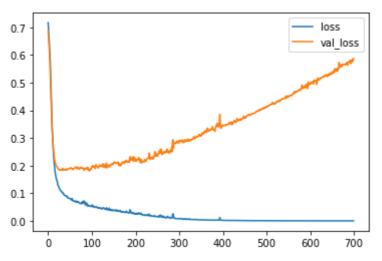
```
0.5160
Epoch 619/700
13/13 [========================] - 0s 5ms/step - loss: 1.7671e-04 - val_loss:
0.5096
Epoch 620/700
0.5110
Epoch 621/700
0.5126
Epoch 622/700
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Epoch 628/700
0.5199
Epoch 629/700
13/13 [========================] - 0s 4ms/step - loss: 1.8113e-04 - val_loss:
0.5202
Epoch 630/700
0.5210
Epoch 631/700
13/13 [========================] - 0s 4ms/step - loss: 1.8093e-04 - val_loss:
0.5228
Epoch 632/700
0.5196
Epoch 633/700
0.5251
Epoch 634/700
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Epoch 639/700
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Epoch 640/700
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Epoch 641/700
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0.5218
Epoch 642/700
13/13 [========================] - 0s 5ms/step - loss: 1.6861e-04 - val_loss:
0.5281
Epoch 643/700
0.5264
Epoch 644/700
0.5318
Epoch 645/700
0.5322
Epoch 646/700
0.5270
Epoch 647/700
0.5312
Epoch 648/700
0.5360
Epoch 649/700
0.5307
Epoch 650/700
0.5332
Epoch 651/700
0.5306
Epoch 652/700
13/13 [========================] - 0s 5ms/step - loss: 1.4676e-04 - val_loss:
0.5357
Epoch 653/700
0.5409
Epoch 654/700
13/13 [========================] - 0s 5ms/step - loss: 1.3775e-04 - val_loss:
0.5350
Epoch 655/700
0.5405
Epoch 656/700
0.5423
Epoch 657/700
0.5503
Epoch 658/700
0.5412
Epoch 659/700
0.5395
Epoch 660/700
0.5449
Epoch 661/700
0.5419
Epoch 662/700
0.5514
Epoch 663/700
0.5464
Epoch 664/700
```

```
0.5422
Epoch 665/700
13/13 [========================] - 0s 5ms/step - loss: 1.6059e-04 - val_loss:
0.5511
Epoch 666/700
0.5503
Epoch 667/700
13/13 [========================] - 0s 5ms/step - loss: 1.7417e-04 - val_loss:
0.5730
Epoch 668/700
0.5535
Epoch 669/700
0.5583
Epoch 670/700
0.5639
Epoch 671/700
0.5584
Epoch 672/700
0.5628
Epoch 673/700
0.5574
Epoch 674/700
0.5667
Epoch 675/700
13/13 [========================] - 0s 5ms/step - loss: 1.2353e-04 - val_loss:
0.5624
Epoch 676/700
0.5650
Epoch 677/700
13/13 [========================] - 0s 5ms/step - loss: 1.1918e-04 - val_loss:
0.5655
Epoch 678/700
0.5658
Epoch 679/700
0.5659
Epoch 680/700
Epoch 681/700
0.5637
Epoch 682/700
0.5679
Epoch 683/700
0.5656
Epoch 684/700
0.5709
Epoch 685/700
0.5714
Epoch 686/700
0.5733
Epoch 687/700
```

```
0.5632
    Epoch 688/700
    13/13 [========================] - 0s 5ms/step - loss: 1.0569e-04 - val_loss:
    0.5736
    Epoch 689/700
    0.5743
    Epoch 690/700
    0.5748
    Epoch 691/700
    0.5737
    Epoch 692/700
    0.5803
    Epoch 693/700
    0.5679
    Epoch 694/700
    0.5802
    Epoch 695/700
    0.5799
    Epoch 696/700
    0.5736
    Epoch 697/700
    0.5809
    Epoch 698/700
    0.5851
    Epoch 699/700
    0.5776
    Epoch 700/700
    0.5870
Out[22]: <tensorflow.python.keras.callbacks.History at 0x1a86ec44790>
In [23]:
    loss df = model1.history.history
    loss_train = loss_df["loss"]
    loss_train = np.array(loss_train)
    loss_test = loss_df["val_loss"]
    loss_test = np.array(loss_test)
    axis = range(0,700)
    plt.plot(axis,loss_train, label="loss")
    plt.plot(axis,loss_test, label="val_loss")
    plt.legend()
    # overfitting aşırı uydurma durumu (epoch fazla geldi)
```

Out[23]: <matplotlib.legend.Legend at 0x1a86d9c4d90>



```
In [25]: # modeli tekrar olusturma

model2 = models.Sequential()

model2.add(layers.Dense(30,activation="relu"))
model2.add(layers.Dense(15,activation="relu"))
model2.add(layers.Dense(15,activation="relu"))
model2.add(layers.Dense(1,activation="sigmoid"))

model2.compile(optimizer="adam",loss="binary_crossentropy")
```

```
Epoch 1/700
1/13 [=>.....] - ETA: 0s - loss: 0.7111WARNING:tensorflow:Ca
llbacks method `on train batch end` is slow compared to the batch time (batch time:
0.0000s vs `on_train_batch_end` time: 0.0040s). Check your callbacks.
13/13 [=============] - 0s 17ms/step - loss: 0.6834 - val_loss: 0.6
714
Epoch 2/700
24
Epoch 3/700
99
Epoch 4/700
06
Epoch 5/700
56
Epoch 6/700
63
Epoch 7/700
17
Epoch 8/700
41
Epoch 9/700
```

```
60
Epoch 10/700
Epoch 11/700
88
Epoch 12/700
Epoch 13/700
55
Epoch 14/700
Epoch 15/700
14
Epoch 16/700
Epoch 17/700
56
Epoch 18/700
Epoch 19/700
37
Epoch 20/700
Epoch 21/700
44
Epoch 22/700
Epoch 23/700
Epoch 24/700
Epoch 25/700
45
Epoch 26/700
Epoch 27/700
32
Epoch 28/700
Epoch 29/700
61
Epoch 30/700
Epoch 31/700
26
Epoch 32/700
```

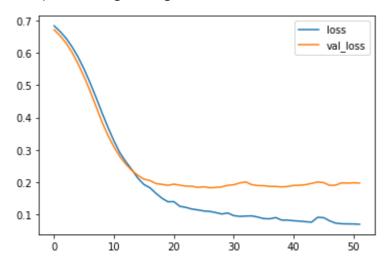
30.03.2021

```
81
 Epoch 33/700
 Epoch 34/700
 Epoch 35/700
 Epoch 36/700
 97
 Epoch 37/700
 Epoch 38/700
 73
 Epoch 39/700
 Epoch 40/700
 72
 Epoch 41/700
 Epoch 42/700
 Epoch 43/700
 Epoch 44/700
 Epoch 45/700
 Epoch 46/700
 Epoch 47/700
 Epoch 48/700
 13
 Epoch 49/700
 Epoch 50/700
 76
 Epoch 51/700
 Epoch 52/700
 77
 Epoch 00052: early stopping
Out[26]: <tensorflow.python.keras.callbacks.History at 0x1a86f0d3ac0>
In [28]:
  loss df = model2.history.history
  loss train = loss df["loss"]
```

```
loss_train = np.array(loss_train)
loss_test = loss_df["val_loss"]
loss_test = np.array(loss_test)

axis = range(0,52) # 52. epoch'da erken durdurma
plt.plot(axis,loss_train, label="loss")
plt.plot(axis,loss_test, label="val_loss")
plt.legend()
```

Out[28]: <matplotlib.legend.Legend at 0x1a87048f370>



```
In [30]: # Dropout (fazla nöronların düşürülmesi - kısa süreli unutma)
# modeli tekrar oluşturma

model3 = models.Sequential()

model3.add(layers.Dense(30,activation="relu"))
model3.add(layers.Dropout(0.4)) #rastgele nöronları düşürmeyi dene

model3.add(layers.Dense(15,activation="relu"))
model3.add(layers.Dropout(0.4))

model3.add(layers.Dense(15,activation="relu"))
model3.add(layers.Dropout(0.4))

model3.add(layers.Dense(1,activation="sigmoid"))
model3.compile(optimizer="adam",loss="binary_crossentropy")
```

```
Epoch 6/700
Epoch 7/700
Epoch 8/700
Epoch 9/700
69
Epoch 10/700
Epoch 11/700
84
Epoch 12/700
Epoch 13/700
94
Epoch 14/700
Epoch 15/700
44
Epoch 16/700
Epoch 17/700
Epoch 18/700
Epoch 19/700
72
Epoch 20/700
Epoch 21/700
47
Epoch 22/700
Epoch 23/700
03
Epoch 24/700
Epoch 25/700
12
Epoch 26/700
Epoch 27/700
30
Epoch 28/700
```

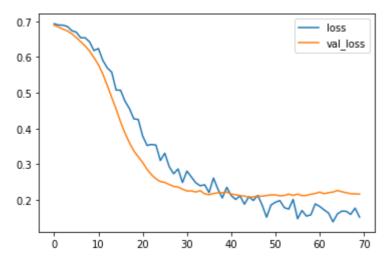
```
Epoch 29/700
Epoch 30/700
Epoch 31/700
Epoch 32/700
53
Epoch 33/700
Epoch 34/700
59
Epoch 35/700
Epoch 36/700
45
Epoch 37/700
Epoch 38/700
98
Epoch 39/700
Epoch 40/700
Epoch 41/700
Epoch 42/700
Epoch 43/700
Epoch 44/700
Epoch 45/700
Epoch 46/700
83
Epoch 47/700
Epoch 48/700
01
Epoch 49/700
Epoch 50/700
40
Epoch 51/700
```

Epoch 52/700

```
Epoch 53/700
  Epoch 54/700
  Epoch 55/700
  26
  Epoch 56/700
  Epoch 57/700
  19
  Epoch 58/700
  Epoch 59/700
  58
  Epoch 60/700
  Epoch 61/700
  19
  Epoch 62/700
  Epoch 63/700
  02
  Epoch 64/700
  Epoch 65/700
  Epoch 66/700
  Epoch 67/700
  Epoch 68/700
  Epoch 69/700
  67
  Epoch 70/700
  Epoch 00070: early stopping
Out[31]: <tensorflow.python.keras.callbacks.History at 0x1a870507d30>
In [32]:
  loss_df = model3.history.history
  loss train = loss df["loss"]
  loss_train = np.array(loss_train)
  loss_test = loss_df["val_loss"]
  loss_test = np.array(loss_test)
```

```
axis = range(0,70) # 70. epoch
plt.plot(axis,loss_train, label="loss")
plt.plot(axis,loss_test, label="val_loss")
plt.legend()
```

Out[32]: <matplotlib.legend.Legend at 0x1a87075c040>



```
In [36]: # siniflandirma sonuçlarını tahmin et
    tahminler = model3.predict_classes(x_test)
    tahminler
```

```
Out[36]: array([[0],
                    [0],
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```
say1 = tahminler.shape[0]
tahminler = pd.Series(tahminler.reshape(say1,))

resultFrame = pd.DataFrame(y_test,columns=["Gerçek Tür"])
resultFrame["Tahmin"] = tahminler
resultFrame
```

Out[37]:		Gerçek Tür	Tahmin
	0	0	0
	1	0	0
	2	0	0
	3	0	0
	4	1	1
	•••		

[0]])

	Gerçek Tür	Tahmin
149	0	0
150	0	0
151	0	0
152	1	1
153	0	0

154 rows × 2 columns

```
In [40]:
          # sınıflandırma sonuçlarını değerlendirme
          from sklearn.metrics import classification_report
          from sklearn.metrics import confusion_matrix
          print(classification_report(y_test,tahminler))
          # %93 accuracy score
                       precision
                                   recall f1-score
                                                       support
                    0
                            0.94
                                                0.94
                                                            95
                                      0.95
                            0.91
                                                0.91
                                      0.90
                                                            59
                    1
                                                0.93
                                                            154
             accuracy
                                                0.92
                                                            154
                            0.93
                                      0.92
            macro avg
                                                0.93
                                                            154
                            0.93
         weighted avg
                                      0.93
In [41]:
          print(confusion_matrix(y_test,tahminler))
         [[90 5]
          [ 6 53]]
In [42]:
          # 0'lar (virüs siteleri tahmininde) 90 doğru, 5 yanlış / %94 doğruluk
          # 1'lar (güvenli siteleri tahmininde) 53 doğru, 6 yanlış / %91 doğruluk
In [ ]:
```