Лабораторная работа №2

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In [35]: import numpy as np
         import math
         import matplotlib.pyplot as plt
         import matplotlib.colors as mclr
         from tensorflow.keras import layers
         from tensorflow.keras import models
         import pandas
         from tensorflow.keras.layers import Dense
         from tensorflow.keras.models import Sequential
         from tensorflow.keras.utils import to categorical
         from sklearn.preprocessing import LabelEncoder
In [36]: dataframe = pandas.read csv("C:\\Users\\loprz\\Downloads\\sonar.csv", header=Nor
         dataset = dataframe.values
         X = dataset[:,0:60].astype(float)
         Y = dataset[:,60]
In [37]:
        encoder = LabelEncoder()
         encoder.fit(Y)
         encoded_Y = encoder.transform(Y)
In [38]:
         model = models.Sequential()
         model.add(layers.Dense(60, input_dim=60, kernel_initializer='normal', activation
         model.add(layers.Dense(1, activation='sigmoid'))
         model2 = models.Sequential()
         model2.add(layers.Dense(30, input_dim=60, kernel_initializer='normal', activation
         model2.add(layers.Dense(1, activation='sigmoid'))
         model3 = models.Sequential()
         model3.add(layers.Dense(60, input_dim=60, kernel_initializer='normal', activation
         model3.add(layers.Dense(15, input dim=60, kernel initializer='normal', activation
         model3.add(layers.Dense(1, activation='sigmoid'))
In [39]:
         model.compile(optimizer='adam',loss='binary_crossentropy', metrics=['accuracy'])
         model2.compile(optimizer='adam',loss='binary crossentropy', metrics=['accuracy']
         model3.compile(optimizer='adam',loss='binary crossentropy', metrics=['accuracy']
In [40]:
        H = model.fit(X, encoded_Y, epochs=100, batch_size=10, validation_split=0.1)
         H2 = model2.fit(X, encoded Y, epochs=100, batch size=10, validation split=0.1)
         H3 = model3.fit(X, encoded_Y, epochs=100, batch_size=10, validation_split=0.1)
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Epoch 1/100
19/19 [============] - 1s 12ms/step - loss: 0.6921 - accuracy:
0.5294 - val_loss: 0.6996 - val_accuracy: 0.3333
Epoch 2/100
0.6631 - val_loss: 0.6884 - val_accuracy: 0.4762
Epoch 3/100
19/19 [============= ] - 0s 4ms/step - loss: 0.6414 - accuracy:
0.7273 - val_loss: 0.7091 - val_accuracy: 0.3810
Epoch 4/100
0.7380 - val loss: 0.7133 - val accuracy: 0.3810
Epoch 5/100
0.7326 - val_loss: 0.6927 - val_accuracy: 0.4286
Epoch 6/100
0.7326 - val loss: 0.7702 - val accuracy: 0.2857
Epoch 7/100
0.7219 - val_loss: 0.7308 - val_accuracy: 0.3810
Epoch 8/100
0.7219 - val_loss: 0.7186 - val_accuracy: 0.3810
Epoch 9/100
19/19 [============== ] - 0s 4ms/step - loss: 0.5349 - accuracy:
0.7380 - val_loss: 0.7197 - val_accuracy: 0.3810
Epoch 10/100
0.7380 - val loss: 0.6799 - val accuracy: 0.4286
Epoch 11/100
0.7380 - val_loss: 0.7129 - val_accuracy: 0.3810
Epoch 12/100
0.7594 - val_loss: 0.7080 - val_accuracy: 0.4286
Epoch 13/100
0.7754 - val_loss: 0.6691 - val_accuracy: 0.4762
Epoch 14/100
0.7701 - val loss: 0.6438 - val accuracy: 0.5238
Epoch 15/100
0.7861 - val_loss: 0.7494 - val_accuracy: 0.3810
Epoch 16/100
0.7861 - val loss: 0.6879 - val accuracy: 0.4762
Epoch 17/100
0.7807 - val_loss: 0.6321 - val_accuracy: 0.5714
Epoch 18/100
0.7701 - val_loss: 0.6075 - val_accuracy: 0.5714
Epoch 19/100
0.7968 - val_loss: 0.7377 - val_accuracy: 0.3810
Epoch 20/100
0.7807 - val_loss: 0.5887 - val_accuracy: 0.6190
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Epoch 21/100
0.7968 - val_loss: 0.5877 - val_accuracy: 0.6190
Epoch 22/100
0.8075 - val_loss: 0.6250 - val_accuracy: 0.5714
Epoch 23/100
0.7754 - val_loss: 0.4991 - val_accuracy: 0.7619
Epoch 24/100
0.7968 - val loss: 0.4856 - val accuracy: 0.7619
Epoch 25/100
0.7968 - val_loss: 0.5117 - val_accuracy: 0.7619
Epoch 26/100
0.8128 - val loss: 0.5663 - val accuracy: 0.6190
Epoch 27/100
0.8182 - val_loss: 0.4942 - val_accuracy: 0.7619
Epoch 28/100
0.8182 - val loss: 0.5893 - val accuracy: 0.6190
Epoch 29/100
19/19 [============== ] - 0s 4ms/step - loss: 0.4103 - accuracy:
0.7968 - val_loss: 0.5385 - val_accuracy: 0.7143
Epoch 30/100
0.8128 - val loss: 0.5592 - val accuracy: 0.6667
Epoch 31/100
0.8182 - val_loss: 0.5084 - val_accuracy: 0.7619
Epoch 32/100
0.8235 - val_loss: 0.5052 - val_accuracy: 0.7619
Epoch 33/100
0.8128 - val_loss: 0.4516 - val_accuracy: 0.7619
Epoch 34/100
0.8235 - val loss: 0.5629 - val accuracy: 0.6190
0.8182 - val_loss: 0.3967 - val_accuracy: 0.8571
Epoch 36/100
0.8235 - val loss: 0.6118 - val accuracy: 0.6190
Epoch 37/100
0.8449 - val_loss: 0.3772 - val_accuracy: 0.9048
Epoch 38/100
0.8289 - val_loss: 0.5023 - val_accuracy: 0.7619
Epoch 39/100
0.8342 - val_loss: 0.5275 - val_accuracy: 0.7143
Epoch 40/100
0.8342 - val_loss: 0.3871 - val_accuracy: 0.8571
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Epoch 41/100
0.8128 - val_loss: 0.4372 - val_accuracy: 0.7619
Epoch 42/100
0.8449 - val_loss: 0.4967 - val_accuracy: 0.7619
Epoch 43/100
0.8503 - val_loss: 0.4725 - val_accuracy: 0.7619
Epoch 44/100
0.8396 - val loss: 0.5280 - val accuracy: 0.7143
Epoch 45/100
0.8182 - val_loss: 0.3222 - val_accuracy: 0.9048
Epoch 46/100
0.8556 - val loss: 0.5325 - val accuracy: 0.6667
Epoch 47/100
0.8610 - val_loss: 0.4315 - val_accuracy: 0.7619
Epoch 48/100
0.8449 - val loss: 0.5006 - val accuracy: 0.7143
Epoch 49/100
19/19 [============== ] - Os 4ms/step - loss: 0.3477 - accuracy:
0.8770 - val_loss: 0.3770 - val_accuracy: 0.8571
Epoch 50/100
0.8556 - val loss: 0.4188 - val accuracy: 0.7619
Epoch 51/100
0.8503 - val_loss: 0.4188 - val_accuracy: 0.7619
Epoch 52/100
0.8770 - val_loss: 0.3947 - val_accuracy: 0.8095
Epoch 53/100
0.8877 - val_loss: 0.4028 - val_accuracy: 0.8095
Epoch 54/100
0.8342 - val loss: 0.2554 - val accuracy: 0.9524
0.8717 - val_loss: 0.4028 - val_accuracy: 0.8095
Epoch 56/100
0.8770 - val loss: 0.3888 - val accuracy: 0.8095
Epoch 57/100
0.8717 - val_loss: 0.4898 - val_accuracy: 0.7143
Epoch 58/100
0.8824 - val_loss: 0.3195 - val_accuracy: 0.9048
Epoch 59/100
0.8717 - val_loss: 0.4089 - val_accuracy: 0.8095
Epoch 60/100
0.8503 - val_loss: 0.5496 - val_accuracy: 0.6190
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Epoch 61/100
0.8824 - val_loss: 0.2679 - val_accuracy: 0.9524
Epoch 62/100
0.8396 - val_loss: 0.3983 - val_accuracy: 0.8095
Epoch 63/100
0.8930 - val_loss: 0.4420 - val_accuracy: 0.8095
Epoch 64/100
0.8824 - val loss: 0.3222 - val accuracy: 0.9048
Epoch 65/100
0.8877 - val_loss: 0.4007 - val_accuracy: 0.8095
Epoch 66/100
0.8930 - val loss: 0.4123 - val accuracy: 0.8095
Epoch 67/100
0.8770 - val_loss: 0.3743 - val_accuracy: 0.8095
Epoch 68/100
0.8770 - val_loss: 0.3535 - val_accuracy: 0.8095
Epoch 69/100
19/19 [============== ] - 0s 4ms/step - loss: 0.3020 - accuracy:
0.8877 - val_loss: 0.3302 - val_accuracy: 0.9048
Epoch 70/100
0.8877 - val loss: 0.3989 - val accuracy: 0.8095
Epoch 71/100
0.8877 - val_loss: 0.2624 - val_accuracy: 0.9048
Epoch 72/100
0.8877 - val_loss: 0.3456 - val_accuracy: 0.8095
Epoch 73/100
0.8930 - val_loss: 0.3105 - val_accuracy: 0.9048
Epoch 74/100
0.8877 - val loss: 0.3681 - val accuracy: 0.8095
Epoch 75/100
0.8930 - val_loss: 0.3417 - val_accuracy: 0.8095
Epoch 76/100
0.8930 - val loss: 0.2997 - val accuracy: 0.9048
Epoch 77/100
0.8984 - val_loss: 0.3219 - val_accuracy: 0.9048
Epoch 78/100
0.8984 - val_loss: 0.2952 - val_accuracy: 0.9048
Epoch 79/100
0.8930 - val_loss: 0.3804 - val_accuracy: 0.8095
Epoch 80/100
0.9037 - val_loss: 0.3481 - val_accuracy: 0.8095
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Epoch 81/100
0.9037 - val_loss: 0.3056 - val_accuracy: 0.9048
Epoch 82/100
0.9037 - val_loss: 0.3274 - val_accuracy: 0.8571
Epoch 83/100
0.9091 - val_loss: 0.3489 - val_accuracy: 0.8095
Epoch 84/100
0.9091 - val loss: 0.3370 - val accuracy: 0.8095
Epoch 85/100
0.9091 - val_loss: 0.2445 - val_accuracy: 0.9524
Epoch 86/100
0.8984 - val loss: 0.3330 - val accuracy: 0.8095
Epoch 87/100
0.9037 - val_loss: 0.3265 - val_accuracy: 0.8571
Epoch 88/100
0.9144 - val loss: 0.3093 - val accuracy: 0.8571
Epoch 89/100
19/19 [============= ] - 0s 4ms/step - loss: 0.2601 - accuracy:
0.9144 - val_loss: 0.2684 - val_accuracy: 0.9048
Epoch 90/100
0.8930 - val loss: 0.2919 - val accuracy: 0.9048
Epoch 91/100
0.9198 - val_loss: 0.2972 - val_accuracy: 0.9048
Epoch 92/100
0.9198 - val_loss: 0.2575 - val_accuracy: 0.9524
Epoch 93/100
0.9091 - val_loss: 0.3247 - val_accuracy: 0.8095
Epoch 94/100
0.9091 - val loss: 0.2730 - val accuracy: 0.9048
0.9091 - val_loss: 0.4036 - val_accuracy: 0.7619
Epoch 96/100
0.9305 - val loss: 0.1957 - val accuracy: 1.0000
Epoch 97/100
0.9144 - val_loss: 0.2510 - val_accuracy: 0.9524
Epoch 98/100
0.9144 - val_loss: 0.2877 - val_accuracy: 0.9048
Epoch 99/100
0.9144 - val_loss: 0.2420 - val_accuracy: 0.9524
Epoch 100/100
0.9305 - val_loss: 0.2750 - val_accuracy: 0.9048
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Epoch 1/100
19/19 [============= ] - 1s 11ms/step - loss: 0.6870 - accuracy:
0.5294 - val_loss: 0.7685 - val_accuracy: 0.0000e+00
Epoch 2/100
0.6310 - val_loss: 0.7617 - val_accuracy: 0.0000e+00
Epoch 3/100
0.6738 - val_loss: 0.7467 - val_accuracy: 0.1429
Epoch 4/100
0.6898 - val loss: 0.7625 - val accuracy: 0.0952
Epoch 5/100
0.7219 - val_loss: 0.7543 - val_accuracy: 0.2381
Epoch 6/100
0.7166 - val loss: 0.7940 - val accuracy: 0.0952
Epoch 7/100
0.7273 - val_loss: 0.7757 - val_accuracy: 0.2381
Epoch 8/100
0.7380 - val_loss: 0.7678 - val_accuracy: 0.3333
Epoch 9/100
19/19 [============== ] - 0s 4ms/step - loss: 0.5873 - accuracy:
0.7219 - val_loss: 0.8029 - val_accuracy: 0.2857
Epoch 10/100
0.7380 - val loss: 0.7588 - val accuracy: 0.3810
Epoch 11/100
0.7433 - val_loss: 0.8200 - val_accuracy: 0.3333
Epoch 12/100
0.7487 - val_loss: 0.7858 - val_accuracy: 0.3333
Epoch 13/100
0.7166 - val_loss: 0.8541 - val_accuracy: 0.3333
Epoch 14/100
0.7540 - val loss: 0.7598 - val accuracy: 0.3810
Epoch 15/100
0.7647 - val_loss: 0.8120 - val_accuracy: 0.3333
Epoch 16/100
0.7701 - val loss: 0.7430 - val accuracy: 0.4286
Epoch 17/100
0.7914 - val_loss: 0.7334 - val_accuracy: 0.4286
Epoch 18/100
0.7594 - val_loss: 0.6941 - val_accuracy: 0.4286
Epoch 19/100
0.7807 - val_loss: 0.7276 - val_accuracy: 0.4286
Epoch 20/100
0.7701 - val_loss: 0.7668 - val_accuracy: 0.3810
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Epoch 21/100
0.7968 - val_loss: 0.7102 - val_accuracy: 0.4286
Epoch 22/100
0.8128 - val_loss: 0.7652 - val_accuracy: 0.3810
Epoch 23/100
0.7487 - val_loss: 0.7574 - val_accuracy: 0.4286
Epoch 24/100
0.7594 - val loss: 0.6741 - val accuracy: 0.4762
Epoch 25/100
0.7647 - val_loss: 0.7535 - val_accuracy: 0.4286
Epoch 26/100
0.8075 - val loss: 0.6235 - val accuracy: 0.5238
Epoch 27/100
0.7968 - val_loss: 0.6407 - val_accuracy: 0.4762
Epoch 28/100
0.7701 - val loss: 0.6737 - val accuracy: 0.4762
Epoch 29/100
19/19 [============== ] - 0s 4ms/step - loss: 0.4455 - accuracy:
0.8075 - val_loss: 0.6434 - val_accuracy: 0.4762
Epoch 30/100
0.7754 - val loss: 0.6683 - val accuracy: 0.4762
Epoch 31/100
0.8075 - val_loss: 0.6303 - val_accuracy: 0.4762
Epoch 32/100
0.7701 - val_loss: 0.5946 - val_accuracy: 0.6667
Epoch 33/100
0.7914 - val_loss: 0.6124 - val_accuracy: 0.4762
Epoch 34/100
0.8182 - val loss: 0.6267 - val accuracy: 0.4762
19/19 [============== ] - 0s 4ms/step - loss: 0.4265 - accuracy:
0.8075 - val_loss: 0.6084 - val_accuracy: 0.5238
Epoch 36/100
0.7861 - val loss: 0.5532 - val accuracy: 0.6667
Epoch 37/100
0.8182 - val_loss: 0.5888 - val_accuracy: 0.6667
Epoch 38/100
0.8128 - val_loss: 0.5762 - val_accuracy: 0.6667
Epoch 39/100
19/19 [============] - 0s 16ms/step - loss: 0.4121 - accuracy:
0.8128 - val_loss: 0.5976 - val_accuracy: 0.6190
Epoch 40/100
0.8182 - val_loss: 0.5974 - val_accuracy: 0.6190
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Epoch 41/100
0.8021 - val_loss: 0.5398 - val_accuracy: 0.7143
Epoch 42/100
0.8075 - val_loss: 0.5958 - val_accuracy: 0.6190
Epoch 43/100
0.8075 - val_loss: 0.5620 - val_accuracy: 0.6667
Epoch 44/100
0.7914 - val loss: 0.5699 - val accuracy: 0.6667
Epoch 45/100
0.8182 - val_loss: 0.5599 - val_accuracy: 0.6667
Epoch 46/100
0.8075 - val loss: 0.5740 - val accuracy: 0.6667
Epoch 47/100
0.8235 - val_loss: 0.4899 - val_accuracy: 0.7619
Epoch 48/100
0.8396 - val loss: 0.5585 - val accuracy: 0.6667
Epoch 49/100
19/19 [============== ] - 0s 4ms/step - loss: 0.3891 - accuracy:
0.8182 - val_loss: 0.5637 - val_accuracy: 0.6667
Epoch 50/100
0.8235 - val loss: 0.5545 - val accuracy: 0.6667
Epoch 51/100
0.8289 - val_loss: 0.5482 - val_accuracy: 0.6667
Epoch 52/100
0.8396 - val_loss: 0.5328 - val_accuracy: 0.7143
Epoch 53/100
0.8289 - val_loss: 0.5290 - val_accuracy: 0.7143
Epoch 54/100
0.8289 - val loss: 0.5214 - val accuracy: 0.7143
0.8235 - val_loss: 0.5619 - val_accuracy: 0.6667
Epoch 56/100
0.8182 - val loss: 0.4913 - val accuracy: 0.7619
Epoch 57/100
0.8449 - val_loss: 0.5075 - val_accuracy: 0.7619
Epoch 58/100
0.8449 - val_loss: 0.5084 - val_accuracy: 0.7143
Epoch 59/100
0.8449 - val_loss: 0.5273 - val_accuracy: 0.7143
Epoch 60/100
0.8449 - val_loss: 0.4931 - val_accuracy: 0.7619
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Epoch 61/100
0.8396 - val_loss: 0.5094 - val_accuracy: 0.7143
Epoch 62/100
0.8503 - val_loss: 0.4378 - val_accuracy: 0.7619
Epoch 63/100
0.8503 - val_loss: 0.4884 - val_accuracy: 0.7619
Epoch 64/100
0.8449 - val loss: 0.5232 - val accuracy: 0.7143
Epoch 65/100
0.8503 - val_loss: 0.4787 - val_accuracy: 0.7619
Epoch 66/100
0.8503 - val loss: 0.4954 - val accuracy: 0.7619
Epoch 67/100
0.8503 - val_loss: 0.4676 - val_accuracy: 0.7619
Epoch 68/100
0.8556 - val loss: 0.4872 - val accuracy: 0.7619
Epoch 69/100
19/19 [============== ] - 0s 4ms/step - loss: 0.3478 - accuracy:
0.8610 - val_loss: 0.4718 - val_accuracy: 0.7619
Epoch 70/100
0.8717 - val loss: 0.4404 - val accuracy: 0.7619
Epoch 71/100
0.8770 - val_loss: 0.4721 - val_accuracy: 0.7619
Epoch 72/100
0.8717 - val_loss: 0.4892 - val_accuracy: 0.7143
Epoch 73/100
0.8610 - val_loss: 0.4088 - val_accuracy: 0.7619
Epoch 74/100
0.8449 - val loss: 0.4996 - val accuracy: 0.7143
Epoch 75/100
0.8663 - val_loss: 0.4516 - val_accuracy: 0.7619
Epoch 76/100
0.8930 - val loss: 0.5123 - val accuracy: 0.7143
Epoch 77/100
0.8824 - val_loss: 0.4677 - val_accuracy: 0.7619
Epoch 78/100
0.8717 - val_loss: 0.4297 - val_accuracy: 0.7619
Epoch 79/100
0.8824 - val_loss: 0.4820 - val_accuracy: 0.7143
Epoch 80/100
0.8824 - val_loss: 0.4758 - val_accuracy: 0.7619
```

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Epoch 81/100
0.8877 - val_loss: 0.4373 - val_accuracy: 0.7619
Epoch 82/100
0.8824 - val_loss: 0.4181 - val_accuracy: 0.7619
Epoch 83/100
0.8824 - val_loss: 0.4453 - val_accuracy: 0.7619
Epoch 84/100
0.8877 - val loss: 0.4373 - val accuracy: 0.7619
Epoch 85/100
0.8877 - val_loss: 0.4406 - val_accuracy: 0.7619
Epoch 86/100
0.8930 - val loss: 0.4438 - val accuracy: 0.7619
Epoch 87/100
0.8877 - val_loss: 0.3976 - val_accuracy: 0.8095
Epoch 88/100
0.8610 - val loss: 0.5478 - val accuracy: 0.6190
Epoch 89/100
19/19 [============== ] - 0s 4ms/step - loss: 0.3248 - accuracy:
0.8824 - val_loss: 0.4159 - val_accuracy: 0.7619
Epoch 90/100
0.8824 - val loss: 0.4040 - val accuracy: 0.8095
Epoch 91/100
0.8877 - val_loss: 0.4718 - val_accuracy: 0.7619
Epoch 92/100
0.8930 - val_loss: 0.3930 - val_accuracy: 0.8095
Epoch 93/100
0.8930 - val_loss: 0.4590 - val_accuracy: 0.7619
Epoch 94/100
0.8877 - val loss: 0.4096 - val accuracy: 0.8095
0.8930 - val_loss: 0.4146 - val_accuracy: 0.8095
Epoch 96/100
0.8930 - val loss: 0.4571 - val accuracy: 0.7619
Epoch 97/100
0.8984 - val_loss: 0.4180 - val_accuracy: 0.8095
Epoch 98/100
0.8877 - val_loss: 0.3985 - val_accuracy: 0.8095
Epoch 99/100
0.8877 - val_loss: 0.4229 - val_accuracy: 0.8095
Epoch 100/100
0.8877 - val_loss: 0.3918 - val_accuracy: 0.8095
```

```
Epoch 1/100
19/19 [============] - 1s 12ms/step - loss: 0.6909 - accuracy:
0.5348 - val_loss: 0.6997 - val_accuracy: 0.3333
Epoch 2/100
0.6845 - val_loss: 0.7159 - val_accuracy: 0.2381
Epoch 3/100
0.7219 - val_loss: 0.7052 - val_accuracy: 0.3810
Epoch 4/100
0.7219 - val loss: 0.7613 - val accuracy: 0.2857
Epoch 5/100
0.7326 - val_loss: 0.8031 - val_accuracy: 0.2381
Epoch 6/100
0.7701 - val loss: 0.6601 - val accuracy: 0.4762
Epoch 7/100
0.7112 - val_loss: 0.7882 - val_accuracy: 0.3333
Epoch 8/100
0.7807 - val loss: 0.9441 - val accuracy: 0.2857
Epoch 9/100
19/19 [============== ] - 0s 4ms/step - loss: 0.5080 - accuracy:
0.7487 - val_loss: 0.7572 - val_accuracy: 0.4286
Epoch 10/100
0.7540 - val loss: 0.3805 - val accuracy: 0.9524
Epoch 11/100
0.7540 - val_loss: 0.6513 - val_accuracy: 0.4762
Epoch 12/100
0.8128 - val_loss: 0.5709 - val_accuracy: 0.6667
Epoch 13/100
0.8021 - val_loss: 0.5919 - val_accuracy: 0.6190
Epoch 14/100
0.7647 - val loss: 0.6446 - val accuracy: 0.4762
Epoch 15/100
19/19 [============== ] - 0s 4ms/step - loss: 0.4192 - accuracy:
0.8289 - val_loss: 0.4728 - val_accuracy: 0.8571
Epoch 16/100
0.8342 - val loss: 0.6953 - val accuracy: 0.4762
Epoch 17/100
0.8396 - val_loss: 0.3880 - val_accuracy: 0.9048
Epoch 18/100
0.7968 - val_loss: 0.6342 - val_accuracy: 0.5238
Epoch 19/100
0.8235 - val_loss: 0.6778 - val_accuracy: 0.4762
Epoch 20/100
0.8182 - val_loss: 0.4586 - val_accuracy: 0.8095
```

```
Epoch 21/100
0.8556 - val_loss: 0.5818 - val_accuracy: 0.6190
Epoch 22/100
0.8342 - val_loss: 0.6465 - val_accuracy: 0.5238
Epoch 23/100
19/19 [============ ] - 0s 4ms/step - loss: 0.3460 - accuracy:
0.8503 - val_loss: 0.3299 - val_accuracy: 0.9524
Epoch 24/100
0.7968 - val loss: 0.2016 - val accuracy: 1.0000
Epoch 25/100
0.8182 - val_loss: 0.4907 - val_accuracy: 0.7619
Epoch 26/100
0.8984 - val loss: 0.6709 - val accuracy: 0.5238
Epoch 27/100
0.8877 - val_loss: 0.4522 - val_accuracy: 0.7619
Epoch 28/100
0.8877 - val loss: 0.5137 - val accuracy: 0.6667
Epoch 29/100
19/19 [============= ] - 0s 4ms/step - loss: 0.3152 - accuracy:
0.8824 - val_loss: 0.4243 - val_accuracy: 0.7619
Epoch 30/100
0.8824 - val loss: 0.7425 - val accuracy: 0.4762
Epoch 31/100
0.8824 - val_loss: 0.5776 - val_accuracy: 0.6190
Epoch 32/100
0.8717 - val_loss: 0.3933 - val_accuracy: 0.8095
Epoch 33/100
0.8503 - val_loss: 0.2984 - val_accuracy: 0.9524
Epoch 34/100
0.8663 - val loss: 0.1951 - val accuracy: 0.9524
0.8717 - val_loss: 0.5564 - val_accuracy: 0.6190
Epoch 36/100
0.9037 - val loss: 0.5415 - val accuracy: 0.6667
Epoch 37/100
0.9144 - val_loss: 0.3880 - val_accuracy: 0.8095
Epoch 38/100
0.9091 - val_loss: 0.3925 - val_accuracy: 0.7619
Epoch 39/100
0.9251 - val_loss: 0.4222 - val_accuracy: 0.7619
Epoch 40/100
0.9144 - val_loss: 0.2720 - val_accuracy: 0.9524
```

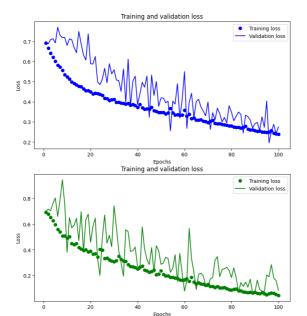
```
Epoch 41/100
0.9144 - val_loss: 0.2463 - val_accuracy: 0.9524
Epoch 42/100
0.8824 - val_loss: 0.5792 - val_accuracy: 0.6190
Epoch 43/100
0.9198 - val_loss: 0.4073 - val_accuracy: 0.7143
Epoch 44/100
0.9305 - val loss: 0.3268 - val accuracy: 0.9048
Epoch 45/100
0.9412 - val_loss: 0.2864 - val_accuracy: 0.9048
Epoch 46/100
0.9251 - val loss: 0.4172 - val accuracy: 0.7619
Epoch 47/100
0.8930 - val_loss: 0.2254 - val_accuracy: 0.9524
Epoch 48/100
0.9412 - val loss: 0.4321 - val accuracy: 0.7143
Epoch 49/100
19/19 [============== ] - 0s 4ms/step - loss: 0.2117 - accuracy:
0.9358 - val_loss: 0.3940 - val_accuracy: 0.7619
Epoch 50/100
0.9091 - val loss: 0.2879 - val accuracy: 0.9048
Epoch 51/100
0.9412 - val_loss: 0.2973 - val_accuracy: 0.9048
Epoch 52/100
0.9519 - val_loss: 0.3428 - val_accuracy: 0.8571
Epoch 53/100
0.9412 - val_loss: 0.3324 - val_accuracy: 0.9048
Epoch 54/100
0.9412 - val loss: 0.2254 - val accuracy: 0.9524
0.9305 - val_loss: 0.2562 - val_accuracy: 0.9524
Epoch 56/100
0.9465 - val loss: 0.3623 - val accuracy: 0.7619
Epoch 57/100
0.9251 - val_loss: 0.1556 - val_accuracy: 1.0000
Epoch 58/100
0.9465 - val_loss: 0.3727 - val_accuracy: 0.7619
Epoch 59/100
0.9465 - val_loss: 0.2196 - val_accuracy: 0.9524
Epoch 60/100
0.9465 - val_loss: 0.0808 - val_accuracy: 1.0000
```

```
Epoch 61/100
0.9358 - val_loss: 0.1604 - val_accuracy: 1.0000
Epoch 62/100
0.9626 - val_loss: 0.5664 - val_accuracy: 0.5714
Epoch 63/100
19/19 [============= ] - 0s 4ms/step - loss: 0.1874 - accuracy:
0.9412 - val_loss: 0.3316 - val_accuracy: 0.8095
Epoch 64/100
0.9465 - val loss: 0.2262 - val accuracy: 0.9524
Epoch 65/100
0.9626 - val_loss: 0.0921 - val_accuracy: 1.0000
Epoch 66/100
0.9572 - val loss: 0.2091 - val accuracy: 0.9524
Epoch 67/100
0.9679 - val_loss: 0.2189 - val_accuracy: 0.9524
Epoch 68/100
0.9572 - val loss: 0.1968 - val accuracy: 0.9524
Epoch 69/100
19/19 [============== ] - 0s 4ms/step - loss: 0.1244 - accuracy:
0.9840 - val_loss: 0.1262 - val_accuracy: 1.0000
Epoch 70/100
0.9626 - val loss: 0.1132 - val accuracy: 1.0000
Epoch 71/100
0.9733 - val_loss: 0.1741 - val_accuracy: 1.0000
Epoch 72/100
0.9679 - val_loss: 0.1868 - val_accuracy: 1.0000
Epoch 73/100
0.9786 - val_loss: 0.3189 - val_accuracy: 0.8095
Epoch 74/100
0.9893 - val loss: 0.3476 - val accuracy: 0.7619
Epoch 75/100
0.9786 - val_loss: 0.1915 - val_accuracy: 0.9524
Epoch 76/100
0.9733 - val loss: 0.2474 - val accuracy: 0.9048
Epoch 77/100
0.9786 - val_loss: 0.2530 - val_accuracy: 0.8571
Epoch 78/100
0.9840 - val_loss: 0.2641 - val_accuracy: 0.8571
Epoch 79/100
0.9840 - val_loss: 0.2413 - val_accuracy: 0.9048
Epoch 80/100
0.9893 - val_loss: 0.1856 - val_accuracy: 0.9524
```

```
Epoch 81/100
0.9893 - val_loss: 0.2593 - val_accuracy: 0.8571
Epoch 82/100
0.9840 - val_loss: 0.1853 - val_accuracy: 1.0000
Epoch 83/100
19/19 [============ ] - 0s 4ms/step - loss: 0.0834 - accuracy:
0.9893 - val_loss: 0.0697 - val_accuracy: 1.0000
Epoch 84/100
0.9893 - val loss: 0.1457 - val accuracy: 1.0000
Epoch 85/100
0.9947 - val_loss: 0.1163 - val_accuracy: 1.0000
Epoch 86/100
0.9947 - val loss: 0.1497 - val accuracy: 1.0000
Epoch 87/100
0.9947 - val_loss: 0.1107 - val_accuracy: 1.0000
Epoch 88/100
19/19 [============== ] - 0s 4ms/step - loss: 0.0713 - accuracy:
0.9893 - val loss: 0.1728 - val accuracy: 1.0000
Epoch 89/100
19/19 [============== ] - 0s 4ms/step - loss: 0.0662 - accuracy:
0.9947 - val_loss: 0.0651 - val_accuracy: 1.0000
Epoch 90/100
0.9947 - val loss: 0.0759 - val accuracy: 1.0000
Epoch 91/100
0.9947 - val_loss: 0.1648 - val_accuracy: 1.0000
Epoch 92/100
0.9947 - val_loss: 0.0938 - val_accuracy: 1.0000
Epoch 93/100
0.9893 - val_loss: 0.0793 - val_accuracy: 1.0000
Epoch 94/100
0.9947 - val loss: 0.0761 - val accuracy: 1.0000
0.9947 - val_loss: 0.2060 - val_accuracy: 0.9048
Epoch 96/100
0.9893 - val loss: 0.1904 - val accuracy: 1.0000
Epoch 97/100
0.9947 - val_loss: 0.2852 - val_accuracy: 0.8095
Epoch 98/100
0.9893 - val_loss: 0.1826 - val_accuracy: 1.0000
Epoch 99/100
0.9947 - val_loss: 0.1612 - val_accuracy: 1.0000
Epoch 100/100
0.9947 - val_loss: 0.0861 - val_accuracy: 1.0000
```

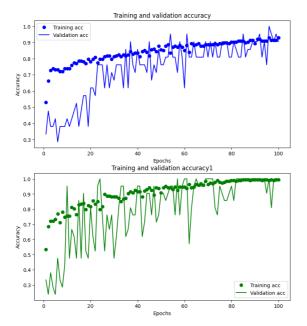
```
In [41]: loss = H.history['loss']
         val_loss = H.history['val_loss']
         acc = H.history['accuracy']
         val_acc = H.history['val_accuracy']
         epochs = range(1, len(loss) + 1)
         loss2 = H2.history['loss']
         val_loss2 = H2.history['val_loss']
         acc2 = H2.history['accuracy']
         val_acc2 = H2.history['val_accuracy']
         epochs2 = range(1, len(loss2) + 1)
         loss3 = H3.history['loss']
         val_loss3 = H3.history['val_loss']
         acc3 = H3.history['accuracy']
         val_acc3 = H3.history['val_accuracy']
         epochs3 = range(1, len(loss3) + 1)
In [42]: plt.figure(figsize=(20, 10))
         plt.subplot(2, 2, 1)
         plt.plot(epochs, loss, 'bo', label='Training loss')
         plt.plot(epochs, val_loss, 'b', label='Validation loss')
         plt.title('Training and validation loss')
         plt.xlabel('Epochs')
         plt.ylabel('Loss')
         plt.legend()
         plt.subplot(2, 2, 2)
         plt.plot(epochs2, loss2, 'yo', label='Training loss')
         plt.plot(epochs2, val_loss2, 'y', label='Validation loss')
         plt.title('Training and validation loss')
         plt.xlabel('Epochs')
         plt.ylabel('Loss')
         plt.legend()
         plt.subplot(2, 2, 3)
         plt.plot(epochs3, loss3, 'go', label='Training loss')
         plt.plot(epochs3, val_loss3, 'g', label='Validation loss')
         plt.title('Training and validation loss')
         plt.xlabel('Epochs')
         plt.ylabel('Loss')
         plt.legend()
         plt.show()
```

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```
In [43]: #Построение графика точности
         plt.figure(figsize=(20, 10))
         plt.subplot(2, 2, 1)
         plt.plot(epochs, acc, 'bo', label='Training acc')
         plt.plot(epochs, val_acc, 'b', label='Validation acc')
         plt.title('Training and validation accuracy')
         plt.xlabel('Epochs')
         plt.ylabel('Accuracy')
         plt.legend()
         plt.subplot(2, 2, 2)
         plt.plot(epochs2, acc2, 'yo', label='Training acc')
         plt.plot(epochs2, val_acc2, 'y', label='Validation acc')
         plt.title('Training and validation accuracy1')
         plt.xlabel('Epochs')
         plt.ylabel('Accuracy')
         plt.legend()
         plt.subplot(2, 2, 3)
         plt.plot(epochs3, acc3, 'go', label='Training acc')
         plt.plot(epochs3, val_acc3, 'g', label='Validation acc')
         plt.title('Training and validation accuracy1')
         plt.xlabel('Epochs')
         plt.ylabel('Accuracy')
         plt.legend()
         plt.show()
```

16.10.2023, 16:36 Untitled4-Copy5





Вывод: мы изучили влияние количества нейронов на слое и количества слоев на результат обучения модели, а так же построили графики ошибок и точности в ходе обучения и сравнили полученные сети. Количество слоев и нейронов в нейронной сети имеет двустороннее влияние на ее эффективность. Более глубокая и широкая сеть может позволить решать более сложные задачи, но требует более тщательной тренировки и может быть более подвержена переобучению. Поэтому выбор количества слоев и неронов в нейронной сети должен быть основан на анализе конкретной задачи и доступных ресурсов. доступных ресурсов.