

Import neccessery libraries

```
In [74]: import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.metrics import confusion_matrix
from keras.models import Sequential
from keras.layers import Dense, Flatten, Activation, Layer, Lambda
import seaborn as sns
from sklearn import preprocessing
from keras.layers import Dropout
from keras import regularizers
from sklearn.model_selection import train_test_split
from matplotlib import pyplot as plt
from sklearn.decomposition import PCA
from mlxtend.plotting import plot_decision_regions
from tensorflow import keras
import warnings
warnings.filterwarnings('ignore')
```

Problem

PREDICT THE BURNED AREA OF FOREST FIRES WITH NEURAL NETWORKS

Import data

```
In [2]: forest_data = pd.read_csv('forestfires.csv')
forest_data
```

```
Out[2]:
```

	month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain	...	monthfeb	monthjan	m
0	mar	fri	86.2	26.2	94.3	5.1	8.2	51	6.7	0.0	...	0	0	
1	oct	tue	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	...	0	0	
2	oct	sat	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0	...	0	0	
3	mar	fri	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2	...	0	0	
4	mar	sun	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0	...	0	0	
...
512	aug	sun	81.6	56.7	665.6	1.9	27.8	32	2.7	0.0	...	0	0	
513	aug	sun	81.6	56.7	665.6	1.9	21.9	71	5.8	0.0	...	0	0	
514	aug	sun	81.6	56.7	665.6	1.9	21.2	70	6.7	0.0	...	0	0	
515	aug	sat	94.4	146.0	614.7	11.3	25.6	42	4.0	0.0	...	0	0	
516	nov	tue	79.5	3.0	106.7	1.1	11.8	31	4.5	0.0	...	0	0	

517 rows × 31 columns

```
In [3]: forest_data.drop(["month", "day"], axis=1, inplace = True)
```

```
In [4]: forest_data["size_category"].value_counts()
```

```
Out[4]: small      378  
large      139  
Name: size_category, dtype: int64
```

```
In [5]: forest_data.isnull().sum()
```

```
Out[5]: FPMC      0  
DMC      0  
DC      0  
ISI      0  
temp      0  
RH      0  
wind      0  
rain      0  
area      0  
dayfri      0  
daymon      0  
daysat      0  
daysun      0  
daythu      0  
daytue      0  
daywed      0  
monthapr      0  
monthaug      0  
monthdec      0  
monthfeb      0  
monthjan      0  
monthjul      0  
monthjun      0  
monthmar      0  
monthmay      0  
monthnov      0  
monthoct      0  
monthsep      0  
size_category      0  
dtype: int64
```

```
In [6]: forest_data.shape
```

```
Out[6]: (517, 29)
```

```
In [7]: df = forest_data.describe().T  
df
```

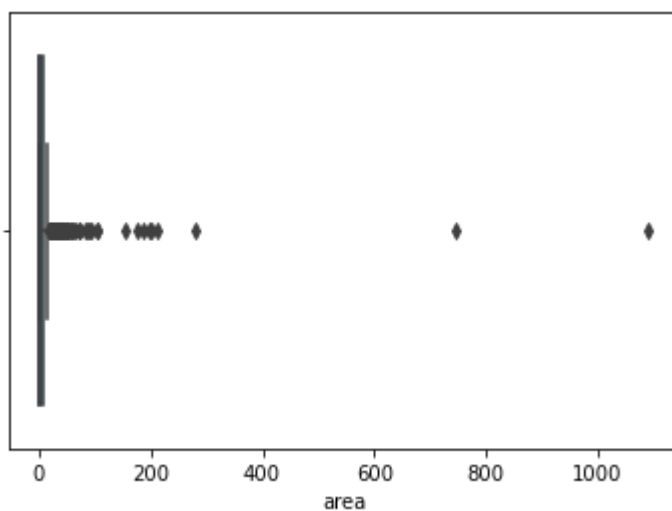
```
Out[7]:
```

	count	mean	std	min	25%	50%	75%	max
FFMC	517.0	90.644681	5.520111	18.7	90.2	91.60	92.90	96.20
DMC	517.0	110.872340	64.046482	1.1	68.6	108.30	142.40	291.30
DC	517.0	547.940039	248.066192	7.9	437.7	664.20	713.90	860.60
ISI	517.0	9.021663	4.559477	0.0	6.5	8.40	10.80	56.10
temp	517.0	18.889168	5.806625	2.2	15.5	19.30	22.80	33.30
RH	517.0	44.288201	16.317469	15.0	33.0	42.00	53.00	100.00

	count	mean	std	min	25%	50%	75%	max
wind	517.0	4.017602	1.791653	0.4	2.7	4.00	4.90	9.40
rain	517.0	0.021663	0.295959	0.0	0.0	0.00	0.00	6.40
area	517.0	12.847292	63.655818	0.0	0.0	0.52	6.57	1090.84
dayfri	517.0	0.164410	0.371006	0.0	0.0	0.00	0.00	1.00
daymon	517.0	0.143133	0.350548	0.0	0.0	0.00	0.00	1.00
daysat	517.0	0.162476	0.369244	0.0	0.0	0.00	0.00	1.00
daysun	517.0	0.183752	0.387657	0.0	0.0	0.00	0.00	1.00
daythu	517.0	0.117988	0.322907	0.0	0.0	0.00	0.00	1.00
daytue	517.0	0.123791	0.329662	0.0	0.0	0.00	0.00	1.00
daywed	517.0	0.104449	0.306138	0.0	0.0	0.00	0.00	1.00
monthapr	517.0	0.017408	0.130913	0.0	0.0	0.00	0.00	1.00
monthaug	517.0	0.355899	0.479249	0.0	0.0	0.00	1.00	1.00
monthdec	517.0	0.017408	0.130913	0.0	0.0	0.00	0.00	1.00
monthfeb	517.0	0.038685	0.193029	0.0	0.0	0.00	0.00	1.00
monthjan	517.0	0.003868	0.062137	0.0	0.0	0.00	0.00	1.00
monthjul	517.0	0.061896	0.241199	0.0	0.0	0.00	0.00	1.00
monthjun	517.0	0.032882	0.178500	0.0	0.0	0.00	0.00	1.00
monthmar	517.0	0.104449	0.306138	0.0	0.0	0.00	0.00	1.00
monthmay	517.0	0.003868	0.062137	0.0	0.0	0.00	0.00	1.00
monthnov	517.0	0.001934	0.043980	0.0	0.0	0.00	0.00	1.00
monthoct	517.0	0.029014	0.168007	0.0	0.0	0.00	0.00	1.00

Outlier Check

```
In [8]: ax = sns.boxplot(forest_data['area'])
```

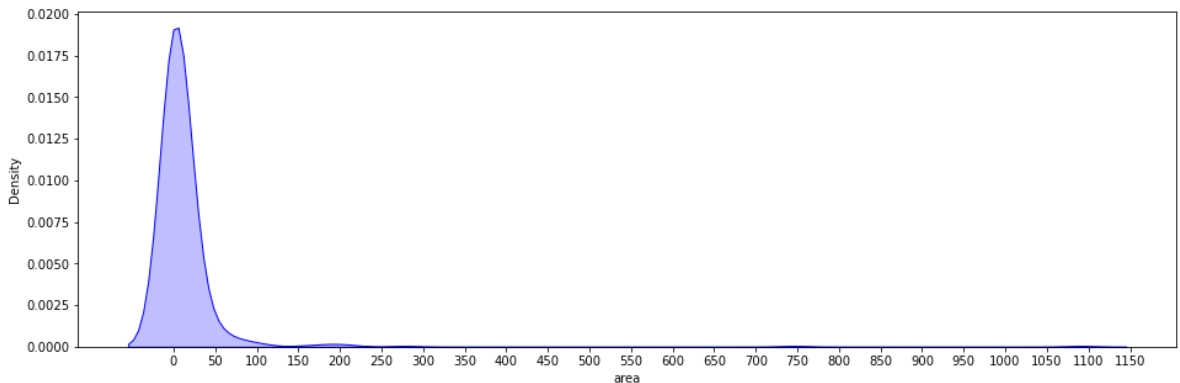


There are 3 Outlier instances in our data

```
In [9]: plt.rcParams["figure.figsize"] = 9,5
```

```
In [10]: plt.figure(figsize=(16,5))
print("Skew: {}".format(forest_data['area'].skew()))
print("Kurtosis: {}".format(forest_data['area'].kurtosis()))
ax = sns.kdeplot(forest_data['area'], shade=True, color='b')
plt.xticks([i for i in range(0,1200,50)])
plt.show()
```

```
Skew: 12.846933533934868
Kurtosis: 194.1407210942299
```



The Data is highly skewed and has large kurtosis value. Majority of the forest fires do not cover a large area, most of the damaged area is under 100 hectares of land

```
In [11]: dfa = forest_data[forest_data.columns[0:10]]
month_colum = dfa.select_dtypes(include='object').columns.tolist()
```

```
In [12]: plt.figure(figsize=(16,10))
for i,col in enumerate(month_colum,1):
    plt.subplot(2,2,i)
    sns.countplot(data=dfa,y=col)
    plt.subplot(2,2,i+2)
    forest_data[col].value_counts(normalize=True).plot.bar()
    plt.ylabel(col)
    plt.xlabel('% distribution per category')
plt.tight_layout()
plt.show()
```

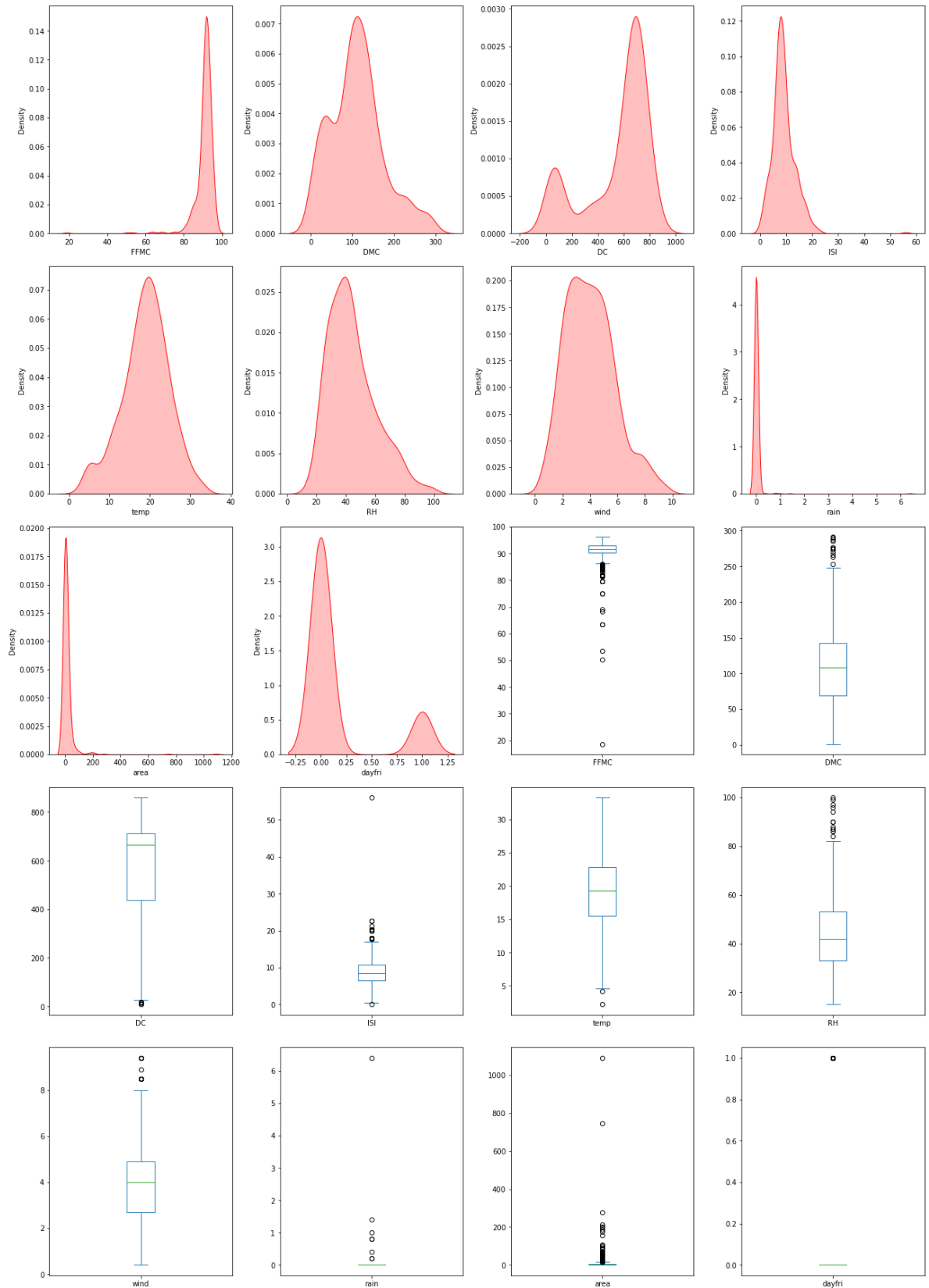
<Figure size 1152x720 with 0 Axes>

Majority of the fire accors in the month August and September
For Days Sunday and Friday have recoreded the most cases

```
In [13]: num_columns = dfa.select_dtypes(exclude='object').columns.tolist()
```

In [14]:

```
plt.figure(figsize=(18,40))
for i,col in enumerate(num_columns,1):
    plt.subplot(8,4,i)
    sns.kdeplot(forest_data[col],color='r',shade=True)
    plt.subplot(8,4,i+10)
    forest_data[col].plot.box()
plt.tight_layout()
plt.show()
num_data = forest_data[num_columns]
pd.DataFrame(data=[num_data.skew(),num_data.kurtosis()],index=['skewness',
```



```
Out[14]:
```

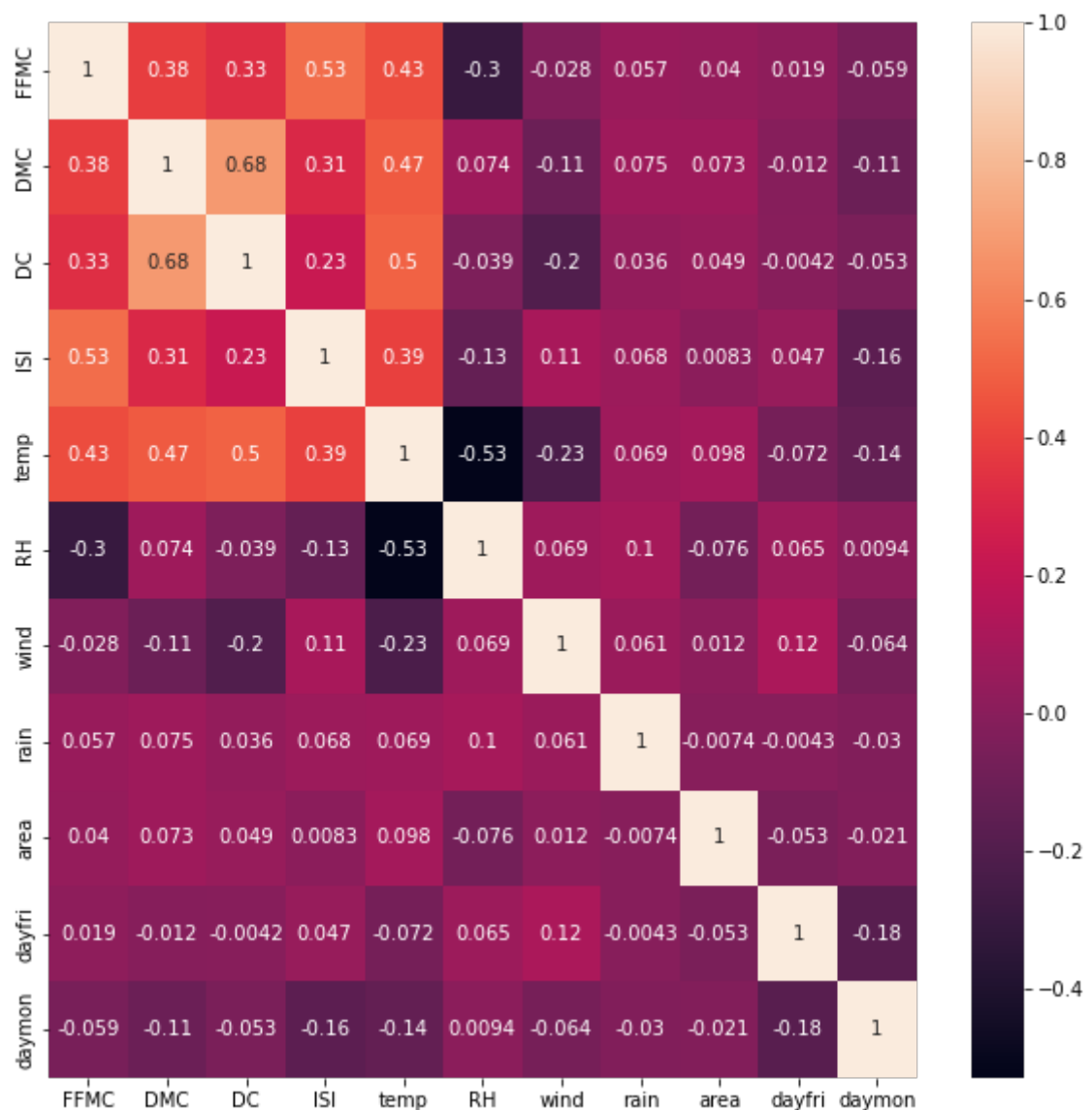
	FFMC	DMC	DC	ISI	temp	RH	wind	rain	
skewness	-6.575606	0.547498	-1.100445	2.536325	-0.331172	0.862904	0.571001	19.816344	
kurtosis	67.066041	0.204822	-0.245244	21.458037	0.136166	0.438183	0.054324	421.295964	1

correlation

```
In [15]: corr = forest_data[forest_data.columns[0:11]].corr()
```

```
In [16]: plt.figure(figsize=(10,10))
sns.heatmap(corr,annot=True)
```

```
Out[16]: <AxesSubplot:>
```



Neural Network Model

```
In [17]: mapping = {'small': 1, 'large': 2}
```

```
In [18]: df1 = forest_data.replace(mapping)
```

```
In [19]: dataset = df1.values
```

```
In [20]: dataset
```

```
Out[20]: array([[ 86.2,  26.2,  94.3, ...,  0. ,  0. ,  1. ],
 [ 90.6,  35.4, 669.1, ...,  1. ,  0. ,  1. ],
 [ 90.6,  43.7, 686.9, ...,  1. ,  0. ,  1. ],
 ...,
 [ 81.6,  56.7, 665.6, ...,  0. ,  0. ,  2. ],
 [ 94.4, 146. , 614.7, ...,  0. ,  0. ,  1. ],
 [ 79.5,   3. , 106.7, ...,  0. ,  0. ,  1. ]])
```

```
In [52]: X = dataset[:,0:10]
Y = dataset[:,10]
```

```
In [53]: min_max_scaler = preprocessing.MinMaxScaler()
X_scale = min_max_scaler.fit_transform(X)
```

```
In [54]: X_scale
```

```
Out[54]: array([[0.87096774, 0.08649207, 0.1013252 , ..., 0. , 0. ,
 1. ,
 [0.92774194, 0.11819435, 0.77541926, ..., 0. , 0. ,
 0. ],
 [0.92774194, 0.14679531, 0.79629412, ..., 0. , 0. ,
 0. ],
 ...,
 [0.8116129 , 0.19159201, 0.77131465, ..., 0. , 0.01023065,
 0. ],
 [0.97677419, 0.49931082, 0.71162191, ..., 0. , 0. ,
 0. ],
 [0.78451613, 0.00654721, 0.11586725, ..., 0. , 0. ,
 0. ]])
```

```
In [55]: X_train, X_val_and_test, Y_train, Y_val_and_test = train_test_split(X_scale,
```

```
In [57]: X_val, X_test, Y_val, Y_test = train_test_split(X_val_and_test, Y_val_and_t
print(X_train.shape, X_val.shape, X_test.shape, Y_train.shape, Y_val.shape,

(361, 10) (78, 10) (78, 10) (361,) (78,) (78,)
```

```
In [58]: model = Sequential([
    Dense(32, activation='relu', input_shape=(10,)),
    Dense(32, activation='relu'),
    Dense(1, activation='sigmoid'),
])
```

```
In [59]: model.compile(optimizer='sgd',  
                    loss='binary_crossentropy',  
                    metrics=['accuracy'])
```

```
In [60]: hist = model.fit(X_train, Y_train,  
                        batch_size=32, epochs=100,  
                        validation_data=(X_val, Y_val))
```

```
Epoch 1/100  
12/12 [=====] - 0s 9ms/step - loss: 0.6599 - accur  
acy: 0.7479 - val_loss: 0.6241 - val_accuracy: 0.8333  
Epoch 2/100  
12/12 [=====] - 0s 2ms/step - loss: 0.6080 - accur  
acy: 0.8504 - val_loss: 0.5780 - val_accuracy: 0.8846  
Epoch 3/100  
12/12 [=====] - 0s 2ms/step - loss: 0.5692 - accur  
acy: 0.8560 - val_loss: 0.5356 - val_accuracy: 0.8846  
Epoch 4/100  
12/12 [=====] - 0s 2ms/step - loss: 0.5337 - accur  
acy: 0.8560 - val_loss: 0.5017 - val_accuracy: 0.8846  
Epoch 5/100  
12/12 [=====] - 0s 3ms/step - loss: 0.5060 - accur  
acy: 0.8560 - val_loss: 0.4730 - val_accuracy: 0.8846  
Epoch 6/100  
12/12 [=====] - 0s 3ms/step - loss: 0.4826 - accur  
acy: 0.8560 - val_loss: 0.4521 - val_accuracy: 0.8846  
Epoch 7/100  
12/12 [=====] - 0s 2ms/step - loss: 0.4660 - accur  
acy: 0.8560 - val_loss: 0.4326 - val_accuracy: 0.8846  
Epoch 8/100  
12/12 [=====] - 0s 2ms/step - loss: 0.4508 - accur  
acy: 0.8560 - val_loss: 0.4170 - val_accuracy: 0.8846  
Epoch 9/100  
12/12 [=====] - 0s 2ms/step - loss: 0.4390 - accur  
acy: 0.8560 - val_loss: 0.4043 - val_accuracy: 0.8846  
Epoch 10/100  
12/12 [=====] - 0s 2ms/step - loss: 0.4298 - accur  
acy: 0.8560 - val_loss: 0.3948 - val_accuracy: 0.8846  
Epoch 11/100  
12/12 [=====] - 0s 2ms/step - loss: 0.4230 - accur  
acy: 0.8560 - val_loss: 0.3866 - val_accuracy: 0.8846  
Epoch 12/100  
12/12 [=====] - 0s 2ms/step - loss: 0.4176 - accur  
acy: 0.8560 - val_loss: 0.3806 - val_accuracy: 0.8846  
Epoch 13/100  
12/12 [=====] - 0s 3ms/step - loss: 0.4137 - accur  
acy: 0.8560 - val_loss: 0.3757 - val_accuracy: 0.8846  
Epoch 14/100  
12/12 [=====] - 0s 2ms/step - loss: 0.4107 - accur  
acy: 0.8560 - val_loss: 0.3719 - val_accuracy: 0.8846  
Epoch 15/100  
12/12 [=====] - 0s 2ms/step - loss: 0.4084 - accur  
acy: 0.8560 - val_loss: 0.3687 - val_accuracy: 0.8846  
Epoch 16/100  
12/12 [=====] - 0s 3ms/step - loss: 0.4066 - accur  
acy: 0.8560 - val_loss: 0.3665 - val_accuracy: 0.8846  
Epoch 17/100  
12/12 [=====] - 0s 5ms/step - loss: 0.4056 - accur  
acy: 0.8560 - val_loss: 0.3641 - val_accuracy: 0.8846  
Epoch 18/100
```


12/12 [=====] - 0s 3ms/step - loss: 0.4044 - accuracy: 0.8560 - val_loss: 0.3624 - val_accuracy: 0.8846
Epoch 19/100
12/12 [=====] - 0s 3ms/step - loss: 0.4034 - accuracy: 0.8560 - val_loss: 0.3614 - val_accuracy: 0.8846
Epoch 20/100
12/12 [=====] - 0s 2ms/step - loss: 0.4030 - accuracy: 0.8560 - val_loss: 0.3605 - val_accuracy: 0.8846
Epoch 21/100
12/12 [=====] - 0s 2ms/step - loss: 0.4026 - accuracy: 0.8560 - val_loss: 0.3600 - val_accuracy: 0.8846
Epoch 22/100
12/12 [=====] - 0s 2ms/step - loss: 0.4022 - accuracy: 0.8560 - val_loss: 0.3587 - val_accuracy: 0.8846
Epoch 23/100
12/12 [=====] - 0s 3ms/step - loss: 0.4017 - accuracy: 0.8560 - val_loss: 0.3587 - val_accuracy: 0.8846
Epoch 24/100
12/12 [=====] - 0s 3ms/step - loss: 0.4017 - accuracy: 0.8560 - val_loss: 0.3581 - val_accuracy: 0.8846
Epoch 25/100
12/12 [=====] - 0s 2ms/step - loss: 0.4014 - accuracy: 0.8560 - val_loss: 0.3576 - val_accuracy: 0.8846
Epoch 26/100
12/12 [=====] - 0s 2ms/step - loss: 0.4011 - accuracy: 0.8560 - val_loss: 0.3566 - val_accuracy: 0.8846
Epoch 27/100
12/12 [=====] - 0s 3ms/step - loss: 0.4007 - accuracy: 0.8560 - val_loss: 0.3561 - val_accuracy: 0.8846
Epoch 28/100
12/12 [=====] - 0s 3ms/step - loss: 0.4005 - accuracy: 0.8560 - val_loss: 0.3557 - val_accuracy: 0.8846
Epoch 29/100
12/12 [=====] - 0s 3ms/step - loss: 0.4003 - accuracy: 0.8560 - val_loss: 0.3552 - val_accuracy: 0.8846
Epoch 30/100
12/12 [=====] - 0s 2ms/step - loss: 0.4002 - accuracy: 0.8560 - val_loss: 0.3547 - val_accuracy: 0.8846
Epoch 31/100
12/12 [=====] - 0s 2ms/step - loss: 0.3998 - accuracy: 0.8560 - val_loss: 0.3544 - val_accuracy: 0.8846
Epoch 32/100
12/12 [=====] - 0s 2ms/step - loss: 0.3997 - accuracy: 0.8560 - val_loss: 0.3546 - val_accuracy: 0.8846
Epoch 33/100
12/12 [=====] - 0s 2ms/step - loss: 0.3996 - accuracy: 0.8560 - val_loss: 0.3543 - val_accuracy: 0.8846
Epoch 34/100
12/12 [=====] - 0s 2ms/step - loss: 0.3994 - accuracy: 0.8560 - val_loss: 0.3537 - val_accuracy: 0.8846
Epoch 35/100
12/12 [=====] - 0s 3ms/step - loss: 0.3992 - accuracy: 0.8560 - val_loss: 0.3533 - val_accuracy: 0.8846
Epoch 36/100
12/12 [=====] - 0s 3ms/step - loss: 0.3991 - accuracy: 0.8560 - val_loss: 0.3528 - val_accuracy: 0.8846
Epoch 37/100
12/12 [=====] - 0s 3ms/step - loss: 0.3990 - accuracy: 0.8560 - val_loss: 0.3527 - val_accuracy: 0.8846
Epoch 38/100
12/12 [=====] - 0s 2ms/step - loss: 0.3989 - accuracy: 0.8560 - val_loss: 0.3523 - val_accuracy: 0.8846
Epoch 39/100

12/12 [=====] - 0s 2ms/step - loss: 0.3987 - accuracy: 0.8560 - val_loss: 0.3524 - val_accuracy: 0.8846
Epoch 40/100
12/12 [=====] - 0s 2ms/step - loss: 0.3985 - accuracy: 0.8560 - val_loss: 0.3523 - val_accuracy: 0.8846
Epoch 41/100
12/12 [=====] - 0s 2ms/step - loss: 0.3984 - accuracy: 0.8560 - val_loss: 0.3523 - val_accuracy: 0.8846
Epoch 42/100
12/12 [=====] - 0s 2ms/step - loss: 0.3982 - accuracy: 0.8560 - val_loss: 0.3523 - val_accuracy: 0.8846
Epoch 43/100
12/12 [=====] - 0s 2ms/step - loss: 0.3982 - accuracy: 0.8560 - val_loss: 0.3521 - val_accuracy: 0.8846
Epoch 44/100
12/12 [=====] - 0s 2ms/step - loss: 0.3982 - accuracy: 0.8560 - val_loss: 0.3519 - val_accuracy: 0.8846
Epoch 45/100
12/12 [=====] - 0s 2ms/step - loss: 0.3979 - accuracy: 0.8560 - val_loss: 0.3520 - val_accuracy: 0.8846
Epoch 46/100
12/12 [=====] - 0s 3ms/step - loss: 0.3979 - accuracy: 0.8560 - val_loss: 0.3516 - val_accuracy: 0.8846
Epoch 47/100
12/12 [=====] - 0s 2ms/step - loss: 0.3976 - accuracy: 0.8560 - val_loss: 0.3514 - val_accuracy: 0.8846
Epoch 48/100
12/12 [=====] - 0s 2ms/step - loss: 0.3977 - accuracy: 0.8560 - val_loss: 0.3515 - val_accuracy: 0.8846
Epoch 49/100
12/12 [=====] - 0s 3ms/step - loss: 0.3973 - accuracy: 0.8560 - val_loss: 0.3510 - val_accuracy: 0.8846
Epoch 50/100
12/12 [=====] - 0s 3ms/step - loss: 0.3973 - accuracy: 0.8560 - val_loss: 0.3511 - val_accuracy: 0.8846
Epoch 51/100
12/12 [=====] - 0s 3ms/step - loss: 0.3971 - accuracy: 0.8560 - val_loss: 0.3507 - val_accuracy: 0.8846
Epoch 52/100
12/12 [=====] - 0s 2ms/step - loss: 0.3970 - accuracy: 0.8560 - val_loss: 0.3506 - val_accuracy: 0.8846
Epoch 53/100
12/12 [=====] - 0s 2ms/step - loss: 0.3968 - accuracy: 0.8560 - val_loss: 0.3509 - val_accuracy: 0.8846
Epoch 54/100
12/12 [=====] - 0s 2ms/step - loss: 0.3967 - accuracy: 0.8560 - val_loss: 0.3509 - val_accuracy: 0.8846
Epoch 55/100
12/12 [=====] - 0s 2ms/step - loss: 0.3964 - accuracy: 0.8560 - val_loss: 0.3505 - val_accuracy: 0.8846
Epoch 56/100
12/12 [=====] - 0s 2ms/step - loss: 0.3964 - accuracy: 0.8560 - val_loss: 0.3507 - val_accuracy: 0.8846
Epoch 57/100
12/12 [=====] - 0s 2ms/step - loss: 0.3962 - accuracy: 0.8560 - val_loss: 0.3503 - val_accuracy: 0.8846
Epoch 58/100
12/12 [=====] - 0s 2ms/step - loss: 0.3961 - accuracy: 0.8560 - val_loss: 0.3505 - val_accuracy: 0.8846
Epoch 59/100
12/12 [=====] - 0s 3ms/step - loss: 0.3960 - accuracy: 0.8560 - val_loss: 0.3502 - val_accuracy: 0.8846
Epoch 60/100

12/12 [=====] - 0s 2ms/step - loss: 0.3960 - accuracy: 0.8560 - val_loss: 0.3498 - val_accuracy: 0.8846
Epoch 61/100
12/12 [=====] - 0s 2ms/step - loss: 0.3957 - accuracy: 0.8560 - val_loss: 0.3494 - val_accuracy: 0.8846
Epoch 62/100
12/12 [=====] - 0s 2ms/step - loss: 0.3958 - accuracy: 0.8560 - val_loss: 0.3495 - val_accuracy: 0.8846
Epoch 63/100
12/12 [=====] - 0s 2ms/step - loss: 0.3954 - accuracy: 0.8560 - val_loss: 0.3495 - val_accuracy: 0.8846
Epoch 64/100
12/12 [=====] - 0s 2ms/step - loss: 0.3953 - accuracy: 0.8560 - val_loss: 0.3500 - val_accuracy: 0.8846
Epoch 65/100
12/12 [=====] - 0s 2ms/step - loss: 0.3952 - accuracy: 0.8560 - val_loss: 0.3495 - val_accuracy: 0.8846
Epoch 66/100
12/12 [=====] - 0s 2ms/step - loss: 0.3951 - accuracy: 0.8560 - val_loss: 0.3493 - val_accuracy: 0.8846
Epoch 67/100
12/12 [=====] - 0s 2ms/step - loss: 0.3948 - accuracy: 0.8560 - val_loss: 0.3492 - val_accuracy: 0.8846
Epoch 68/100
12/12 [=====] - 0s 2ms/step - loss: 0.3948 - accuracy: 0.8560 - val_loss: 0.3492 - val_accuracy: 0.8846
Epoch 69/100
12/12 [=====] - 0s 2ms/step - loss: 0.3946 - accuracy: 0.8560 - val_loss: 0.3489 - val_accuracy: 0.8846
Epoch 70/100
12/12 [=====] - 0s 2ms/step - loss: 0.3945 - accuracy: 0.8560 - val_loss: 0.3484 - val_accuracy: 0.8846
Epoch 71/100
12/12 [=====] - 0s 2ms/step - loss: 0.3944 - accuracy: 0.8560 - val_loss: 0.3482 - val_accuracy: 0.8846
Epoch 72/100
12/12 [=====] - 0s 2ms/step - loss: 0.3942 - accuracy: 0.8560 - val_loss: 0.3480 - val_accuracy: 0.8846
Epoch 73/100
12/12 [=====] - 0s 2ms/step - loss: 0.3943 - accuracy: 0.8560 - val_loss: 0.3482 - val_accuracy: 0.8846
Epoch 74/100
12/12 [=====] - 0s 2ms/step - loss: 0.3938 - accuracy: 0.8560 - val_loss: 0.3479 - val_accuracy: 0.8846
Epoch 75/100
12/12 [=====] - 0s 2ms/step - loss: 0.3938 - accuracy: 0.8560 - val_loss: 0.3478 - val_accuracy: 0.8846
Epoch 76/100
12/12 [=====] - 0s 2ms/step - loss: 0.3935 - accuracy: 0.8560 - val_loss: 0.3475 - val_accuracy: 0.8846
Epoch 77/100
12/12 [=====] - 0s 2ms/step - loss: 0.3933 - accuracy: 0.8560 - val_loss: 0.3477 - val_accuracy: 0.8846
Epoch 78/100
12/12 [=====] - 0s 2ms/step - loss: 0.3933 - accuracy: 0.8560 - val_loss: 0.3474 - val_accuracy: 0.8846
Epoch 79/100
12/12 [=====] - 0s 2ms/step - loss: 0.3931 - accuracy: 0.8560 - val_loss: 0.3471 - val_accuracy: 0.8846
Epoch 80/100
12/12 [=====] - 0s 2ms/step - loss: 0.3931 - accuracy: 0.8560 - val_loss: 0.3469 - val_accuracy: 0.8846
Epoch 81/100

```

12/12 [=====] - 0s 2ms/step - loss: 0.3927 - accur
acy: 0.8560 - val_loss: 0.3466 - val_accuracy: 0.8846
Epoch 82/100
12/12 [=====] - 0s 2ms/step - loss: 0.3927 - accur
acy: 0.8560 - val_loss: 0.3461 - val_accuracy: 0.8846
Epoch 83/100
12/12 [=====] - 0s 2ms/step - loss: 0.3925 - accur
acy: 0.8560 - val_loss: 0.3461 - val_accuracy: 0.8846
Epoch 84/100
12/12 [=====] - 0s 2ms/step - loss: 0.3923 - accur
acy: 0.8560 - val_loss: 0.3461 - val_accuracy: 0.8846
Epoch 85/100
12/12 [=====] - 0s 2ms/step - loss: 0.3922 - accur
acy: 0.8560 - val_loss: 0.3460 - val_accuracy: 0.8846
Epoch 86/100
12/12 [=====] - 0s 2ms/step - loss: 0.3920 - accur
acy: 0.8560 - val_loss: 0.3460 - val_accuracy: 0.8846
Epoch 87/100
12/12 [=====] - 0s 2ms/step - loss: 0.3920 - accur
acy: 0.8560 - val_loss: 0.3462 - val_accuracy: 0.8846
Epoch 88/100
12/12 [=====] - 0s 2ms/step - loss: 0.3920 - accur
acy: 0.8560 - val_loss: 0.3459 - val_accuracy: 0.8846
Epoch 89/100
12/12 [=====] - 0s 2ms/step - loss: 0.3918 - accur
acy: 0.8560 - val_loss: 0.3458 - val_accuracy: 0.8846
Epoch 90/100
12/12 [=====] - 0s 2ms/step - loss: 0.3916 - accur
acy: 0.8560 - val_loss: 0.3458 - val_accuracy: 0.8846
Epoch 91/100
12/12 [=====] - 0s 2ms/step - loss: 0.3915 - accur
acy: 0.8560 - val_loss: 0.3461 - val_accuracy: 0.8846
Epoch 92/100
12/12 [=====] - 0s 2ms/step - loss: 0.3913 - accur
acy: 0.8560 - val_loss: 0.3461 - val_accuracy: 0.8846
Epoch 93/100
12/12 [=====] - 0s 2ms/step - loss: 0.3914 - accur
acy: 0.8560 - val_loss: 0.3454 - val_accuracy: 0.8846
Epoch 94/100
12/12 [=====] - 0s 2ms/step - loss: 0.3911 - accur
acy: 0.8560 - val_loss: 0.3451 - val_accuracy: 0.8846
Epoch 95/100
12/12 [=====] - 0s 2ms/step - loss: 0.3910 - accur
acy: 0.8560 - val_loss: 0.3452 - val_accuracy: 0.8846
Epoch 96/100
12/12 [=====] - 0s 2ms/step - loss: 0.3908 - accur
acy: 0.8560 - val_loss: 0.3452 - val_accuracy: 0.8846
Epoch 97/100
12/12 [=====] - 0s 2ms/step - loss: 0.3907 - accur
acy: 0.8560 - val_loss: 0.3452 - val_accuracy: 0.8846
Epoch 98/100
12/12 [=====] - 0s 2ms/step - loss: 0.3907 - accur
acy: 0.8560 - val_loss: 0.3448 - val_accuracy: 0.8846
Epoch 99/100
12/12 [=====] - 0s 2ms/step - loss: 0.3905 - accur
acy: 0.8560 - val_loss: 0.3447 - val_accuracy: 0.8846
Epoch 100/100

```

```
In [61]: model.evaluate(X_test, Y_test)[1]
```

```

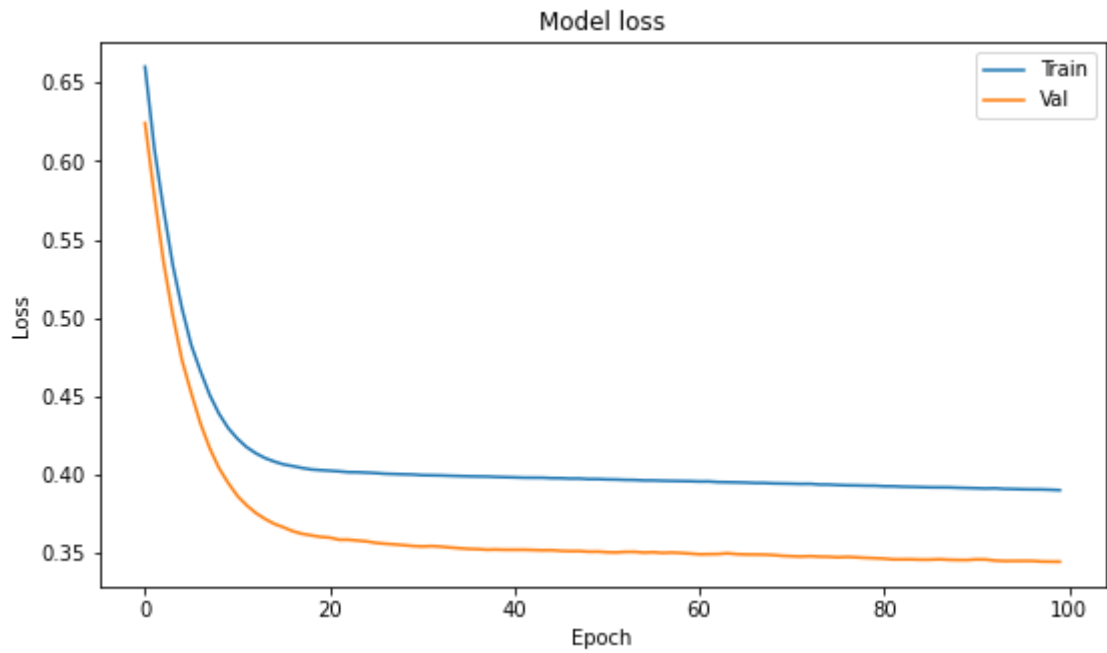
3/3 [=====] - 0s 997us/step - loss: 0.4276 - accur
acy: 0.8333

```

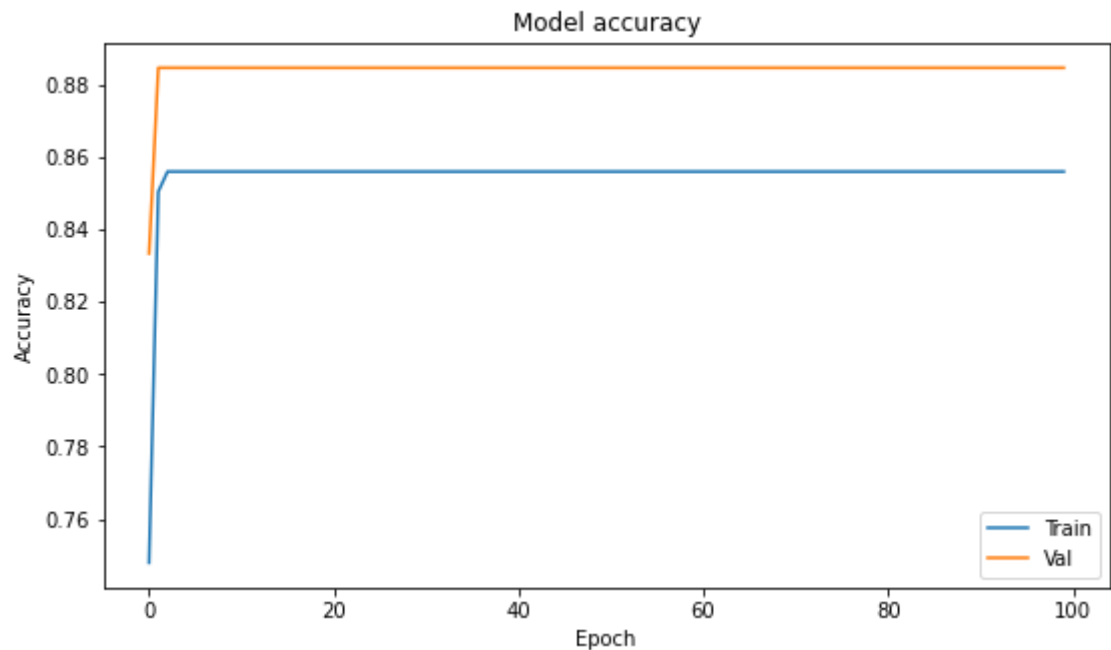
Out[61]: 0.8333333134651184

Visualizing Loss and Accuracy

```
In [66]: plt.plot(hist.history['loss'])
plt.plot(hist.history['val_loss'])
plt.title('Model loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['Train', 'Val'], loc='upper right')
plt.show()
```



```
In [69]: plt.plot(hist.history['accuracy'])
plt.plot(hist.history['val_accuracy'])
plt.title('Model accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(['Train', 'Val'], loc='lower right')
plt.show()
```



Adding Regularization to our Neural Network

In [70]:

```
model_2 = Sequential([
    Dense(1000, activation='relu', input_shape=(10,)),
    Dense(1000, activation='relu'),
    Dense(1000, activation='relu'),
    Dense(1000, activation='relu'),
    Dense(1, activation='sigmoid'),
])

model_2.compile(optimizer='adam',
                loss='binary_crossentropy',
                metrics=['accuracy'])

hist_2 = model_2.fit(X_train, Y_train,
                    batch_size=32, epochs=100,
                    validation_data=(X_val, Y_val))
```

```
Epoch 1/100
12/12 [=====] - 1s 32ms/step - loss: 0.4316 - accuracy: 0.8560 - val_loss: 0.3390 - val_accuracy: 0.8846
Epoch 2/100
12/12 [=====] - 0s 26ms/step - loss: 0.3928 - accuracy: 0.8560 - val_loss: 0.3187 - val_accuracy: 0.8846
Epoch 3/100
12/12 [=====] - 0s 26ms/step - loss: 0.3771 - accuracy: 0.8560 - val_loss: 0.3126 - val_accuracy: 0.8846
Epoch 4/100
12/12 [=====] - 0s 24ms/step - loss: 0.3684 - accuracy: 0.8560 - val_loss: 0.3125 - val_accuracy: 0.8846
Epoch 5/100
12/12 [=====] - 0s 25ms/step - loss: 0.3931 - accuracy: 0.8560 - val_loss: 0.3088 - val_accuracy: 0.8846
Epoch 6/100
12/12 [=====] - 0s 25ms/step - loss: 0.3665 - accuracy: 0.8560 - val_loss: 0.3059 - val_accuracy: 0.8846
Epoch 7/100
12/12 [=====] - 0s 25ms/step - loss: 0.3571 - accuracy: 0.8560 - val_loss: 0.3045 - val_accuracy: 0.8846
Epoch 8/100
```

12/12 [=====] - 0s 25ms/step - loss: 0.3633 - accuracy: 0.8560 - val_loss: 0.3105 - val_accuracy: 0.8846
Epoch 9/100
12/12 [=====] - 0s 25ms/step - loss: 0.3702 - accuracy: 0.8560 - val_loss: 0.3017 - val_accuracy: 0.8846
Epoch 10/100
12/12 [=====] - 0s 28ms/step - loss: 0.3692 - accuracy: 0.8560 - val_loss: 0.3087 - val_accuracy: 0.8846
Epoch 11/100
12/12 [=====] - 0s 27ms/step - loss: 0.3567 - accuracy: 0.8560 - val_loss: 0.2948 - val_accuracy: 0.8846
Epoch 12/100
12/12 [=====] - 0s 25ms/step - loss: 0.3506 - accuracy: 0.8560 - val_loss: 0.2961 - val_accuracy: 0.8846
Epoch 13/100
12/12 [=====] - 0s 25ms/step - loss: 0.3402 - accuracy: 0.8560 - val_loss: 0.2909 - val_accuracy: 0.8846
Epoch 14/100
12/12 [=====] - 0s 26ms/step - loss: 0.3443 - accuracy: 0.8643 - val_loss: 0.3159 - val_accuracy: 0.8846
Epoch 15/100
12/12 [=====] - 0s 25ms/step - loss: 0.3877 - accuracy: 0.8587 - val_loss: 0.3007 - val_accuracy: 0.8718
Epoch 16/100
12/12 [=====] - 0s 26ms/step - loss: 0.3473 - accuracy: 0.8615 - val_loss: 0.3022 - val_accuracy: 0.8846
Epoch 17/100
12/12 [=====] - 0s 26ms/step - loss: 0.3414 - accuracy: 0.8615 - val_loss: 0.2963 - val_accuracy: 0.8718
Epoch 18/100
12/12 [=====] - 0s 26ms/step - loss: 0.3286 - accuracy: 0.8615 - val_loss: 0.3077 - val_accuracy: 0.8718
Epoch 19/100
12/12 [=====] - 0s 25ms/step - loss: 0.3328 - accuracy: 0.8643 - val_loss: 0.3006 - val_accuracy: 0.8718
Epoch 20/100
12/12 [=====] - 0s 25ms/step - loss: 0.3303 - accuracy: 0.8643 - val_loss: 0.3067 - val_accuracy: 0.8718
Epoch 21/100
12/12 [=====] - 0s 27ms/step - loss: 0.3149 - accuracy: 0.8698 - val_loss: 0.2872 - val_accuracy: 0.8718
Epoch 22/100
12/12 [=====] - 0s 26ms/step - loss: 0.3213 - accuracy: 0.8670 - val_loss: 0.3016 - val_accuracy: 0.8718
Epoch 23/100
12/12 [=====] - 0s 25ms/step - loss: 0.3313 - accuracy: 0.8615 - val_loss: 0.3025 - val_accuracy: 0.8718
Epoch 24/100
12/12 [=====] - 0s 25ms/step - loss: 0.3096 - accuracy: 0.8698 - val_loss: 0.2933 - val_accuracy: 0.8718
Epoch 25/100
12/12 [=====] - 0s 27ms/step - loss: 0.3320 - accuracy: 0.8670 - val_loss: 0.3522 - val_accuracy: 0.8718
Epoch 26/100
12/12 [=====] - 0s 26ms/step - loss: 0.3469 - accuracy: 0.8643 - val_loss: 0.3117 - val_accuracy: 0.8718
Epoch 27/100
12/12 [=====] - 0s 25ms/step - loss: 0.3214 - accuracy: 0.8643 - val_loss: 0.2898 - val_accuracy: 0.8718
Epoch 28/100
12/12 [=====] - 0s 26ms/step - loss: 0.3437 - accuracy: 0.8643 - val_loss: 0.3258 - val_accuracy: 0.8718
Epoch 29/100

12/12 [=====] - 0s 25ms/step - loss: 0.3242 - accuracy: 0.8615 - val_loss: 0.2981 - val_accuracy: 0.8718
Epoch 30/100
12/12 [=====] - 0s 26ms/step - loss: 0.3127 - accuracy: 0.8615 - val_loss: 0.2901 - val_accuracy: 0.8718
Epoch 31/100
12/12 [=====] - 0s 25ms/step - loss: 0.3005 - accuracy: 0.8615 - val_loss: 0.2986 - val_accuracy: 0.8718
Epoch 32/100
12/12 [=====] - 0s 25ms/step - loss: 0.2897 - accuracy: 0.8670 - val_loss: 0.3018 - val_accuracy: 0.8718
Epoch 33/100
12/12 [=====] - 0s 29ms/step - loss: 0.3004 - accuracy: 0.8670 - val_loss: 0.2877 - val_accuracy: 0.8718
Epoch 34/100
12/12 [=====] - 0s 25ms/step - loss: 0.2801 - accuracy: 0.8781 - val_loss: 0.3068 - val_accuracy: 0.8846
Epoch 35/100
12/12 [=====] - 0s 27ms/step - loss: 0.2727 - accuracy: 0.8809 - val_loss: 0.3125 - val_accuracy: 0.8846
Epoch 36/100
12/12 [=====] - 0s 25ms/step - loss: 0.2731 - accuracy: 0.8809 - val_loss: 0.3740 - val_accuracy: 0.8333
Epoch 37/100
12/12 [=====] - 0s 24ms/step - loss: 0.2756 - accuracy: 0.8753 - val_loss: 0.3008 - val_accuracy: 0.8846
Epoch 38/100
12/12 [=====] - 0s 27ms/step - loss: 0.3018 - accuracy: 0.8670 - val_loss: 0.2893 - val_accuracy: 0.8718
Epoch 39/100
12/12 [=====] - 0s 24ms/step - loss: 0.2738 - accuracy: 0.8726 - val_loss: 0.3052 - val_accuracy: 0.8718
Epoch 40/100
12/12 [=====] - 0s 24ms/step - loss: 0.2423 - accuracy: 0.8892 - val_loss: 0.3229 - val_accuracy: 0.8718
Epoch 41/100
12/12 [=====] - 0s 24ms/step - loss: 0.2526 - accuracy: 0.8920 - val_loss: 0.3207 - val_accuracy: 0.8718
Epoch 42/100
12/12 [=====] - 0s 25ms/step - loss: 0.2707 - accuracy: 0.8753 - val_loss: 0.3609 - val_accuracy: 0.8846
Epoch 43/100
12/12 [=====] - 0s 26ms/step - loss: 0.3053 - accuracy: 0.8698 - val_loss: 0.2919 - val_accuracy: 0.8718
Epoch 44/100
12/12 [=====] - 0s 23ms/step - loss: 0.2721 - accuracy: 0.8809 - val_loss: 0.3259 - val_accuracy: 0.8462
Epoch 45/100
12/12 [=====] - 0s 25ms/step - loss: 0.2476 - accuracy: 0.8920 - val_loss: 0.3313 - val_accuracy: 0.8718
Epoch 46/100
12/12 [=====] - 0s 27ms/step - loss: 0.2402 - accuracy: 0.8864 - val_loss: 0.3097 - val_accuracy: 0.8974
Epoch 47/100
12/12 [=====] - 0s 28ms/step - loss: 0.2448 - accuracy: 0.8809 - val_loss: 0.3221 - val_accuracy: 0.8846
Epoch 48/100
12/12 [=====] - 0s 26ms/step - loss: 0.2465 - accuracy: 0.8892 - val_loss: 0.3427 - val_accuracy: 0.8590
Epoch 49/100
12/12 [=====] - 0s 25ms/step - loss: 0.2541 - accuracy: 0.8864 - val_loss: 0.3118 - val_accuracy: 0.8590
Epoch 50/100

12/12 [=====] - 0s 28ms/step - loss: 0.2443 - accuracy: 0.8809 - val_loss: 0.3906 - val_accuracy: 0.8590
Epoch 51/100
12/12 [=====] - 0s 24ms/step - loss: 0.2579 - accuracy: 0.9003 - val_loss: 0.2950 - val_accuracy: 0.8718
Epoch 52/100
12/12 [=====] - 0s 24ms/step - loss: 0.2354 - accuracy: 0.8892 - val_loss: 0.3443 - val_accuracy: 0.8205
Epoch 53/100
12/12 [=====] - 0s 29ms/step - loss: 0.2601 - accuracy: 0.8864 - val_loss: 0.3230 - val_accuracy: 0.8590
Epoch 54/100
12/12 [=====] - 0s 26ms/step - loss: 0.2719 - accuracy: 0.8726 - val_loss: 0.2829 - val_accuracy: 0.9103
Epoch 55/100
12/12 [=====] - 0s 25ms/step - loss: 0.2339 - accuracy: 0.9030 - val_loss: 0.3110 - val_accuracy: 0.8718
Epoch 56/100
12/12 [=====] - 0s 24ms/step - loss: 0.2247 - accuracy: 0.8975 - val_loss: 0.3901 - val_accuracy: 0.8205
Epoch 57/100
12/12 [=====] - 0s 27ms/step - loss: 0.2360 - accuracy: 0.9003 - val_loss: 0.2978 - val_accuracy: 0.8846
Epoch 58/100
12/12 [=====] - 0s 28ms/step - loss: 0.2130 - accuracy: 0.8975 - val_loss: 0.3724 - val_accuracy: 0.8590
Epoch 59/100
12/12 [=====] - 0s 26ms/step - loss: 0.2157 - accuracy: 0.9030 - val_loss: 0.3632 - val_accuracy: 0.8718
Epoch 60/100
12/12 [=====] - 0s 25ms/step - loss: 0.2229 - accuracy: 0.9058 - val_loss: 0.3355 - val_accuracy: 0.8590
Epoch 61/100
12/12 [=====] - 0s 24ms/step - loss: 0.1913 - accuracy: 0.8947 - val_loss: 0.4383 - val_accuracy: 0.8590
Epoch 62/100
12/12 [=====] - 0s 25ms/step - loss: 0.2070 - accuracy: 0.9030 - val_loss: 0.3222 - val_accuracy: 0.8462
Epoch 63/100
12/12 [=====] - 0s 25ms/step - loss: 0.2532 - accuracy: 0.9141 - val_loss: 0.2825 - val_accuracy: 0.8846
Epoch 64/100
12/12 [=====] - 0s 25ms/step - loss: 0.2150 - accuracy: 0.9169 - val_loss: 0.3282 - val_accuracy: 0.8974
Epoch 65/100
12/12 [=====] - 0s 25ms/step - loss: 0.2018 - accuracy: 0.9114 - val_loss: 0.4163 - val_accuracy: 0.8462
Epoch 66/100
12/12 [=====] - 0s 27ms/step - loss: 0.2922 - accuracy: 0.8864 - val_loss: 0.2872 - val_accuracy: 0.8718
Epoch 67/100
12/12 [=====] - 0s 25ms/step - loss: 0.2572 - accuracy: 0.8947 - val_loss: 0.3535 - val_accuracy: 0.8590
Epoch 68/100
12/12 [=====] - 0s 23ms/step - loss: 0.2253 - accuracy: 0.8947 - val_loss: 0.3635 - val_accuracy: 0.8846
Epoch 69/100
12/12 [=====] - 0s 25ms/step - loss: 0.2031 - accuracy: 0.8947 - val_loss: 0.3683 - val_accuracy: 0.8846
Epoch 70/100
12/12 [=====] - 0s 28ms/step - loss: 0.2065 - accuracy: 0.9030 - val_loss: 0.4701 - val_accuracy: 0.8846
Epoch 71/100

12/12 [=====] - 0s 24ms/step - loss: 0.2574 - accuracy: 0.9003 - val_loss: 0.2965 - val_accuracy: 0.8974
Epoch 72/100
12/12 [=====] - 0s 26ms/step - loss: 0.2196 - accuracy: 0.8947 - val_loss: 0.3904 - val_accuracy: 0.8462
Epoch 73/100
12/12 [=====] - 0s 24ms/step - loss: 0.2063 - accuracy: 0.9058 - val_loss: 0.3151 - val_accuracy: 0.8718
Epoch 74/100
12/12 [=====] - 0s 27ms/step - loss: 0.1897 - accuracy: 0.9197 - val_loss: 0.4009 - val_accuracy: 0.8590
Epoch 75/100
12/12 [=====] - 0s 26ms/step - loss: 0.1643 - accuracy: 0.9307 - val_loss: 0.5197 - val_accuracy: 0.8718
Epoch 76/100
12/12 [=====] - 0s 24ms/step - loss: 0.2227 - accuracy: 0.9141 - val_loss: 0.3429 - val_accuracy: 0.8974
Epoch 77/100
12/12 [=====] - 0s 26ms/step - loss: 0.1701 - accuracy: 0.9280 - val_loss: 0.3196 - val_accuracy: 0.8846
Epoch 78/100
12/12 [=====] - 0s 26ms/step - loss: 0.1680 - accuracy: 0.9307 - val_loss: 0.4103 - val_accuracy: 0.8846
Epoch 79/100
12/12 [=====] - 0s 24ms/step - loss: 0.1767 - accuracy: 0.9335 - val_loss: 0.3969 - val_accuracy: 0.8846
Epoch 80/100
12/12 [=====] - 0s 23ms/step - loss: 0.2094 - accuracy: 0.8975 - val_loss: 0.3562 - val_accuracy: 0.8846
Epoch 81/100
12/12 [=====] - 0s 23ms/step - loss: 0.1695 - accuracy: 0.9280 - val_loss: 0.4328 - val_accuracy: 0.8462
Epoch 82/100
12/12 [=====] - 0s 26ms/step - loss: 0.1691 - accuracy: 0.9224 - val_loss: 0.3936 - val_accuracy: 0.8846
Epoch 83/100
12/12 [=====] - 0s 24ms/step - loss: 0.1365 - accuracy: 0.9557 - val_loss: 0.3536 - val_accuracy: 0.8974
Epoch 84/100
12/12 [=====] - 0s 24ms/step - loss: 0.1393 - accuracy: 0.9446 - val_loss: 0.4879 - val_accuracy: 0.7949
Epoch 85/100
12/12 [=====] - 0s 26ms/step - loss: 0.1962 - accuracy: 0.9030 - val_loss: 0.5662 - val_accuracy: 0.8333
Epoch 86/100
12/12 [=====] - 0s 24ms/step - loss: 0.1860 - accuracy: 0.9252 - val_loss: 0.4006 - val_accuracy: 0.8718
Epoch 87/100
12/12 [=====] - 0s 26ms/step - loss: 0.1778 - accuracy: 0.9224 - val_loss: 0.3889 - val_accuracy: 0.8205
Epoch 88/100
12/12 [=====] - 0s 25ms/step - loss: 0.1406 - accuracy: 0.9474 - val_loss: 0.3560 - val_accuracy: 0.8718
Epoch 89/100
12/12 [=====] - 0s 24ms/step - loss: 0.1194 - accuracy: 0.9474 - val_loss: 0.4761 - val_accuracy: 0.8205
Epoch 90/100
12/12 [=====] - 0s 27ms/step - loss: 0.1285 - accuracy: 0.9474 - val_loss: 0.4399 - val_accuracy: 0.8333
Epoch 91/100
12/12 [=====] - 0s 23ms/step - loss: 0.1313 - accuracy: 0.9474 - val_loss: 0.4035 - val_accuracy: 0.8205
Epoch 92/100

```

12/12 [=====] - 0s 26ms/step - loss: 0.1706 - accu
racy: 0.9252 - val_loss: 0.5791 - val_accuracy: 0.8846
Epoch 93/100
12/12 [=====] - 0s 23ms/step - loss: 0.2398 - accu
racy: 0.8975 - val_loss: 0.4048 - val_accuracy: 0.9103
Epoch 94/100
12/12 [=====] - 0s 26ms/step - loss: 0.2582 - accu
racy: 0.8864 - val_loss: 0.3005 - val_accuracy: 0.8846
Epoch 95/100
12/12 [=====] - 0s 25ms/step - loss: 0.2174 - accu
racy: 0.9030 - val_loss: 0.2741 - val_accuracy: 0.8846
Epoch 96/100
12/12 [=====] - 0s 23ms/step - loss: 0.1804 - accu
racy: 0.9335 - val_loss: 0.3465 - val_accuracy: 0.8974
Epoch 97/100
12/12 [=====] - 0s 26ms/step - loss: 0.1626 - accu
racy: 0.9307 - val_loss: 0.2817 - val_accuracy: 0.8846
Epoch 98/100
12/12 [=====] - 0s 24ms/step - loss: 0.1542 - accu
racy: 0.9280 - val_loss: 0.3053 - val_accuracy: 0.8590
Epoch 99/100
12/12 [=====] - 0s 26ms/step - loss: 0.1467 - accu
racy: 0.9391 - val_loss: 0.3768 - val_accuracy: 0.8590
Epoch 100/100
12/12 [=====] - 0s 23ms/step - loss: 0.1407 - accu

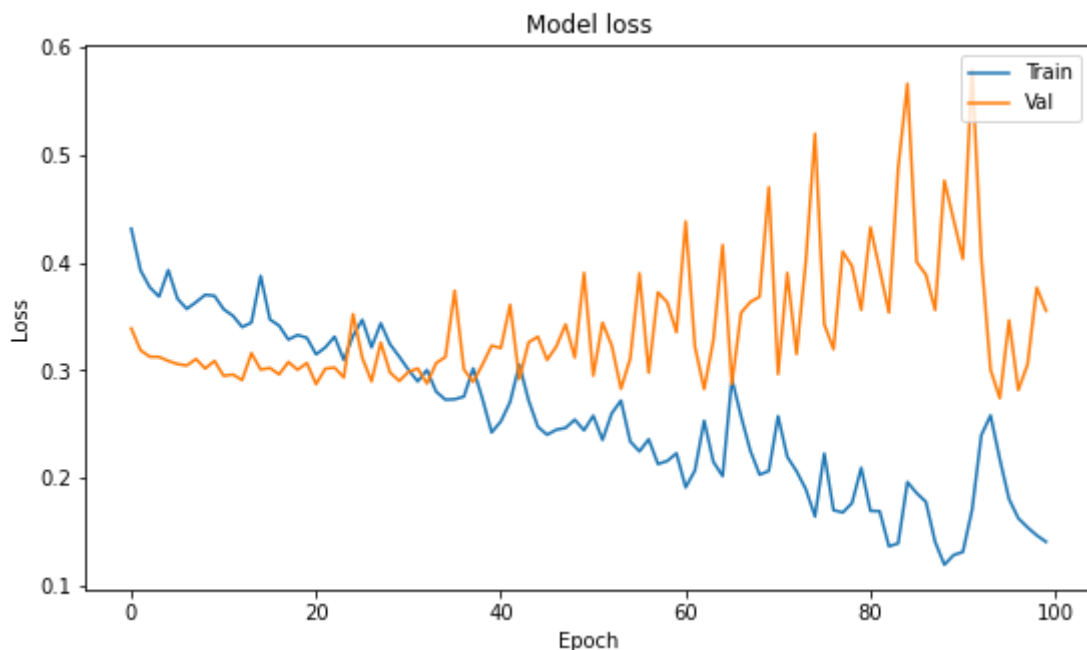
```

In [71]:

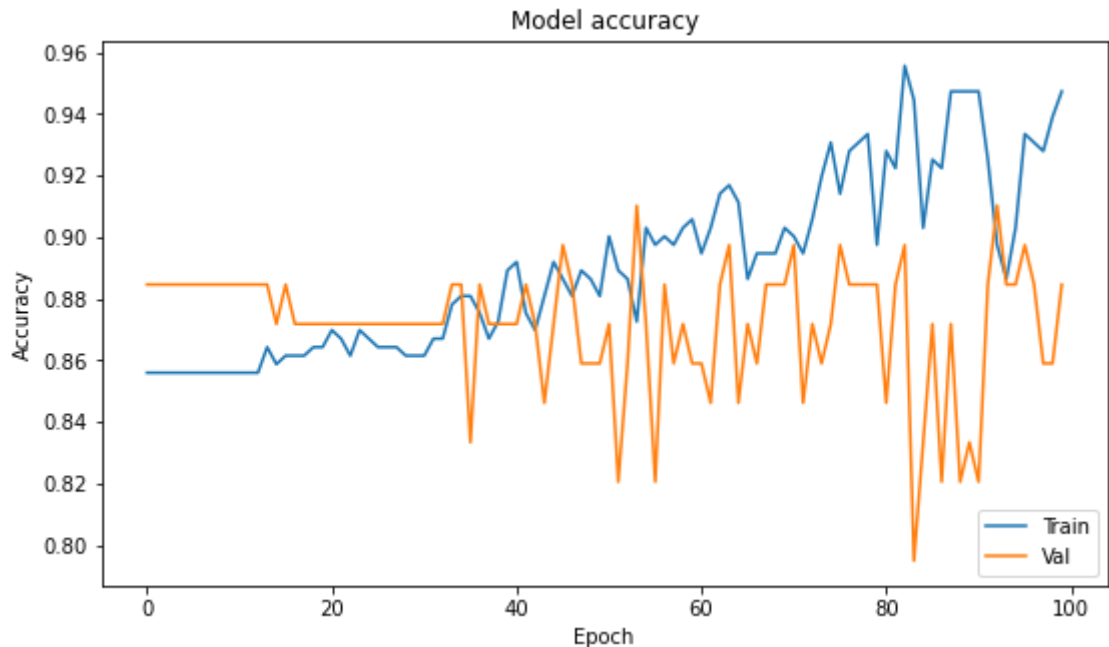
```

plt.plot(hist_2.history['loss'])
plt.plot(hist_2.history['val_loss'])
plt.title('Model loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['Train', 'Val'], loc='upper right')
plt.show()

```



```
In [73]: plt.plot(hist_2.history['accuracy'])
plt.plot(hist_2.history['val_accuracy'])
plt.title('Model accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(['Train', 'Val'], loc='lower right')
plt.show()
```



```
In [75]: model_3 = Sequential([
    Dense(1000, activation='relu', kernel_regularizer=regularizers.l2(0.01),
    Dropout(0.3),
    Dense(1000, activation='relu', kernel_regularizer=regularizers.l2(0.01),
    Dropout(0.3),
    Dense(1000, activation='relu', kernel_regularizer=regularizers.l2(0.01),
    Dropout(0.3),
    Dense(1000, activation='relu', kernel_regularizer=regularizers.l2(0.01),
    Dropout(0.3),
    Dense(1, activation='sigmoid', kernel_regularizer=regularizers.l2(0.01),
])
```

```
In [76]: model_3.compile(optimizer='adam',
    loss='binary_crossentropy',
    metrics=['accuracy'])

hist_3 = model_3.fit(X_train, Y_train,
    batch_size=32, epochs=100,
    validation_data=(X_val, Y_val))
```

```
Epoch 1/100
12/12 [=====] - 1s 48ms/step - loss: 23.5925 - acc
uracy: 0.8199 - val_loss: 15.5078 - val_accuracy: 0.8846
Epoch 2/100
12/12 [=====] - 0s 39ms/step - loss: 11.4112 - acc
uracy: 0.8560 - val_loss: 6.9854 - val_accuracy: 0.8846
Epoch 3/100
12/12 [=====] - 0s 38ms/step - loss: 5.0096 - accu
racy: 0.8560 - val_loss: 2.9277 - val_accuracy: 0.8846
Epoch 4/100
```

12/12 [=====] - 0s 39ms/step - loss: 2.1268 - accuracy: 0.8560 - val_loss: 1.2712 - val_accuracy: 0.8846
Epoch 5/100
12/12 [=====] - 0s 38ms/step - loss: 1.0031 - accuracy: 0.8560 - val_loss: 0.6730 - val_accuracy: 0.8846
Epoch 6/100
12/12 [=====] - 0s 41ms/step - loss: 0.6291 - accuracy: 0.8560 - val_loss: 0.5209 - val_accuracy: 0.8846
Epoch 7/100
12/12 [=====] - 0s 39ms/step - loss: 0.5240 - accuracy: 0.8560 - val_loss: 0.4398 - val_accuracy: 0.8846
Epoch 8/100
12/12 [=====] - 0s 38ms/step - loss: 0.4897 - accuracy: 0.8560 - val_loss: 0.4287 - val_accuracy: 0.8846
Epoch 9/100
12/12 [=====] - 0s 39ms/step - loss: 0.4781 - accuracy: 0.8560 - val_loss: 0.4197 - val_accuracy: 0.8846
Epoch 10/100
12/12 [=====] - 0s 40ms/step - loss: 0.4616 - accuracy: 0.8560 - val_loss: 0.4097 - val_accuracy: 0.8846
Epoch 11/100
12/12 [=====] - 0s 39ms/step - loss: 0.4640 - accuracy: 0.8560 - val_loss: 0.4210 - val_accuracy: 0.8846
Epoch 12/100
12/12 [=====] - 0s 39ms/step - loss: 0.4590 - accuracy: 0.8560 - val_loss: 0.4019 - val_accuracy: 0.8846
Epoch 13/100
12/12 [=====] - 1s 43ms/step - loss: 0.4580 - accuracy: 0.8560 - val_loss: 0.4007 - val_accuracy: 0.8846
Epoch 14/100
12/12 [=====] - 0s 38ms/step - loss: 0.4637 - accuracy: 0.8560 - val_loss: 0.4153 - val_accuracy: 0.8846
Epoch 15/100
12/12 [=====] - 0s 40ms/step - loss: 0.4551 - accuracy: 0.8560 - val_loss: 0.3982 - val_accuracy: 0.8846
Epoch 16/100
12/12 [=====] - 0s 38ms/step - loss: 0.4475 - accuracy: 0.8560 - val_loss: 0.3984 - val_accuracy: 0.8846
Epoch 17/100
12/12 [=====] - 0s 38ms/step - loss: 0.4468 - accuracy: 0.8560 - val_loss: 0.3992 - val_accuracy: 0.8846
Epoch 18/100
12/12 [=====] - 0s 39ms/step - loss: 0.4452 - accuracy: 0.8560 - val_loss: 0.3955 - val_accuracy: 0.8846
Epoch 19/100
12/12 [=====] - 0s 38ms/step - loss: 0.4502 - accuracy: 0.8560 - val_loss: 0.3955 - val_accuracy: 0.8846
Epoch 20/100
12/12 [=====] - 0s 39ms/step - loss: 0.4411 - accuracy: 0.8560 - val_loss: 0.3965 - val_accuracy: 0.8846
Epoch 21/100
12/12 [=====] - 0s 38ms/step - loss: 0.4441 - accuracy: 0.8560 - val_loss: 0.3940 - val_accuracy: 0.8846
Epoch 22/100
12/12 [=====] - 0s 39ms/step - loss: 0.4451 - accuracy: 0.8560 - val_loss: 0.3960 - val_accuracy: 0.8846
Epoch 23/100
12/12 [=====] - 0s 37ms/step - loss: 0.4449 - accuracy: 0.8560 - val_loss: 0.3958 - val_accuracy: 0.8846
Epoch 24/100
12/12 [=====] - 0s 38ms/step - loss: 0.4459 - accuracy: 0.8560 - val_loss: 0.3955 - val_accuracy: 0.8846
Epoch 25/100

12/12 [=====] - 0s 37ms/step - loss: 0.4423 - accuracy: 0.8560 - val_loss: 0.3924 - val_accuracy: 0.8846
Epoch 26/100
12/12 [=====] - 0s 39ms/step - loss: 0.4466 - accuracy: 0.8560 - val_loss: 0.3949 - val_accuracy: 0.8846
Epoch 27/100
12/12 [=====] - 0s 38ms/step - loss: 0.4485 - accuracy: 0.8560 - val_loss: 0.3966 - val_accuracy: 0.8846
Epoch 28/100
12/12 [=====] - 0s 38ms/step - loss: 0.4438 - accuracy: 0.8560 - val_loss: 0.3956 - val_accuracy: 0.8846
Epoch 29/100
12/12 [=====] - 0s 40ms/step - loss: 0.4380 - accuracy: 0.8560 - val_loss: 0.3909 - val_accuracy: 0.8846
Epoch 30/100
12/12 [=====] - 0s 40ms/step - loss: 0.4434 - accuracy: 0.8560 - val_loss: 0.3927 - val_accuracy: 0.8846
Epoch 31/100
12/12 [=====] - 0s 42ms/step - loss: 0.4432 - accuracy: 0.8560 - val_loss: 0.3961 - val_accuracy: 0.8846
Epoch 32/100
12/12 [=====] - 1s 46ms/step - loss: 0.4400 - accuracy: 0.8560 - val_loss: 0.3895 - val_accuracy: 0.8846
Epoch 33/100
12/12 [=====] - 0s 41ms/step - loss: 0.4452 - accuracy: 0.8560 - val_loss: 0.3927 - val_accuracy: 0.8846
Epoch 34/100
12/12 [=====] - 1s 48ms/step - loss: 0.4436 - accuracy: 0.8560 - val_loss: 0.3900 - val_accuracy: 0.8846
Epoch 35/100
12/12 [=====] - 1s 43ms/step - loss: 0.4395 - accuracy: 0.8560 - val_loss: 0.3887 - val_accuracy: 0.8846
Epoch 36/100
12/12 [=====] - 0s 39ms/step - loss: 0.4417 - accuracy: 0.8560 - val_loss: 0.3938 - val_accuracy: 0.8846
Epoch 37/100
12/12 [=====] - 1s 44ms/step - loss: 0.4443 - accuracy: 0.8560 - val_loss: 0.3885 - val_accuracy: 0.8846
Epoch 38/100
12/12 [=====] - 1s 43ms/step - loss: 0.4390 - accuracy: 0.8560 - val_loss: 0.3936 - val_accuracy: 0.8846
Epoch 39/100
12/12 [=====] - 1s 44ms/step - loss: 0.4441 - accuracy: 0.8560 - val_loss: 0.3882 - val_accuracy: 0.8846
Epoch 40/100
12/12 [=====] - 0s 38ms/step - loss: 0.4420 - accuracy: 0.8560 - val_loss: 0.3907 - val_accuracy: 0.8846
Epoch 41/100
12/12 [=====] - 0s 38ms/step - loss: 0.4432 - accuracy: 0.8560 - val_loss: 0.3867 - val_accuracy: 0.8846
Epoch 42/100
12/12 [=====] - 1s 43ms/step - loss: 0.4425 - accuracy: 0.8560 - val_loss: 0.3934 - val_accuracy: 0.8846
Epoch 43/100
12/12 [=====] - 0s 40ms/step - loss: 0.4392 - accuracy: 0.8560 - val_loss: 0.3909 - val_accuracy: 0.8846
Epoch 44/100
12/12 [=====] - 0s 39ms/step - loss: 0.4355 - accuracy: 0.8560 - val_loss: 0.3837 - val_accuracy: 0.8846
Epoch 45/100
12/12 [=====] - 1s 43ms/step - loss: 0.4390 - accuracy: 0.8560 - val_loss: 0.3865 - val_accuracy: 0.8846
Epoch 46/100

12/12 [=====] - 0s 39ms/step - loss: 0.4360 - accuracy: 0.8560 - val_loss: 0.3885 - val_accuracy: 0.8846
Epoch 47/100
12/12 [=====] - 0s 39ms/step - loss: 0.4349 - accuracy: 0.8560 - val_loss: 0.3823 - val_accuracy: 0.8846
Epoch 48/100
12/12 [=====] - 0s 40ms/step - loss: 0.4330 - accuracy: 0.8560 - val_loss: 0.3820 - val_accuracy: 0.8846
Epoch 49/100
12/12 [=====] - 0s 39ms/step - loss: 0.4326 - accuracy: 0.8560 - val_loss: 0.3815 - val_accuracy: 0.8846
Epoch 50/100
12/12 [=====] - 0s 41ms/step - loss: 0.4309 - accuracy: 0.8560 - val_loss: 0.3798 - val_accuracy: 0.8846
Epoch 51/100
12/12 [=====] - 0s 40ms/step - loss: 0.4320 - accuracy: 0.8560 - val_loss: 0.3811 - val_accuracy: 0.8846
Epoch 52/100
12/12 [=====] - 0s 39ms/step - loss: 0.4290 - accuracy: 0.8560 - val_loss: 0.3773 - val_accuracy: 0.8846
Epoch 53/100
12/12 [=====] - 0s 40ms/step - loss: 0.4279 - accuracy: 0.8560 - val_loss: 0.3775 - val_accuracy: 0.8846
Epoch 54/100
12/12 [=====] - 0s 40ms/step - loss: 0.4250 - accuracy: 0.8560 - val_loss: 0.3753 - val_accuracy: 0.8846
Epoch 55/100
12/12 [=====] - 0s 42ms/step - loss: 0.4265 - accuracy: 0.8560 - val_loss: 0.3744 - val_accuracy: 0.8846
Epoch 56/100
12/12 [=====] - 0s 39ms/step - loss: 0.4241 - accuracy: 0.8560 - val_loss: 0.3756 - val_accuracy: 0.8846
Epoch 57/100
12/12 [=====] - 0s 40ms/step - loss: 0.4260 - accuracy: 0.8560 - val_loss: 0.3716 - val_accuracy: 0.8846
Epoch 58/100
12/12 [=====] - 0s 42ms/step - loss: 0.4242 - accuracy: 0.8560 - val_loss: 0.3741 - val_accuracy: 0.8846
Epoch 59/100
12/12 [=====] - 0s 39ms/step - loss: 0.4248 - accuracy: 0.8560 - val_loss: 0.3721 - val_accuracy: 0.8846
Epoch 60/100
12/12 [=====] - 0s 40ms/step - loss: 0.4222 - accuracy: 0.8560 - val_loss: 0.3766 - val_accuracy: 0.8846
Epoch 61/100
12/12 [=====] - 1s 45ms/step - loss: 0.4245 - accuracy: 0.8560 - val_loss: 0.3733 - val_accuracy: 0.8846
Epoch 62/100
12/12 [=====] - 0s 42ms/step - loss: 0.4235 - accuracy: 0.8560 - val_loss: 0.3761 - val_accuracy: 0.8846
Epoch 63/100
12/12 [=====] - 1s 43ms/step - loss: 0.4249 - accuracy: 0.8560 - val_loss: 0.3729 - val_accuracy: 0.8846
Epoch 64/100
12/12 [=====] - 0s 39ms/step - loss: 0.4227 - accuracy: 0.8560 - val_loss: 0.3694 - val_accuracy: 0.8846
Epoch 65/100
12/12 [=====] - 0s 39ms/step - loss: 0.4208 - accuracy: 0.8560 - val_loss: 0.3735 - val_accuracy: 0.8846
Epoch 66/100
12/12 [=====] - 0s 41ms/step - loss: 0.4246 - accuracy: 0.8560 - val_loss: 0.3758 - val_accuracy: 0.8846
Epoch 67/100

12/12 [=====] - 0s 38ms/step - loss: 0.4236 - accuracy: 0.8560 - val_loss: 0.3706 - val_accuracy: 0.8846
Epoch 68/100
12/12 [=====] - 0s 40ms/step - loss: 0.4215 - accuracy: 0.8560 - val_loss: 0.3775 - val_accuracy: 0.8846
Epoch 69/100
12/12 [=====] - 0s 40ms/step - loss: 0.4204 - accuracy: 0.8560 - val_loss: 0.3715 - val_accuracy: 0.8846
Epoch 70/100
12/12 [=====] - 0s 38ms/step - loss: 0.4208 - accuracy: 0.8560 - val_loss: 0.3685 - val_accuracy: 0.8846
Epoch 71/100
12/12 [=====] - 0s 41ms/step - loss: 0.4236 - accuracy: 0.8560 - val_loss: 0.3696 - val_accuracy: 0.8846
Epoch 72/100
12/12 [=====] - 0s 41ms/step - loss: 0.4211 - accuracy: 0.8560 - val_loss: 0.3731 - val_accuracy: 0.8846
Epoch 73/100
12/12 [=====] - 0s 39ms/step - loss: 0.4211 - accuracy: 0.8560 - val_loss: 0.3700 - val_accuracy: 0.8846
Epoch 74/100
12/12 [=====] - 0s 41ms/step - loss: 0.4203 - accuracy: 0.8560 - val_loss: 0.3740 - val_accuracy: 0.8846
Epoch 75/100
12/12 [=====] - 0s 39ms/step - loss: 0.4201 - accuracy: 0.8560 - val_loss: 0.3702 - val_accuracy: 0.8846
Epoch 76/100
12/12 [=====] - 0s 39ms/step - loss: 0.4209 - accuracy: 0.8560 - val_loss: 0.3704 - val_accuracy: 0.8846
Epoch 77/100
12/12 [=====] - 1s 42ms/step - loss: 0.4224 - accuracy: 0.8560 - val_loss: 0.3746 - val_accuracy: 0.8846
Epoch 78/100
12/12 [=====] - 0s 40ms/step - loss: 0.4208 - accuracy: 0.8560 - val_loss: 0.3709 - val_accuracy: 0.8846
Epoch 79/100
12/12 [=====] - 0s 41ms/step - loss: 0.4203 - accuracy: 0.8560 - val_loss: 0.3682 - val_accuracy: 0.8846
Epoch 80/100
12/12 [=====] - 0s 39ms/step - loss: 0.4196 - accuracy: 0.8560 - val_loss: 0.3703 - val_accuracy: 0.8846
Epoch 81/100
12/12 [=====] - 0s 39ms/step - loss: 0.4210 - accuracy: 0.8560 - val_loss: 0.3684 - val_accuracy: 0.8846
Epoch 82/100
12/12 [=====] - 0s 40ms/step - loss: 0.4216 - accuracy: 0.8560 - val_loss: 0.3704 - val_accuracy: 0.8846
Epoch 83/100
12/12 [=====] - 0s 40ms/step - loss: 0.4209 - accuracy: 0.8560 - val_loss: 0.3690 - val_accuracy: 0.8846
Epoch 84/100
12/12 [=====] - 0s 38ms/step - loss: 0.4202 - accuracy: 0.8560 - val_loss: 0.3703 - val_accuracy: 0.8846
Epoch 85/100
12/12 [=====] - 0s 40ms/step - loss: 0.4209 - accuracy: 0.8560 - val_loss: 0.3730 - val_accuracy: 0.8846
Epoch 86/100
12/12 [=====] - 0s 40ms/step - loss: 0.4208 - accuracy: 0.8560 - val_loss: 0.3690 - val_accuracy: 0.8846
Epoch 87/100
12/12 [=====] - 1s 43ms/step - loss: 0.4189 - accuracy: 0.8560 - val_loss: 0.3672 - val_accuracy: 0.8846
Epoch 88/100


```

12/12 [=====] - 1s 43ms/step - loss: 0.4208 - accuracy: 0.8560 - val_loss: 0.3682 - val_accuracy: 0.8846
Epoch 89/100
12/12 [=====] - 0s 41ms/step - loss: 0.4184 - accuracy: 0.8560 - val_loss: 0.3680 - val_accuracy: 0.8846
Epoch 90/100
12/12 [=====] - 0s 42ms/step - loss: 0.4185 - accuracy: 0.8560 - val_loss: 0.3699 - val_accuracy: 0.8846
Epoch 91/100
12/12 [=====] - 1s 42ms/step - loss: 0.4205 - accuracy: 0.8560 - val_loss: 0.3671 - val_accuracy: 0.8846
Epoch 92/100
12/12 [=====] - 1s 43ms/step - loss: 0.4197 - accuracy: 0.8560 - val_loss: 0.3718 - val_accuracy: 0.8846
Epoch 93/100
12/12 [=====] - 0s 41ms/step - loss: 0.4190 - accuracy: 0.8560 - val_loss: 0.3688 - val_accuracy: 0.8846
Epoch 94/100
12/12 [=====] - 0s 40ms/step - loss: 0.4214 - accuracy: 0.8560 - val_loss: 0.3702 - val_accuracy: 0.8846
Epoch 95/100
12/12 [=====] - 0s 41ms/step - loss: 0.4211 - accuracy: 0.8560 - val_loss: 0.3672 - val_accuracy: 0.8846
Epoch 96/100
12/12 [=====] - 0s 40ms/step - loss: 0.4205 - accuracy: 0.8560 - val_loss: 0.3697 - val_accuracy: 0.8846
Epoch 97/100
12/12 [=====] - 0s 41ms/step - loss: 0.4178 - accuracy: 0.8560 - val_loss: 0.3668 - val_accuracy: 0.8846
Epoch 98/100
12/12 [=====] - 0s 39ms/step - loss: 0.4187 - accuracy: 0.8560 - val_loss: 0.3659 - val_accuracy: 0.8846
Epoch 99/100
12/12 [=====] - 0s 41ms/step - loss: 0.4175 - accuracy: 0.8560 - val_loss: 0.3665 - val_accuracy: 0.8846
Epoch 100/100

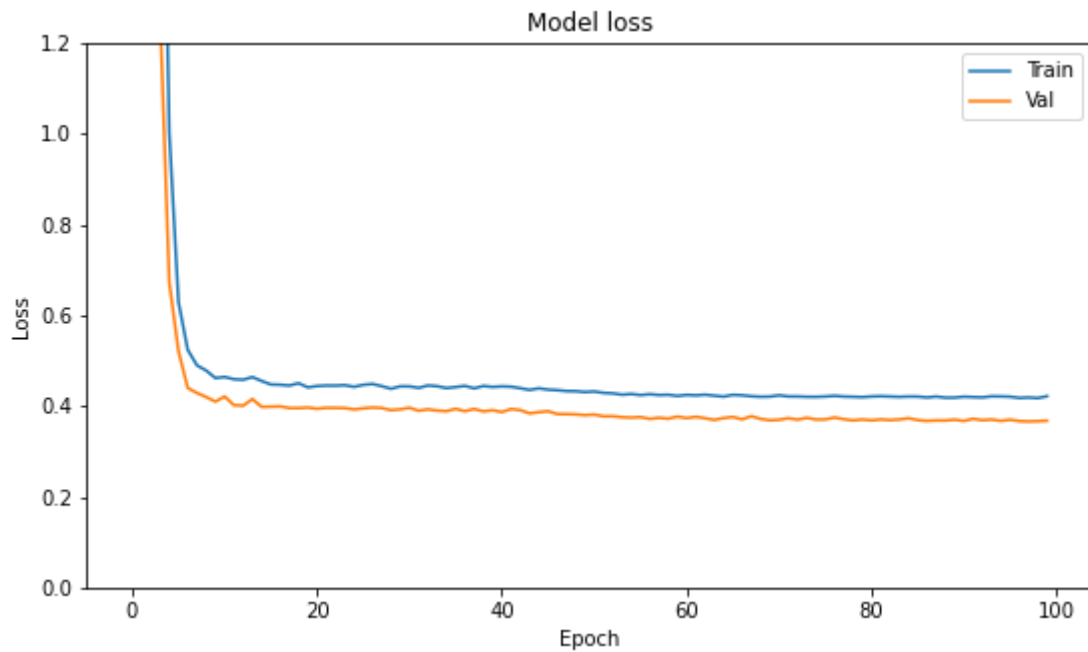
```

In [77]:

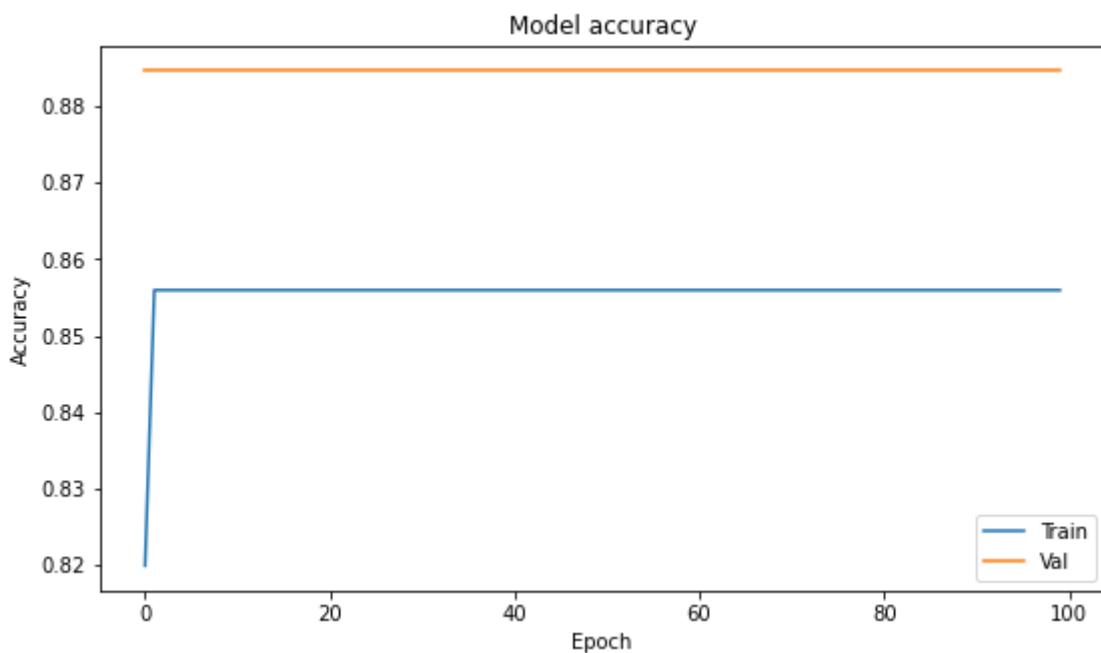
```

plt.plot(hist_3.history['loss'])
plt.plot(hist_3.history['val_loss'])
plt.title('Model loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['Train', 'Val'], loc='upper right')
plt.ylim(top=1.2, bottom=0)
plt.show()

```



```
In [78]: plt.plot(hist_3.history['accuracy'])
plt.plot(hist_3.history['val_accuracy'])
plt.title('Model accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(['Train', 'Val'], loc='lower right')
plt.show()
```



conclusion

```
In [83]: predictions = model.predict(X_train)
```

```
In [82]: for i in range(5):
          print('%s => %d (expected %d)' % (X[i].tolist(), predictions[i], y[i]))
```

[86.2, 26.2, 94.3, 5.1, 8.2, 51.0, 6.7, 0.0, 0.0, 1.0] => 0 (expected 1)

```
[90.6, 35.4, 669.1, 6.7, 18.0, 33.0, 0.9, 0.0, 0.0, 0.0] => 0 (expected 1)
[90.6, 43.7, 686.9, 6.7, 14.6, 33.0, 1.3, 0.0, 0.0, 0.0] => 0 (expected 1)
[91.7, 33.3, 77.5, 9.0, 8.3, 97.0, 4.0, 0.2, 0.0, 1.0] => 0 (expected 1)
[88.3, 51.3, 102.2, 9.6, 11.4, 99.0, 1.8, 0.0, 0.0, 0.0] => 0 (expected 1)
```

In []: