

```
#
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy_score, f1_score
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense

data = pd.read_csv('cancer data (1).csv') # Importing the dataset
```

data

	id	diagnosis	radius_mean	texture_mean	perimeter_mean
area_mean \					
0	842302	M	17.99	10.38	122.80
1001.0					
1	842517	M	20.57	17.77	132.90
1326.0					
2	84300903	M	19.69	21.25	130.00
1203.0					
3	84348301	M	11.42	20.38	77.58
386.1					
4	84358402	M	20.29	14.34	135.10
1297.0					
..	...	...	...	...	...
...					
564	926424	M	21.56	22.39	142.00
1479.0					
565	926682	M	20.13	28.25	131.20
1261.0					
566	926954	M	16.60	28.08	108.30
858.1					
567	927241	M	20.60	29.33	140.10
1265.0					
568	92751	B	7.76	24.54	47.92
181.0					

	smoothness_mean	compactness_mean	concavity_mean	concave
points_mean \				
0	0.11840	0.27760	0.30010	
0.14710				
1	0.08474	0.07864	0.08690	
0.07017				
2	0.10960	0.15990	0.19740	
0.12790				
3	0.14250	0.28390	0.24140	
0.10520				
4	0.10030	0.13280	0.19800	
0.10430				

...	...	...	...
564	0.11100	0.11590	0.24390
0.13890			
565	0.09780	0.10340	0.14400
0.09791			
566	0.08455	0.10230	0.09251
0.05302			
567	0.11780	0.27700	0.35140
0.15200			
568	0.05263	0.04362	0.00000
0.00000			

	...	texture_worst	perimeter_worst	area_worst	smoothness_worst
\					
0	...	17.33	184.60	2019.0	0.16220
1	...	23.41	158.80	1956.0	0.12380
2	...	25.53	152.50	1709.0	0.14440
3	...	26.50	98.87	567.7	0.20980
4	...	16.67	152.20	1575.0	0.13740
...	...	...	...	...	...
564	...	26.40	166.10	2027.0	0.14100
565	...	38.25	155.00	1731.0	0.11660
566	...	34.12	126.70	1124.0	0.11390
567	...	39.42	184.60	1821.0	0.16500
568	...	30.37	59.16	268.6	0.08996

	compactness_worst	concavity_worst	concave points_worst
symmetry_worst \			
0	0.66560	0.7119	0.2654
0.4601			
1	0.18660	0.2416	0.1860
0.2750			
2	0.42450	0.4504	0.2430
0.3613			
3	0.86630	0.6869	0.2575
0.6638			
4	0.20500	0.4000	0.1625

```

0.2364
...
...
564      0.21130      0.4107      0.2216
0.2060
565      0.19220      0.3215      0.1628
0.2572
566      0.30940      0.3403      0.1418
0.2218
567      0.86810      0.9387      0.2650
0.4087
568      0.06444      0.0000      0.0000
0.2871

```

```

      fractal_dimension_worst  Unnamed: 32
0      0.11890      NaN
1      0.08902      NaN
2      0.08758      NaN
3      0.17300      NaN
4      0.07678      NaN
...
564      0.07115      NaN
565      0.06637      NaN
566      0.07820      NaN
567      0.12400      NaN
568      0.07039      NaN

```

[569 rows x 33 columns]

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 569 entries, 0 to 568
```

```
Data columns (total 33 columns):
```

#	Column	Non-Null Count	Dtype
0	id	569 non-null	int64
1	diagnosis	569 non-null	object
2	radius_mean	569 non-null	float64
3	texture_mean	569 non-null	float64
4	perimeter_mean	569 non-null	float64
5	area_mean	569 non-null	float64
6	smoothness_mean	569 non-null	float64
7	compactness_mean	569 non-null	float64
8	concavity_mean	569 non-null	float64
9	concave points_mean	569 non-null	float64
10	symmetry_mean	569 non-null	float64
11	fractal_dimension_mean	569 non-null	float64
12	radius_se	569 non-null	float64
13	texture_se	569 non-null	float64

14	perimeter_se	569	non-null	float64
15	area_se	569	non-null	float64
16	smoothness_se	569	non-null	float64
17	compactness_se	569	non-null	float64
18	concavity_se	569	non-null	float64
19	concave points_se	569	non-null	float64
20	symmetry_se	569	non-null	float64
21	fractal_dimension_se	569	non-null	float64
22	radius_worst	569	non-null	float64
23	texture_worst	569	non-null	float64
24	perimeter_worst	569	non-null	float64
25	area_worst	569	non-null	float64
26	smoothness_worst	569	non-null	float64
27	compactness_worst	569	non-null	float64
28	concavity_worst	569	non-null	float64
29	concave points_worst	569	non-null	float64
30	symmetry_worst	569	non-null	float64
31	fractal_dimension_worst	569	non-null	float64
32	Unnamed: 32	0	non-null	float64

dtypes: float64(31), int64(1), object(1)  
memory usage: 146.8+ KB

data.isnull().sum()

id	0
diagnosis	0
radius_mean	0
texture_mean	0
perimeter_mean	0
area_mean	0
smoothness_mean	0
compactness_mean	0
concavity_mean	0
concave points_mean	0
symmetry_mean	0
fractal_dimension_mean	0
radius_se	0
texture_se	0
perimeter_se	0
area_se	0
smoothness_se	0
compactness_se	0
concavity_se	0
concave points_se	0
symmetry_se	0
fractal_dimension_se	0
radius_worst	0
texture_worst	0
perimeter_worst	0
area_worst	0
smoothness_worst	0

```
compactness_worst      0
concavity_worst        0
concave points_worst   0
symmetry_worst         0
fractal_dimension_worst 0
Unnamed: 32            569
dtype: int64
```

```
data = data.drop(["id","Unnamed: 32"], axis="columns") # removing
the unwanted columns
```

```
data
```

	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	\
0	M	17.99	10.38	122.80	1001.0	
1	M	20.57	17.77	132.90	1326.0	
2	M	19.69	21.25	130.00	1203.0	
3	M	11.42	20.38	77.58	386.1	
4	M	20.29	14.34	135.10	1297.0	
...	...	...	...	...	...	
564	M	21.56	22.39	142.00	1479.0	
565	M	20.13	28.25	131.20	1261.0	
566	M	16.60	28.08	108.30	858.1	
567	M	20.60	29.33	140.10	1265.0	
568	B	7.76	24.54	47.92	181.0	

	smoothness_mean	compactness_mean	concavity_mean	concave
points_mean \				
0	0.11840	0.27760	0.30010	
0.14710				
1	0.08474	0.07864	0.08690	
0.07017				
2	0.10960	0.15990	0.19740	
0.12790				
3	0.14250	0.28390	0.24140	
0.10520				
4	0.10030	0.13280	0.19800	
0.10430				
...	...	...	...	
...				
564	0.11100	0.11590	0.24390	
0.13890				
565	0.09780	0.10340	0.14400	
0.09791				
566	0.08455	0.10230	0.09251	
0.05302				
567	0.11780	0.27700	0.35140	
0.15200				
568	0.05263	0.04362	0.00000	
0.00000				

	symmetry_mean	...	radius_worst	texture_worst	perimeter_worst
\ 0	0.2419	...	25.380	17.33	184.60
1	0.1812	...	24.990	23.41	158.80
2	0.2069	...	23.570	25.53	152.50
3	0.2597	...	14.910	26.50	98.87
4	0.1809	...	22.540	16.67	152.20
..	...	...	...	...	...
564	0.1726	...	25.450	26.40	166.10
565	0.1752	...	23.690	38.25	155.00
566	0.1590	...	18.980	34.12	126.70
567	0.2397	...	25.740	39.42	184.60
568	0.1587	...	9.456	30.37	59.16

	area_worst	smoothness_worst	compactness_worst	concavity_worst
\ 0	2019.0	0.16220	0.66560	0.7119
1	1956.0	0.12380	0.18660	0.2416
2	1709.0	0.14440	0.42450	0.4504
3	567.7	0.20980	0.86630	0.6869
4	1575.0	0.13740	0.20500	0.4000
..	...	...	...	...
564	2027.0	0.14100	0.21130	0.4107
565	1731.0	0.11660	0.19220	0.3215
566	1124.0	0.11390	0.30940	0.3403
567	1821.0	0.16500	0.86810	0.9387
568	268.6	0.08996	0.06444	0.0000

	concave points_worst	symmetry_worst	fractal_dimension_worst
0	0.2654	0.4601	0.11890
1	0.1860	0.2750	0.08902
2	0.2430	0.3613	0.08758
3	0.2575	0.6638	0.17300
4	0.1625	0.2364	0.07678
..	...	...	...
564	0.2216	0.2060	0.07115
565	0.1628	0.2572	0.06637
566	0.1418	0.2218	0.07820
567	0.2650	0.4087	0.12400
568	0.0000	0.2871	0.07039

[569 rows x 31 columns]

data.info()

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 569 entries, 0 to 568

Data columns (total 31 columns):

#	Column	Non-Null Count	Dtype
0	diagnosis	569 non-null	object
1	radius_mean	569 non-null	float64
2	texture_mean	569 non-null	float64
3	perimeter_mean	569 non-null	float64
4	area_mean	569 non-null	float64
5	smoothness_mean	569 non-null	float64
6	compactness_mean	569 non-null	float64
7	concavity_mean	569 non-null	float64
8	concave points_mean	569 non-null	float64
9	symmetry_mean	569 non-null	float64
10	fractal_dimension_mean	569 non-null	float64
11	radius_se	569 non-null	float64
12	texture_se	569 non-null	float64
13	perimeter_se	569 non-null	float64
14	area_se	569 non-null	float64
15	smoothness_se	569 non-null	float64
16	compactness_se	569 non-null	float64
17	concavity_se	569 non-null	float64
18	concave points_se	569 non-null	float64
19	symmetry_se	569 non-null	float64
20	fractal_dimension_se	569 non-null	float64
21	radius_worst	569 non-null	float64
22	texture_worst	569 non-null	float64
23	perimeter_worst	569 non-null	float64
24	area_worst	569 non-null	float64
25	smoothness_worst	569 non-null	float64

```

26 compactness_worst      569 non-null    float64
27 concavity_worst        569 non-null    float64
28 concave points_worst    569 non-null    float64
29 symmetry_worst         569 non-null    float64
30 fractal_dimension_worst 569 non-null    float64

```

```
dtypes: float64(30), object(1)
```

```
memory usage: 137.9+ KB
```

```
data["diagnosis"].unique()
```

```
array(['M', 'B'], dtype=object)
```

```

from sklearn import preprocessing #applying lable-encoder to
convert categorical variable into numerical.

```

```
Encode = preprocessing.LabelEncoder()
```

```
Encode.fit(['M','B'])
```

```
data["diagnosis"] = Encode.transform(data['diagnosis'])
```

```
data["diagnosis"].unique()
```

```
array([1, 0])
```

```
data
```

	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean \
0	1	17.99	10.38	122.80	1001.0
1	1	20.57	17.77	132.90	1326.0
2	1	19.69	21.25	130.00	1203.0
3	1	11.42	20.38	77.58	386.1
4	1	20.29	14.34	135.10	1297.0
..	...	...	...	...	...
564	1	21.56	22.39	142.00	1479.0
565	1	20.13	28.25	131.20	1261.0
566	1	16.60	28.08	108.30	858.1
567	1	20.60	29.33	140.10	1265.0
568	0	7.76	24.54	47.92	181.0

```
smoothness_mean compactness_mean concavity_mean concave
```



points_mean \			
0	0.11840	0.27760	0.30010
0.14710			
1	0.08474	0.07864	0.08690
0.07017			
2	0.10960	0.15990	0.19740
0.12790			
3	0.14250	0.28390	0.24140
0.10520			
4	0.10030	0.13280	0.19800
0.10430			
..	...	...	...
...			
564	0.11100	0.11590	0.24390
0.13890			
565	0.09780	0.10340	0.14400
0.09791			
566	0.08455	0.10230	0.09251
0.05302			
567	0.11780	0.27700	0.35140
0.15200			
568	0.05263	0.04362	0.00000
0.00000			

	symmetry_mean	...	radius_worst	texture_worst	perimeter_worst
\					
0	0.2419	...	25.380	17.33	184.60
1	0.1812	...	24.990	23.41	158.80
2	0.2069	...	23.570	25.53	152.50
3	0.2597	...	14.910	26.50	98.87
4	0.1809	...	22.540	16.67	152.20
..	...	...	...	...	...
564	0.1726	...	25.450	26.40	166.10
565	0.1752	...	23.690	38.25	155.00
566	0.1590	...	18.980	34.12	126.70
567	0.2397	...	25.740	39.42	184.60
568	0.1587	...	9.456	30.37	59.16

	area_worst	smoothness_worst	compactness_worst	concavity_worst
0	2019.0	0.16220	0.66560	0.7119
1	1956.0	0.12380	0.18660	0.2416
2	1709.0	0.14440	0.42450	0.4504
3	567.7	0.20980	0.86630	0.6869
4	1575.0	0.13740	0.20500	0.4000
..	...	...	...	...
564	2027.0	0.14100	0.21130	0.4107
565	1731.0	0.11660	0.19220	0.3215
566	1124.0	0.11390	0.30940	0.3403
567	1821.0	0.16500	0.86810	0.9387
568	268.6	0.08996	0.06444	0.0000

	concave points_worst	symmetry_worst	fractal_dimension_worst
0	0.2654	0.4601	0.11890
1	0.1860	0.2750	0.08902
2	0.2430	0.3613	0.08758
3	0.2575	0.6638	0.17300
4	0.1625	0.2364	0.07678
..	...	...	...
564	0.2216	0.2060	0.07115
565	0.1628	0.2572	0.06637
566	0.1418	0.2218	0.07820
567	0.2650	0.4087	0.12400
568	0.0000	0.2871	0.07039

[569 rows x 31 columns]

```
x = data.drop("diagnosis", axis= 1)
x
```

	radius_mean	texture_mean	perimeter_mean	area_mean
smoothness_mean \				
0	17.99	10.38	122.80	1001.0
0.11840				
1	20.57	17.77	132.90	1326.0
0.08474				

2	19.69	21.25	130.00	1203.0
0.10960				
3	11.42	20.38	77.58	386.1
0.14250				
4	20.29	14.34	135.10	1297.0
0.10030				
..	...	...	...	...
...				
564	21.56	22.39	142.00	1479.0
0.11100				
565	20.13	28.25	131.20	1261.0
0.09780				
566	16.60	28.08	108.30	858.1
0.08455				
567	20.60	29.33	140.10	1265.0
0.11780				
568	7.76	24.54	47.92	181.0
0.05263				

	compactness_mean	concavity_mean	concave points_mean
symmetry_mean \			
0	0.27760	0.30010	0.14710
0.2419			
1	0.07864	0.08690	0.07017
0.1812			
2	0.15990	0.19740	0.12790
0.2069			
3	0.28390	0.24140	0.10520
0.2597			
4	0.13280	0.19800	0.10430
0.1809			
..	...	...	...
...			
564	0.11590	0.24390	0.13890
0.1726			
565	0.10340	0.14400	0.09791
0.1752			
566	0.10230	0.09251	0.05302
0.1590			
567	0.27700	0.35140	0.15200
0.2397			
568	0.04362	0.00000	0.00000
0.1587			

	fractal_dimension_mean	...	radius_worst	texture_worst	\
0	0.07871	...	25.380	17.33	
1	0.05667	...	24.990	23.41	
2	0.05999	...	23.570	25.53	
3	0.09744	...	14.910	26.50	
4	0.05883	...	22.540	16.67	

..	...	...	...
564	0.05623	...	25.450
565	0.05533	...	23.690
566	0.05648	...	18.980
567	0.07016	...	25.740
568	0.05884	...	9.456

	perimeter_worst	area_worst	smoothness_worst	compactness_worst
\				
0	184.60	2019.0	0.16220	0.66560
1	158.80	1956.0	0.12380	0.18660
2	152.50	1709.0	0.14440	0.42450
3	98.87	567.7	0.20980	0.86630
4	152.20	1575.0	0.13740	0.20500
..	...	...	...	...
564	166.10	2027.0	0.14100	0.21130
565	155.00	1731.0	0.11660	0.19220
566	126.70	1124.0	0.11390	0.30940
567	184.60	1821.0	0.16500	0.86810
568	59.16	268.6	0.08996	0.06444

	concavity_worst	concave points_worst	symmetry_worst	\
0	0.7119	0.2654	0.4601	
1	0.2416	0.1860	0.2750	
2	0.4504	0.2430	0.3613	
3	0.6869	0.2575	0.6638	
4	0.4000	0.1625	0.2364	
..	...	...	...	
564	0.4107	0.2216	0.2060	
565	0.3215	0.1628	0.2572	
566	0.3403	0.1418	0.2218	
567	0.9387	0.2650	0.4087	
568	0.0000	0.0000	0.2871	

	fractal_dimension_worst
0	0.11890
1	0.08902
2	0.08758

```

3          0.17300
4          0.07678
..          ...
564        0.07115
565        0.06637
566        0.07820
567        0.12400
568        0.07039

```

[569 rows x 30 columns]

```
x = np.asanyarray(x)      # Independent variable
```

```
x
```

```

array([[1.799e+01, 1.038e+01, 1.228e+02, ..., 2.654e-01, 4.601e-01,
        1.189e-01],
       [2.057e+01, 1.777e+01, 1.329e+02, ..., 1.860e-01, 2.750e-01,
        8.902e-02],
       [1.969e+01, 2.125e+01, 1.300e+02, ..., 2.430e-01, 3.613e-01,
        8.758e-02],
       ...,
       [1.660e+01, 2.808e+01, 1.083e+02, ..., 1.418e-01, 2.218e-01,
        7.820e-02],
       [2.060e+01, 2.933e+01, 1.401e+02, ..., 2.650e-01, 4.087e-01,
        1.240e-01],
       [7.760e+00, 2.454e+01, 4.792e+01, ..., 0.000e+00, 2.871e-01,
        7.039e-02]])

```

```
y = np.asanyarray(data["diagnosis"])  #Dependent variable
```

```
y
```

```

array([1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 0,
0,
       1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1,
1,
       1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 0, 1, 0, 1,
1,
       0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 1, 1, 0, 1, 1,
1,
       0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1,
0,
       0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1, 0, 1, 0,
1,
       1, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
0,
       0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0,
0,
       0, 1, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1,
1,
       1, 1, 0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 1,

```

```

1,      0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 0,
0,      0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
1,      1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0,
0,      0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0,
0,      0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1,
1,      1, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0,
1,      1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 1, 1, 0, 1,
1,      0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0,
0,      0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0,
1,      0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0,
0,      0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1,
1,      0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
0,      0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0,
0,      0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0,
0,      0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0,      0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0])

```

*# Split the dataset into training and testing sets*

```

x_train,x_test,y_train,y_test =
train_test_split(x,y,test_size=0.2,random_state=40)

```

```

print(x_train.shape, y_train.shape)

```

```

print(x_test.shape, y_test.shape)

```

```

(455, 30) (455,)

```

```

(114, 30) (114,)

```

*# Define the model architecture*

```

model = Sequential()

```

```

model.add(Dense(16, activation = "relu", input_dim = 30)) # inpte

```

*layer*

```

model.add(Dense(32, activation = "relu"))

```

*#1st hidden*

*layer*

```

model.add(Dense(1, activation = "sigmoid"))

```

*#output*

*layer*

```

# Compile the model
model.compile(optimizer="adam",          # optimizer adjust the
              loss = "binary_crossentropy", # For integer target we
              using "sparse_categorical_crossentropy"
              metrics = ['accuracy'])    # For finalizing the model
& make it completely ready for use

history = model.fit(x_train, y_train, batch_size=32, epochs=100,
                    verbose=0, validation_data=(x_test, y_test))
history

<keras.callbacks.History at 0x1b76542c130>

# Evaluate the model on training and testing data
train_loss, train_accuracy = model.evaluate(x_train, y_train,
                                             verbose=0)
test_loss, test_accuracy = model.evaluate(x_test, y_test, verbose=0)

# Calculating F1-Score
y_train_pred = model.predict(x_train).round()
y_test_pred = model.predict(x_test).round()

15/15 [=====] - 0s 1ms/step
4/4 [=====] - 0s 0s/step

f1_train = f1_score(y_train, y_train_pred)
f1_test = f1_score(y_test, y_test_pred)

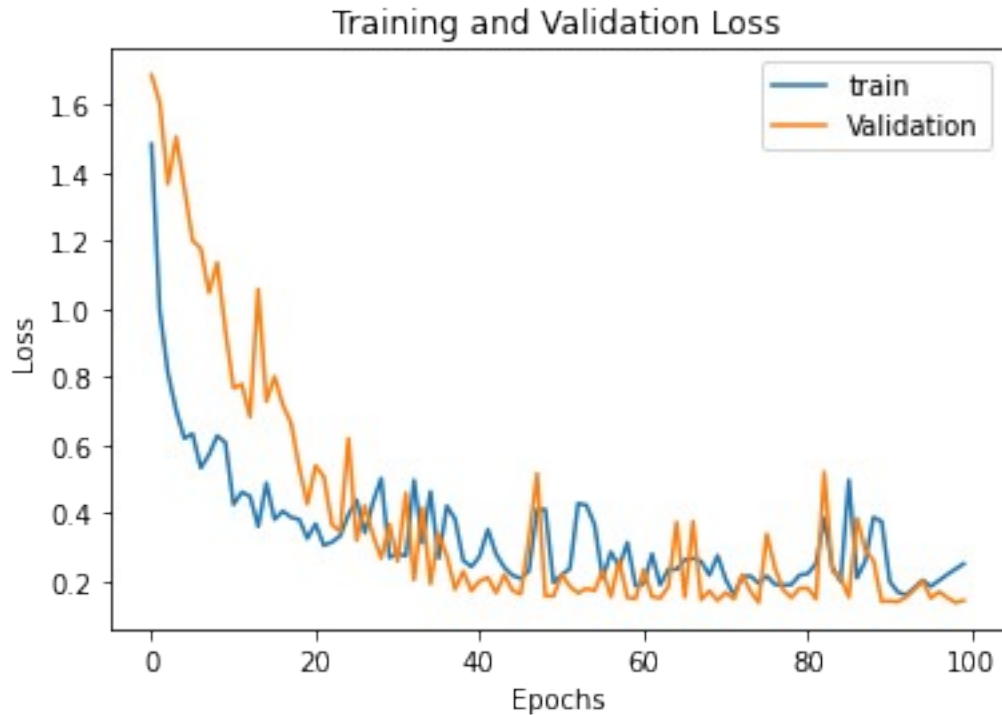
# Print the evaluation metrics
print("Train Loss:", train_loss*100)
print("Test Loss:", test_loss*100)
print("Train Accuracy:", train_accuracy*100)
print("Test Accuracy:", test_accuracy*100)
print("Train F1-Score:", f1_train*100)
print("Test F1-Score:", f1_test*100)

Train Loss: 17.658890783786774
Test Loss: 14.237135648727417
Train Accuracy: 92.74725317955017
Test Accuracy: 95.61403393745422
Train F1-Score: 89.58990536277604
Test F1-Score: 93.33333333333333

# Plotting the loss function
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('Training and Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend(["train", "Validation"], loc="upper right")

```

<matplotlib.legend.Legend at 0x1b76533e520>



```
# Plotting the accuracy function
print(history.history.keys())
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.title('Training and Validation Accuracy')
plt.xlabel('Epochs')
plt.ylabel('accuracy')
plt.legend(["train", "Validation"], loc="lower right")

dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])

<matplotlib.legend.Legend at 0x1b7673ce2e0>
```





*#F1-Score for both Training and Testing*

```
print("Train F1-Score:", f1_train*100)
```

```
print("Test F1-Score:", f1_test*100)
```

Train F1-Score: 89.58990536277604

Test F1-Score: 93.33333333333333