



Deep Learning Project Report

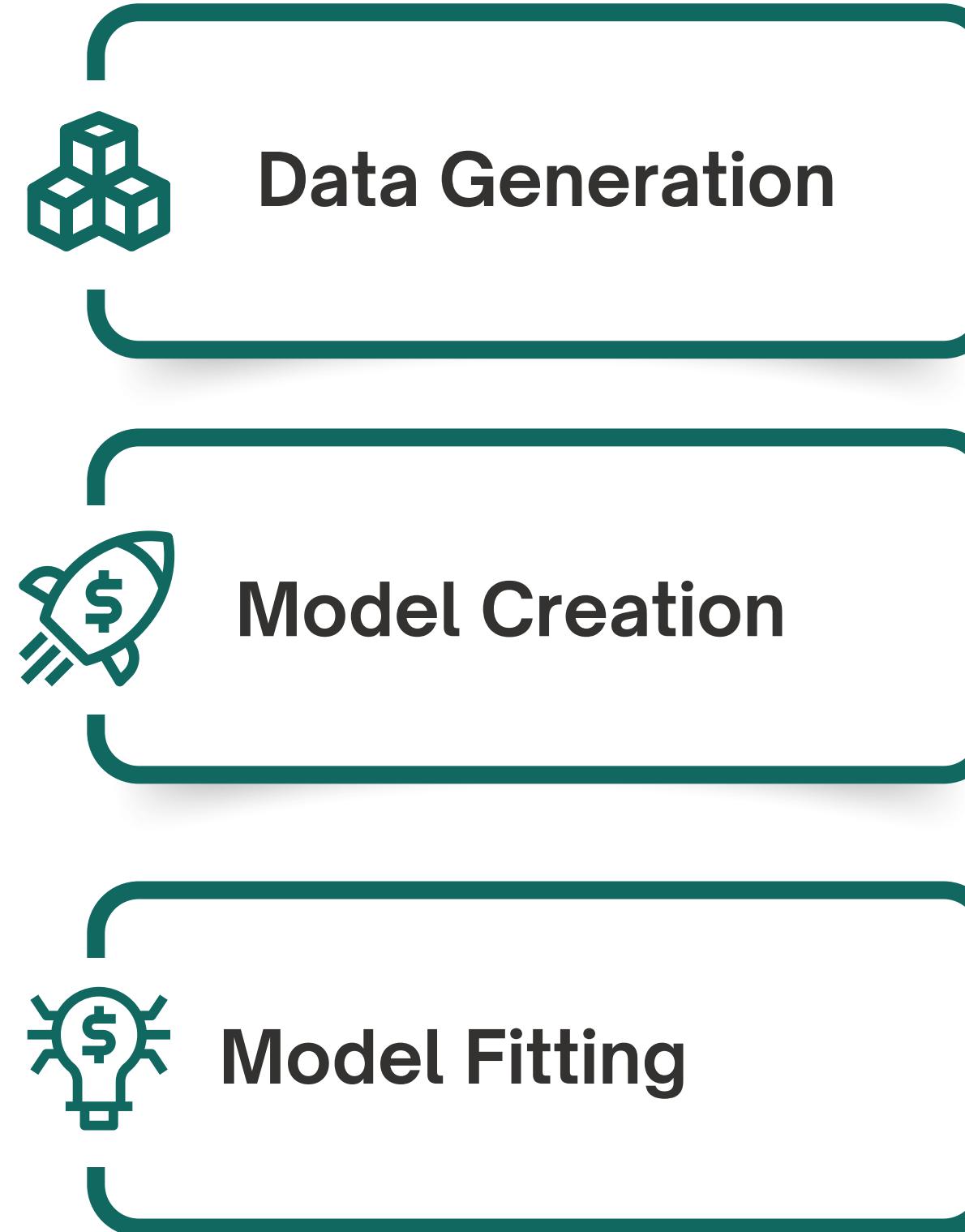
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Objective:

Consider a single point load of known magnitude at the top right end of the beam.
Generate the random topologies of the beam by altering the positions, sizes, and quantities of
the holes in a random manner. Use DeepOnet to predict the displacement field and compare
its performance with ANN.



Workflow





Data Generation



Data Generation

Generating property files:

We obtained a folder containing a file named prop.dat. which contains young's modulus of each node , to create voids, we give value 5 to that node. Ultimately, we produced 1000 new files. Furthermore, we adjusted the second-to-last line of each duplicated file, setting the Poisson's ratio to 0.33.

Generating displacement files:

Using these 1000 files we generated 1000 displacement files which we will use as output of our Neural Network. Below you can find code snippet for generating these displacement files.

```
import os
for i in range(1000):
    os.system(f'cp prop{i}.dat prop.dat')
    os.system('./main.e > shape')
    os.system(f'cp displacement displacement{i}')
```



Model Generation

DeepOnet

Model Summary

Layer (type)	Output Shape	Param #	Connected to
Input_2 (InputLayer)	[(None, 400, 2)]	0	[]
Input_1 (InputLayer)	[(None, 400)]	0	[]
flatten_3 (Flatten)	(None, 800)	0	['Input_2[0][0]']
dense_37 (Dense)	(None, 800)	320800	['Input_1[0][0]']
dense_42 (Dense)	(None, 1600)	1281600	['flatten_3[0][0]']
dense_38 (Dense)	(None, 1600)	1281600	['dense_37[0][0]']
dense_43 (Dense)	(None, 2000)	3202000	['dense_42[0][0]']
dense_39 (Dense)	(None, 2400)	3842400	['dense_38[0][0]']
dense_44 (Dense)	(None, 2400)	4802400	['dense_43[0][0]']
dense_40 (Dense)	(None, 2000)	4802000	['dense_39[0][0]']
dense_45 (Dense)	(None, 2000)	4802000	['dense_44[0][0]']
dense_41 (Dense)	(None, 1200)	2401200	['dense_40[0][0]']
dense_46 (Dense)	(None, 1600)	3201600	['dense_45[0][0]']
concatenate_3 (Concatenate)	(None, 2800)	0	['dense_41[0][0]', 'dense_46[0][0]']
dense_47 (Dense)	(None, 1681)	4708481	['concatenate_3[0][0]']

ANN

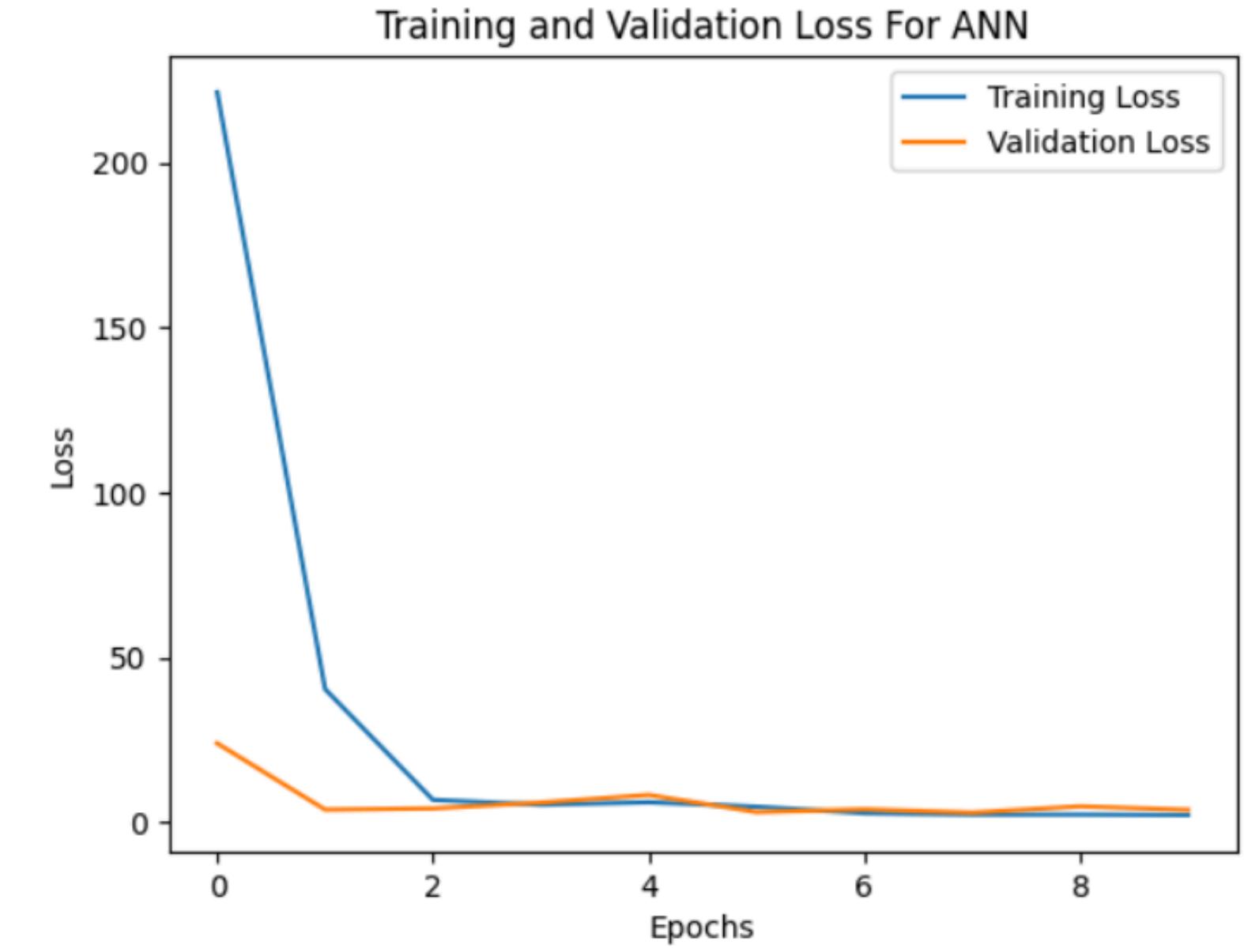
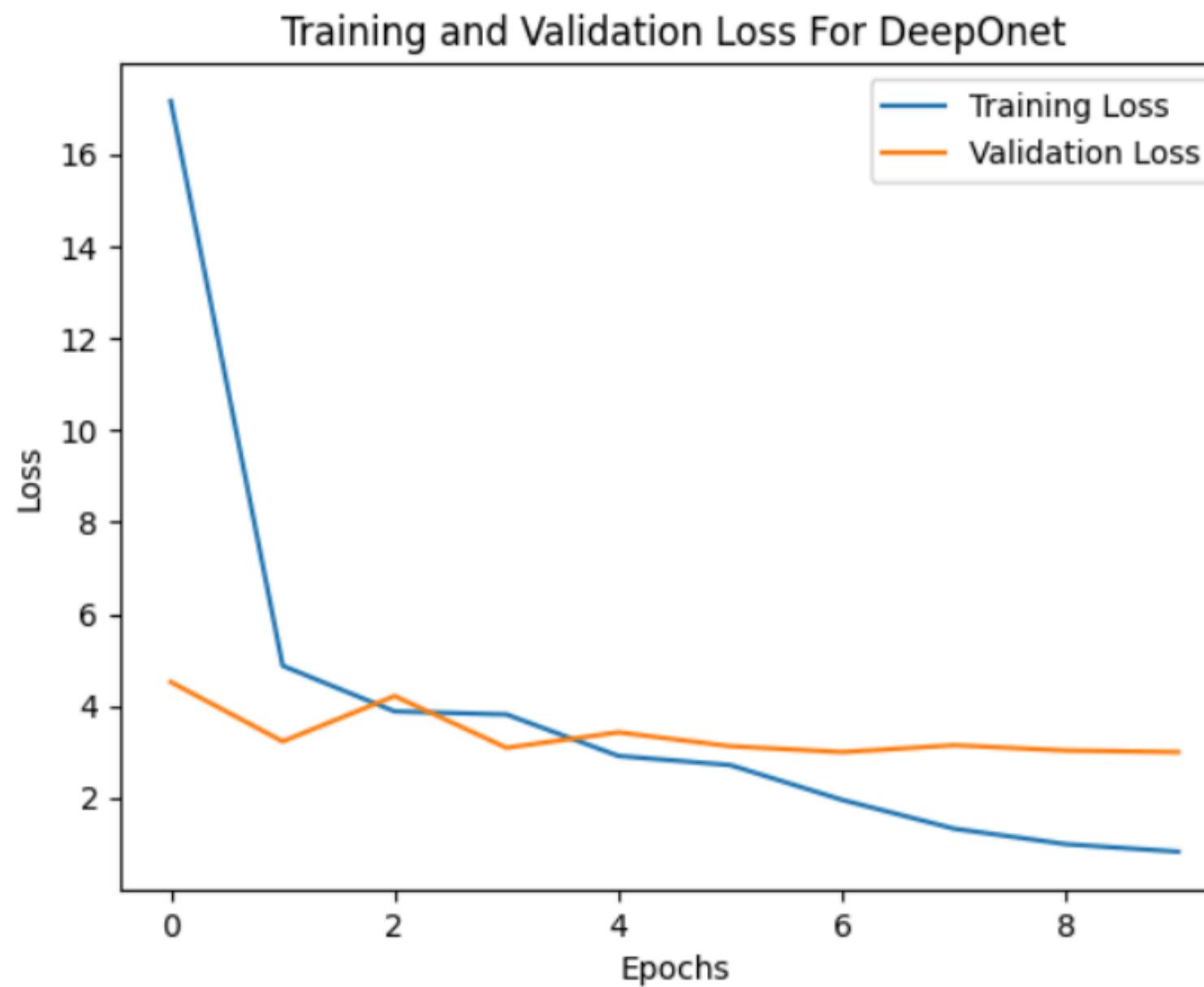
Model Summary

```
[26] # Define the ANN model
model1 = Sequential()
model1.add(Dense(400, input_shape=input_shape, activation='relu', name = 'Input', kernel_initializer=initializer)) #input layer
model1.add(Dense(1000, activation='relu', kernel_initializer=initializer))
model1.add(Dense(1600, activation='relu', kernel_initializer=initializer))
model1.add(Dense(2400, activation='relu', kernel_initializer=initializer))
model1.add(Dense(3000, activation='relu', kernel_initializer=initializer))
model1.add(Dense(2400, activation='relu', kernel_initializer=initializer))
model1.add(Dense(2000, activation='relu', kernel_initializer=initializer))
model1.add(Dense(1681, name = 'Output')) # output layer
model1.summary()
```

Model: "sequential_3"

Layer (type)	Output Shape	Param #
Input (Dense)	(None, 400)	160400
dense_17 (Dense)	(None, 1000)	401000
dense_18 (Dense)	(None, 1600)	1601600
dense_19 (Dense)	(None, 2400)	3842400
dense_20 (Dense)	(None, 3000)	7203000
dense_21 (Dense)	(None, 2400)	7202400
dense_22 (Dense)	(None, 2000)	4802000
Output (Dense)	(None, 1681)	3363681

Comparing DeepOnet with ANN



Comparing DeepOnet with ANN

Predictions of DeepOnet

```
▶ model2.predict([X_test,X_test2])  
[ 7/7 [=====] - 0s 4ms/step  
array([[ 0.08046146, -0.03824379,  0.02767699, ...,  6.8436694 ,  
       7.0134106 ,  7.023118 ],  
      [ 0.05779466,  0.03676069,  0.03526133, ...,  5.0360136 ,  
       5.2075396 ,  5.2245035 ],  
      [ 0.05487629,  0.13955292,  0.08827313, ..., 10.66084 ,  
       10.606489 ,  10.786817 ],  
      ...,  
      [ 0.02120928,  0.09356874,  0.0258595 , ...,  7.3240466 ,  
       7.5284977 ,  7.482904 ],  
      [ 0.09149098,  0.24523059,  0.13268027, ..., 14.68873 ,  
       15.205969 ,  14.894193 ],  
      [-0.02354654,  0.10249826,  0.03398355, ...,  6.4211154 ,  
       6.5375595 ,  6.524995 ]], dtype=float32)
```

```
[ ] y_test  
array([[ 0.        ,  0.        ,  0.        , ...,  6.408853,  6.505128,  
       6.638718],  
      [ 0.        ,  0.        ,  0.        , ...,  6.031897,  6.128132,  
       6.261719],  
      [ 0.        ,  0.        ,  0.        , ...,  9.539615,  9.639187,  
       9.773179],  
      ...,  
      [ 0.        ,  0.        ,  0.        , ...,  5.966433,  6.062663,  
       6.19624 ],  
      [ 0.        ,  0.        ,  0.        , ..., 12.36107 , 12.46937 ,  
       12.60473 ],  
      [ 0.        ,  0.        ,  0.        , ...,  6.146101,  6.24233 ,  
       6.375905]])
```

Predictions of ANN

```
▶ # FOR ANN  
model1.predict(X_test)  
[ 7/7 [=====] - 0s 3ms/step  
array([[ -0.04480736, -0.12804836,  0.04999661, ...,  6.8914566 ,  
       6.9320283 ,  7.0880456 ],  
      [ -0.10278296, -0.20266807,  0.06799543, ..., 12.008759 ,  
       12.027733 , 12.257994 ],  
      [ -0.07187152, -0.16343889,  0.05670393, ...,  8.627366 ,  
       8.659404 ,  8.841123 ],  
      ...,  
      [ -0.05123809, -0.13325232,  0.05823982, ..., 10.443156 ,  
       10.47608 , 10.68141 ],  
      [ -0.14568543, -0.26829395,  0.0731923 , ..., 13.024769 ,  
       13.033874 , 13.2835 ],  
      [ -0.08441925, -0.17755523,  0.06273252, ..., 10.881477 ,  
       10.905769 , 11.118708 ]], dtype=float32)
```

```
[33] y_test  
array([[ 0.        ,  0.        ,  0.        , ...,  6.752358,  6.84864 ,  
       6.982224],  
      [ 0.        ,  0.        ,  0.        , ...,  7.736916,  7.833137,  
       7.966711],  
      [ 0.        ,  0.        ,  0.        , ...,  6.514533,  6.610765,  
       6.744342],  
      ...,  
      [ 0.        ,  0.        ,  0.        , ..., 12.89767 , 12.9942 ,  
       13.12775 ],  
      [ 0.        ,  0.        ,  0.        , ..., 12.21668 , 12.31295 ,  
       12.4465 ],  
      [ 0.        ,  0.        ,  0.        , ..., 11.31705 , 11.41336 ,  
       11.54686 ]])
```

Thank
you

