

# 2348542\_\_CIA\_\_Component\_2

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## 1 CIA Component 2

Name: Purusharth Malik

Registration No.: 2348542

### 1.0.1 Question 1

```
[26]: import numpy as np
import sympy as sp
import random
import matplotlib.pyplot as plt

# Creating the symbols
d, s = sp.symbols("d s")

# Creating the function
f_d_s = 0.1*(s**2) + d/s + 0.6*d

# Calculating the gradient of the function
grad_f = sp.lambdify((d, s), sp.derive_by_array(f_d_s, (d, s)))

# Gradient Descent parameters
alpha = 0.01
max_iter = 1000
momentum = 0.9
initial_s, initial_d = 50, 100
starting_point = sp.Matrix([initial_d, initial_s])
prev_grad = sp.Matrix([0,0])

# Values to plot the graph
d_descent, s_descent = [], []

for i in range(max_iter):
    # Computing the gradient at the current point
    grad = sp.Matrix(grad_f(starting_point[0], starting_point[1]))
    # Updating the values of d and s
    starting_point -= (momentum*grad + alpha*prev_grad)
```

```

# Keeping track of the starting points for every iteration
d_descent.append(starting_point[0])
s_descent.append(starting_point[1])
# Updating the value of the prev_grad
prev_grad = grad

# Checking for convergence
if grad.norm() < 0.02:
    print(f"Number of iterations taken to converge: {i}")
    minimum_ds = starting_point
    # Evaluate f_d_s at the minimum
    minimum_cost = f_d_s.subs({d:minimum_ds[0], s:minimum_ds[1]})
    print(f"Minimum value of d and s: {minimum_ds}")
    print(f"Minimum cost: {minimum_cost}")
    break
else:
    print("Failed to converge")

# Printing the value after the third iteration
print(f'The values after the thrid iteration were:\ns:{s_descent[2]}\nd:
↳{d_descent[2]}')

# Plotting
feature_d = np.linspace(-0.5, 2.0, 70)
feature_s = np.linspace(-0.5, 2.0, 70)

# Creating 2-D grid of features
[D, S] = np.meshgrid(feature_d, feature_s)

fig, ax = plt.subplots(1, 1)

func = 0.1*(S**2) + D/S + 0.6*D

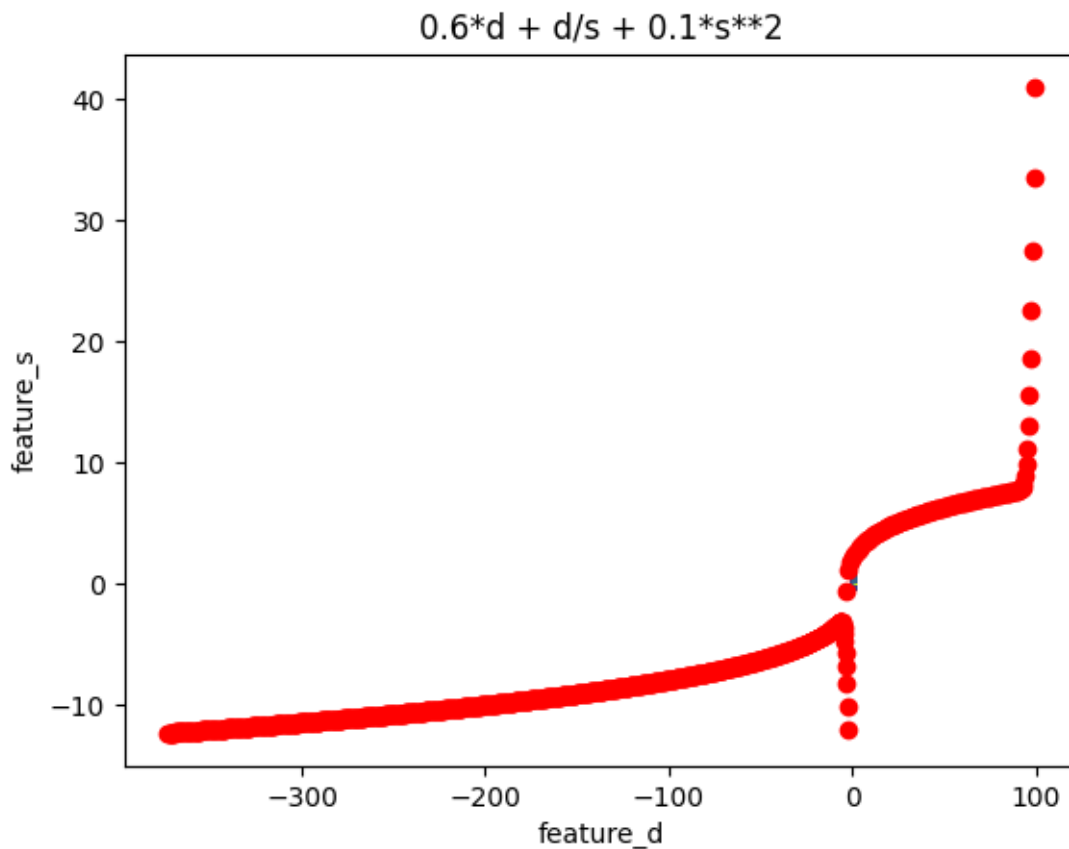
# plots filled contour plot
ax.contourf(D, S, func)
ax.scatter(d_descent, s_descent, c='r')

ax.set_title(f_d_s)
ax.set_xlabel('feature_d')
ax.set_ylabel('feature_s')

plt.show()

```

Failed to converge  
The values after the thrid iteration were:  
s:27.5518410771068  
d:98.3008410801840



## 1.0.2 Question 2

### 1.0.3 Data Preprocessing

```
[18]: import pandas as pd
import seaborn as sns

df = pd.read_csv('GME_stock.csv')
df
```

```
[18]:
```

	date	open_price	high_price	low_price	close_price	\
0	2021-01-28	265.000000	483.000000	112.250000	193.600006	
1	2021-01-27	354.829987	380.000000	249.000000	347.510010	
2	2021-01-26	88.559998	150.000000	80.199997	147.979996	
3	2021-01-25	96.730003	159.179993	61.130001	76.790001	
4	2021-01-22	42.590000	76.760002	42.320000	65.010002	
...	...	...	...	...	...	
4768	2002-02-20	9.600000	9.875000	9.525000	9.875000	
4769	2002-02-19	9.900000	9.900000	9.375000	9.550000	
4770	2002-02-15	10.000000	10.025000	9.850000	9.950000	

4771	2002-02-14	10.175000	10.195000	9.925000	10.000000
4772	2002-02-13	9.625000	10.060000	9.525000	10.050000

	volume	adjclose_price
0	58815800.0	193.600006
1	93396700.0	347.510010
2	178588000.0	147.979996
3	177874000.0	76.790001
4	196784300.0	65.010002
...	...	...
4768	1723200.0	6.648838
4769	1852600.0	6.430017
4770	2097400.0	6.699336
4771	2755400.0	6.733003
4772	19054000.0	6.766666

[4773 rows x 7 columns]

```
[20]: # Checking for missing values
df.isna().sum()
```

```
[20]: date            0
open_price          0
high_price          0
low_price           0
close_price         0
volume             0
adjclose_price      0
dtype: int64
```

There are no missing values and no categorical variables. We will proceed by dropping the date feature and scaling the numerical (all) features.

```
[28]: df.drop('date', inplace=True, axis=1)
```

```
[30]: # Scaling the features
from sklearn.preprocessing import StandardScaler

ss = StandardScaler()
df = pd.DataFrame(ss.fit_transform(df), columns=df.columns)
df
```

```
[30]:
```

	open_price	high_price	low_price	close_price	volume	\
0	16.455242	28.337343	6.570575	11.742172	8.465412	
1	22.568477	21.982661	16.605972	22.347605	13.747883	
2	4.447901	7.792593	4.218586	8.598647	26.761441	
3	5.003898	8.358962	2.819135	3.693177	26.652373	
4	1.319487	3.273982	1.438763	2.881456	29.541052	

...	...	...	...	...	...
4768	-0.925594	-0.852551	-0.967897	-0.917716	-0.255876
4769	-0.905179	-0.851009	-0.978905	-0.940110	-0.236109
4770	-0.898373	-0.843297	-0.944047	-0.912548	-0.198715
4771	-0.886464	-0.832808	-0.938543	-0.909102	-0.098201
4772	-0.923893	-0.841137	-0.967897	-0.905657	2.391523

	adjclose_price
0	16.037784
1	30.005731
2	11.897585
3	5.436809
4	4.367727

...	...
4768	-0.928782
4769	-0.948641
4770	-0.924199
4771	-0.921143
4772	-0.918088

[4773 rows x 6 columns]

For anomaly detection, since we have multivariate data, we will use Hotelling's T2 on the multidimensional space of PCA.

```
[34]: from pca import pca

# Instantiating PCA
model = pca(normalize=True, detect_outliers=['ht2'], n_std=2)

# Getting the results
results = model.fit_transform(df)
```

```
[pca] >Extracting column labels from dataframe.
[pca] >Extracting row labels from dataframe.
[pca] >Normalizing input data per feature (zero mean and unit variance)..
[pca] >The PCA reduction is performed to capture [95.0%] explained variance
using the [6] columns of the input data.
[pca] >Fit using PCA.
[pca] >Compute loadings and PCs.
[pca] >Compute explained variance.
[pca] >Number of components is [2] that covers the [95.00%] explained variance.
[pca] >The PCA reduction is performed on the [6] columns of the input dataframe.
[pca] >Fit using PCA.
[pca] >Compute loadings and PCs.
[pca] >Outlier detection using Hotelling T2 test with alpha=[0.05] and
n_components=[2]
[pca] >Multiple test correction applied for Hotelling T2 test: [fdr_bh]
```

```
[37]: # Looking at the p-values of outliers
outliers = results['outliers'][results['outliers']['y_bool']]
outliers
```

```
[37]:
```

	y_proba	p_raw	y_score	y_bool
0	2.388153e-298	2.501731e-301	1397.423597	True
1	8.623688e-302	5.420294e-305	1414.321950	True
2	0.000000e+00	0.000000e+00	2763.102112	True
3	0.000000e+00	0.000000e+00	1453.638363	True
4	9.563799e-302	8.014916e-305	1413.538534	True
5	6.487987e-04	1.495241e-06	32.523428	True
7	2.306379e-07	3.865709e-10	49.858141	True
9	2.123054e-12	3.113634e-15	74.083336	True
10	2.585578e-296	3.250255e-299	1387.675776	True
75	6.040898e-07	1.139076e-09	47.608112	True
76	6.386411e-07	1.338029e-09	47.272578	True

```
[40]: # Removing the outliers
df = df.drop(outliers.index)
df
```

```
[40]:
```

	open_price	high_price	low_price	close_price	volume	adjclose_price
6	0.964248	1.079457	0.979373	1.097461	4.593941	2.018106
8	1.040468	1.052311	0.828934	0.848019	6.622614	1.689576
11	-0.220561	-0.203201	-0.249091	-0.223481	0.559464	0.278352
12	-0.257991	-0.187777	-0.271841	-0.224170	1.761187	0.277444
13	-0.341696	-0.332763	-0.413474	-0.379210	0.468390	0.073248
...	...	...	...	...	...	...
4768	-0.925594	-0.852551	-0.967897	-0.917716	-0.255876	-0.928782
4769	-0.905179	-0.851009	-0.978905	-0.940110	-0.236109	-0.948641
4770	-0.898373	-0.843297	-0.944047	-0.912548	-0.198715	-0.924199
4771	-0.886464	-0.832808	-0.938543	-0.909102	-0.098201	-0.921143
4772	-0.923893	-0.841137	-0.967897	-0.905657	2.391523	-0.918088

[4762 rows x 6 columns]

#### 1.0.4 Splitting the dataset

```
[47]: from sklearn.model_selection import train_test_split

X = df.drop('close_price', axis=1).values
y = df['close_price']

# Splitting the dataset into training and testing set
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2)

# Splitting the training set into training and validation set
```

```
X_val, X_train, y_val, y_train = train_test_split(X_train, y_train, test_size=0.
↪9)

X_train.shape, X_val.shape, X_test.shape
```

[47]: ((3429, 5), (380, 5), (953, 5))

### 1.0.5 Building the neural network

```
[55]: import tensorflow as tf
from tensorflow.keras.layers import *

class MyModel(tf.keras.models.Model):

    def __init__(self):
        super().__init__()
        self.dense_1 = Dense(128, activation=tf.nn.relu)
        self.dense_2 = Dense(128, activation=tf.nn.relu)
        self.dense_3 = Dense(128, activation=tf.nn.relu)
        self.dropout_1 = Dropout(0.2)
        self.dropout_2 = Dropout(0.2)
        self.regression = Dense(1)

    def call(self, input_tensor):
        x = self.dense_1(input_tensor)
        x = self.dropout_1(x)
        x = self.dense_2(x)
        x = self.dropout_2(x)
        x = self.dense_3(x)
        return self.regression(x)

model = MyModel()

model.compile(loss=tf.keras.losses.MeanSquaredError(),
              optimizer=tf.keras.optimizers.Adam(),
              metrics=tf.keras.metrics.MeanSquaredError())

history = model.fit(X_train, y_train, epochs=100,
                    validation_data=(X_val, y_val),
                    batch_size=16)
```

Epoch 1/100

215/215 [=====] - 2s 3ms/step - loss: 0.0416 -  
mean\_squared\_error: 0.0416 - val\_loss: 0.0183 - val\_mean\_squared\_error: 0.0183

Epoch 2/100

215/215 [=====] - 1s 3ms/step - loss: 0.0114 -  
mean\_squared\_error: 0.0114 - val\_loss: 0.0089 - val\_mean\_squared\_error: 0.0089

Epoch 3/100

215/215 [=====] - 1s 2ms/step - loss: 0.0107 -  
 mean\_squared\_error: 0.0107 - val\_loss: 0.0194 - val\_mean\_squared\_error: 0.0194  
 Epoch 4/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0081 -  
 mean\_squared\_error: 0.0081 - val\_loss: 0.0099 - val\_mean\_squared\_error: 0.0099  
 Epoch 5/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0087 -  
 mean\_squared\_error: 0.0087 - val\_loss: 0.0164 - val\_mean\_squared\_error: 0.0164  
 Epoch 6/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0071 -  
 mean\_squared\_error: 0.0071 - val\_loss: 0.0260 - val\_mean\_squared\_error: 0.0260  
 Epoch 7/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0075 -  
 mean\_squared\_error: 0.0075 - val\_loss: 0.0345 - val\_mean\_squared\_error: 0.0345  
 Epoch 8/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0066 -  
 mean\_squared\_error: 0.0066 - val\_loss: 0.0299 - val\_mean\_squared\_error: 0.0299  
 Epoch 9/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0068 -  
 mean\_squared\_error: 0.0068 - val\_loss: 0.0222 - val\_mean\_squared\_error: 0.0222  
 Epoch 10/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0074 -  
 mean\_squared\_error: 0.0074 - val\_loss: 0.0139 - val\_mean\_squared\_error: 0.0139  
 Epoch 11/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0053 -  
 mean\_squared\_error: 0.0053 - val\_loss: 0.0184 - val\_mean\_squared\_error: 0.0184  
 Epoch 12/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0048 -  
 mean\_squared\_error: 0.0048 - val\_loss: 0.0358 - val\_mean\_squared\_error: 0.0358  
 Epoch 13/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0051 -  
 mean\_squared\_error: 0.0051 - val\_loss: 0.0319 - val\_mean\_squared\_error: 0.0319  
 Epoch 14/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0051 -  
 mean\_squared\_error: 0.0051 - val\_loss: 0.0296 - val\_mean\_squared\_error: 0.0296  
 Epoch 15/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0051 -  
 mean\_squared\_error: 0.0051 - val\_loss: 0.0277 - val\_mean\_squared\_error: 0.0277  
 Epoch 16/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0052 -  
 mean\_squared\_error: 0.0052 - val\_loss: 0.0179 - val\_mean\_squared\_error: 0.0179  
 Epoch 17/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0044 -  
 mean\_squared\_error: 0.0044 - val\_loss: 0.0274 - val\_mean\_squared\_error: 0.0274  
 Epoch 18/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0043 -  
 mean\_squared\_error: 0.0043 - val\_loss: 0.0273 - val\_mean\_squared\_error: 0.0273  
 Epoch 19/100



```

215/215 [=====] - 1s 2ms/step - loss: 0.0043 -
mean_squared_error: 0.0043 - val_loss: 0.0442 - val_mean_squared_error: 0.0442
Epoch 20/100
215/215 [=====] - 0s 2ms/step - loss: 0.0041 -
mean_squared_error: 0.0041 - val_loss: 0.0248 - val_mean_squared_error: 0.0248
Epoch 21/100
215/215 [=====] - 1s 2ms/step - loss: 0.0038 -
mean_squared_error: 0.0038 - val_loss: 0.0275 - val_mean_squared_error: 0.0275
Epoch 22/100
215/215 [=====] - 1s 2ms/step - loss: 0.0041 -
mean_squared_error: 0.0041 - val_loss: 0.0396 - val_mean_squared_error: 0.0396
Epoch 23/100
215/215 [=====] - 1s 2ms/step - loss: 0.0036 -
mean_squared_error: 0.0036 - val_loss: 0.0245 - val_mean_squared_error: 0.0245
Epoch 24/100
215/215 [=====] - 1s 2ms/step - loss: 0.0036 -
mean_squared_error: 0.0036 - val_loss: 0.0275 - val_mean_squared_error: 0.0275
Epoch 25/100
215/215 [=====] - 1s 2ms/step - loss: 0.0038 -
mean_squared_error: 0.0038 - val_loss: 0.0225 - val_mean_squared_error: 0.0225
Epoch 26/100
215/215 [=====] - 1s 2ms/step - loss: 0.0034 -
mean_squared_error: 0.0034 - val_loss: 0.0284 - val_mean_squared_error: 0.0284
Epoch 27/100
215/215 [=====] - 1s 2ms/step - loss: 0.0034 -
mean_squared_error: 0.0034 - val_loss: 0.0281 - val_mean_squared_error: 0.0281
Epoch 28/100
215/215 [=====] - 1s 2ms/step - loss: 0.0030 -
mean_squared_error: 0.0030 - val_loss: 0.0250 - val_mean_squared_error: 0.0250
Epoch 29/100
215/215 [=====] - 1s 2ms/step - loss: 0.0030 -
mean_squared_error: 0.0030 - val_loss: 0.0312 - val_mean_squared_error: 0.0312
Epoch 30/100
215/215 [=====] - 1s 2ms/step - loss: 0.0028 -
mean_squared_error: 0.0028 - val_loss: 0.0230 - val_mean_squared_error: 0.0230
Epoch 31/100
215/215 [=====] - 1s 2ms/step - loss: 0.0034 -
mean_squared_error: 0.0034 - val_loss: 0.0306 - val_mean_squared_error: 0.0306
Epoch 32/100
215/215 [=====] - 1s 2ms/step - loss: 0.0027 -
mean_squared_error: 0.0027 - val_loss: 0.0305 - val_mean_squared_error: 0.0305
Epoch 33/100
215/215 [=====] - 1s 2ms/step - loss: 0.0027 -
mean_squared_error: 0.0027 - val_loss: 0.0267 - val_mean_squared_error: 0.0267
Epoch 34/100
215/215 [=====] - 1s 2ms/step - loss: 0.0027 -
mean_squared_error: 0.0027 - val_loss: 0.0208 - val_mean_squared_error: 0.0208
Epoch 35/100

```

215/215 [=====] - 1s 2ms/step - loss: 0.0030 -  
 mean\_squared\_error: 0.0030 - val\_loss: 0.0295 - val\_mean\_squared\_error: 0.0295  
 Epoch 36/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0026 -  
 mean\_squared\_error: 0.0026 - val\_loss: 0.0258 - val\_mean\_squared\_error: 0.0258  
 Epoch 37/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0028 -  
 mean\_squared\_error: 0.0028 - val\_loss: 0.0277 - val\_mean\_squared\_error: 0.0277  
 Epoch 38/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0027 -  
 mean\_squared\_error: 0.0027 - val\_loss: 0.0275 - val\_mean\_squared\_error: 0.0275  
 Epoch 39/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0027 -  
 mean\_squared\_error: 0.0027 - val\_loss: 0.0318 - val\_mean\_squared\_error: 0.0318  
 Epoch 40/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0024 -  
 mean\_squared\_error: 0.0024 - val\_loss: 0.0330 - val\_mean\_squared\_error: 0.0330  
 Epoch 41/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0029 -  
 mean\_squared\_error: 0.0029 - val\_loss: 0.0246 - val\_mean\_squared\_error: 0.0246  
 Epoch 42/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0025 -  
 mean\_squared\_error: 0.0025 - val\_loss: 0.0343 - val\_mean\_squared\_error: 0.0343  
 Epoch 43/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0035 -  
 mean\_squared\_error: 0.0035 - val\_loss: 0.0247 - val\_mean\_squared\_error: 0.0247  
 Epoch 44/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0023 -  
 mean\_squared\_error: 0.0023 - val\_loss: 0.0258 - val\_mean\_squared\_error: 0.0258  
 Epoch 45/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0029 -  
 mean\_squared\_error: 0.0029 - val\_loss: 0.0318 - val\_mean\_squared\_error: 0.0318  
 Epoch 46/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0023 -  
 mean\_squared\_error: 0.0023 - val\_loss: 0.0272 - val\_mean\_squared\_error: 0.0272  
 Epoch 47/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0021 -  
 mean\_squared\_error: 0.0021 - val\_loss: 0.0255 - val\_mean\_squared\_error: 0.0255  
 Epoch 48/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0024 -  
 mean\_squared\_error: 0.0024 - val\_loss: 0.0259 - val\_mean\_squared\_error: 0.0259  
 Epoch 49/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0030 -  
 mean\_squared\_error: 0.0030 - val\_loss: 0.0172 - val\_mean\_squared\_error: 0.0172  
 Epoch 50/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0023 -  
 mean\_squared\_error: 0.0023 - val\_loss: 0.0252 - val\_mean\_squared\_error: 0.0252  
 Epoch 51/100

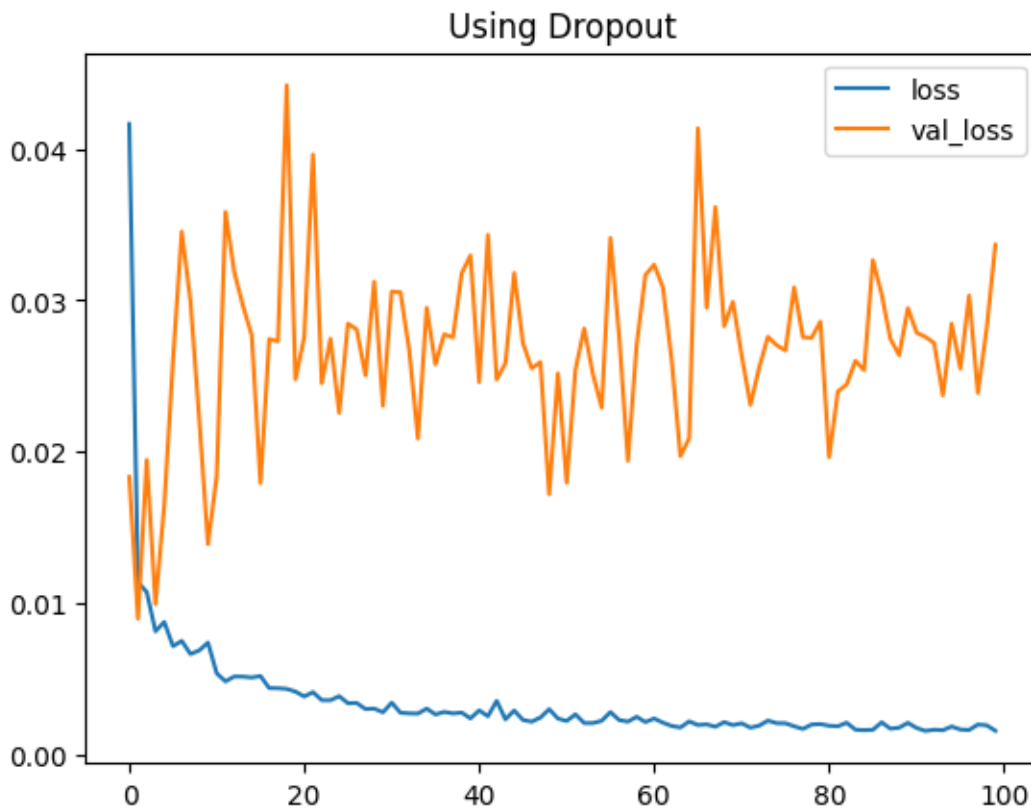
215/215 [=====] - 1s 2ms/step - loss: 0.0022 -  
mean\_squared\_error: 0.0022 - val\_loss: 0.0179 - val\_mean\_squared\_error: 0.0179  
Epoch 52/100  
215/215 [=====] - 1s 3ms/step - loss: 0.0026 -  
mean\_squared\_error: 0.0026 - val\_loss: 0.0254 - val\_mean\_squared\_error: 0.0254  
Epoch 53/100  
215/215 [=====] - 1s 3ms/step - loss: 0.0020 -  
mean\_squared\_error: 0.0020 - val\_loss: 0.0281 - val\_mean\_squared\_error: 0.0281  
Epoch 54/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0020 -  
mean\_squared\_error: 0.0020 - val\_loss: 0.0251 - val\_mean\_squared\_error: 0.0251  
Epoch 55/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0022 -  
mean\_squared\_error: 0.0022 - val\_loss: 0.0229 - val\_mean\_squared\_error: 0.0229  
Epoch 56/100  
215/215 [=====] - 1s 3ms/step - loss: 0.0028 -  
mean\_squared\_error: 0.0028 - val\_loss: 0.0341 - val\_mean\_squared\_error: 0.0341  
Epoch 57/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0022 -  
mean\_squared\_error: 0.0022 - val\_loss: 0.0276 - val\_mean\_squared\_error: 0.0276  
Epoch 58/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0021 -  
mean\_squared\_error: 0.0021 - val\_loss: 0.0194 - val\_mean\_squared\_error: 0.0194  
Epoch 59/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0025 -  
mean\_squared\_error: 0.0025 - val\_loss: 0.0270 - val\_mean\_squared\_error: 0.0270  
Epoch 60/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0021 -  
mean\_squared\_error: 0.0021 - val\_loss: 0.0317 - val\_mean\_squared\_error: 0.0317  
Epoch 61/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0023 -  
mean\_squared\_error: 0.0023 - val\_loss: 0.0323 - val\_mean\_squared\_error: 0.0323  
Epoch 62/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0021 -  
mean\_squared\_error: 0.0021 - val\_loss: 0.0308 - val\_mean\_squared\_error: 0.0308  
Epoch 63/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0019 -  
mean\_squared\_error: 0.0019 - val\_loss: 0.0260 - val\_mean\_squared\_error: 0.0260  
Epoch 64/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0018 -  
mean\_squared\_error: 0.0018 - val\_loss: 0.0197 - val\_mean\_squared\_error: 0.0197  
Epoch 65/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0022 -  
mean\_squared\_error: 0.0022 - val\_loss: 0.0209 - val\_mean\_squared\_error: 0.0209  
Epoch 66/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0019 -  
mean\_squared\_error: 0.0019 - val\_loss: 0.0414 - val\_mean\_squared\_error: 0.0414  
Epoch 67/100

215/215 [=====] - 1s 2ms/step - loss: 0.0020 -  
mean\_squared\_error: 0.0020 - val\_loss: 0.0295 - val\_mean\_squared\_error: 0.0295  
Epoch 68/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0018 -  
mean\_squared\_error: 0.0018 - val\_loss: 0.0362 - val\_mean\_squared\_error: 0.0362  
Epoch 69/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0021 -  
mean\_squared\_error: 0.0021 - val\_loss: 0.0283 - val\_mean\_squared\_error: 0.0283  
Epoch 70/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0019 -  
mean\_squared\_error: 0.0019 - val\_loss: 0.0299 - val\_mean\_squared\_error: 0.0299  
Epoch 71/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0020 -  
mean\_squared\_error: 0.0020 - val\_loss: 0.0264 - val\_mean\_squared\_error: 0.0264  
Epoch 72/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0017 -  
mean\_squared\_error: 0.0017 - val\_loss: 0.0231 - val\_mean\_squared\_error: 0.0231  
Epoch 73/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0019 -  
mean\_squared\_error: 0.0019 - val\_loss: 0.0255 - val\_mean\_squared\_error: 0.0255  
Epoch 74/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0022 -  
mean\_squared\_error: 0.0022 - val\_loss: 0.0276 - val\_mean\_squared\_error: 0.0276  
Epoch 75/100  
215/215 [=====] - 1s 3ms/step - loss: 0.0020 -  
mean\_squared\_error: 0.0020 - val\_loss: 0.0270 - val\_mean\_squared\_error: 0.0270  
Epoch 76/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0020 -  
mean\_squared\_error: 0.0020 - val\_loss: 0.0267 - val\_mean\_squared\_error: 0.0267  
Epoch 77/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0018 -  
mean\_squared\_error: 0.0018 - val\_loss: 0.0308 - val\_mean\_squared\_error: 0.0308  
Epoch 78/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0017 -  
mean\_squared\_error: 0.0017 - val\_loss: 0.0276 - val\_mean\_squared\_error: 0.0276  
Epoch 79/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0019 -  
mean\_squared\_error: 0.0019 - val\_loss: 0.0275 - val\_mean\_squared\_error: 0.0275  
Epoch 80/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0020 -  
mean\_squared\_error: 0.0020 - val\_loss: 0.0286 - val\_mean\_squared\_error: 0.0286  
Epoch 81/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0019 -  
mean\_squared\_error: 0.0019 - val\_loss: 0.0196 - val\_mean\_squared\_error: 0.0196  
Epoch 82/100  
215/215 [=====] - 1s 3ms/step - loss: 0.0018 -  
mean\_squared\_error: 0.0018 - val\_loss: 0.0240 - val\_mean\_squared\_error: 0.0240  
Epoch 83/100

215/215 [=====] - 1s 3ms/step - loss: 0.0021 -  
 mean\_squared\_error: 0.0021 - val\_loss: 0.0244 - val\_mean\_squared\_error: 0.0244  
 Epoch 84/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0016 -  
 mean\_squared\_error: 0.0016 - val\_loss: 0.0260 - val\_mean\_squared\_error: 0.0260  
 Epoch 85/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0016 -  
 mean\_squared\_error: 0.0016 - val\_loss: 0.0254 - val\_mean\_squared\_error: 0.0254  
 Epoch 86/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0016 -  
 mean\_squared\_error: 0.0016 - val\_loss: 0.0327 - val\_mean\_squared\_error: 0.0327  
 Epoch 87/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0021 -  
 mean\_squared\_error: 0.0021 - val\_loss: 0.0304 - val\_mean\_squared\_error: 0.0304  
 Epoch 88/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0017 -  
 mean\_squared\_error: 0.0017 - val\_loss: 0.0274 - val\_mean\_squared\_error: 0.0274  
 Epoch 89/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0017 -  
 mean\_squared\_error: 0.0017 - val\_loss: 0.0264 - val\_mean\_squared\_error: 0.0264  
 Epoch 90/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0021 -  
 mean\_squared\_error: 0.0021 - val\_loss: 0.0295 - val\_mean\_squared\_error: 0.0295  
 Epoch 91/100  
 215/215 [=====] - 1s 3ms/step - loss: 0.0017 -  
 mean\_squared\_error: 0.0017 - val\_loss: 0.0278 - val\_mean\_squared\_error: 0.0278  
 Epoch 92/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0015 -  
 mean\_squared\_error: 0.0015 - val\_loss: 0.0276 - val\_mean\_squared\_error: 0.0276  
 Epoch 93/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0016 -  
 mean\_squared\_error: 0.0016 - val\_loss: 0.0272 - val\_mean\_squared\_error: 0.0272  
 Epoch 94/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0016 -  
 mean\_squared\_error: 0.0016 - val\_loss: 0.0237 - val\_mean\_squared\_error: 0.0237  
 Epoch 95/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0018 -  
 mean\_squared\_error: 0.0018 - val\_loss: 0.0285 - val\_mean\_squared\_error: 0.0285  
 Epoch 96/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0016 -  
 mean\_squared\_error: 0.0016 - val\_loss: 0.0255 - val\_mean\_squared\_error: 0.0255  
 Epoch 97/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0016 -  
 mean\_squared\_error: 0.0016 - val\_loss: 0.0303 - val\_mean\_squared\_error: 0.0303  
 Epoch 98/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0019 -  
 mean\_squared\_error: 0.0019 - val\_loss: 0.0239 - val\_mean\_squared\_error: 0.0239  
 Epoch 99/100

```
215/215 [=====] - 1s 2ms/step - loss: 0.0019 -  
mean_squared_error: 0.0019 - val_loss: 0.0281 - val_mean_squared_error: 0.0281  
Epoch 100/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0015 -  
mean_squared_error: 0.0015 - val_loss: 0.0337 - val_mean_squared_error: 0.0337
```

```
[56]: plt.plot(history.history['loss'], label='loss')  
plt.plot(history.history['val_loss'], label='val_loss')  
plt.legend()  
plt.title('Using Dropout')  
plt.show()
```



### 1.0.6 Implementing Regularization Techniques

```
[57]: import tensorflow as tf  
from tensorflow.keras.layers import *  
  
class L1Model(tf.keras.models.Model):  
  
    def __init__(self):  
        super().__init__()
```

```

        self.dense_1 = Dense(128, activation=tf.nn.relu,
↪kernel_regularizer='l1')
        self.dense_2 = Dense(128, activation=tf.nn.relu,
↪kernel_regularizer='l1')
        self.dense_3 = Dense(128, activation=tf.nn.relu,
↪kernel_regularizer='l1')
        self.regression = Dense(1)

    def call(self, input_tensor):
        x = self.dense_1(input_tensor)
        x = self.dense_2(x)
        x = self.dense_3(x)
        return self.regression(x)

l1_model = L1Model()

l1_model.compile(loss=tf.keras.losses.MeanSquaredError(),
                 optimizer=tf.keras.optimizers.Adam(),
                 metrics=tf.keras.metrics.MeanSquaredError())

l1_history = l1_model.fit(X_train, y_train, epochs=100,
                         validation_data=(X_val, y_val),
                         batch_size=16)

```

Epoch 1/100

215/215 [=====] - 2s 3ms/step - loss: 7.2121 -  
mean\_squared\_error: 0.0405 - val\_loss: 0.3964 - val\_mean\_squared\_error: 0.0085

Epoch 2/100

215/215 [=====] - 0s 2ms/step - loss: 0.2543 -  
mean\_squared\_error: 0.0082 - val\_loss: 0.1849 - val\_mean\_squared\_error: 0.0072

Epoch 3/100

215/215 [=====] - 1s 2ms/step - loss: 0.1620 -  
mean\_squared\_error: 0.0055 - val\_loss: 0.1463 - val\_mean\_squared\_error: 0.0043

Epoch 4/100

215/215 [=====] - 0s 2ms/step - loss: 0.1372 -  
mean\_squared\_error: 0.0038 - val\_loss: 0.1292 - val\_mean\_squared\_error: 0.0031

Epoch 5/100

215/215 [=====] - 0s 2ms/step - loss: 0.1230 -  
mean\_squared\_error: 0.0031 - val\_loss: 0.1167 - val\_mean\_squared\_error: 0.0027

Epoch 6/100

215/215 [=====] - 0s 2ms/step - loss: 0.1113 -  
mean\_squared\_error: 0.0027 - val\_loss: 0.1056 - val\_mean\_squared\_error: 0.0030

Epoch 7/100

215/215 [=====] - 0s 2ms/step - loss: 0.1000 -  
mean\_squared\_error: 0.0022 - val\_loss: 0.0950 - val\_mean\_squared\_error: 0.0015

Epoch 8/100

215/215 [=====] - 0s 2ms/step - loss: 0.0911 -

mean\_squared\_error: 0.0013 - val\_loss: 0.0897 - val\_mean\_squared\_error: 0.0022  
Epoch 9/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0859 -  
mean\_squared\_error: 0.0011 - val\_loss: 0.0834 - val\_mean\_squared\_error:  
9.0297e-04  
Epoch 10/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0824 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0813 - val\_mean\_squared\_error:  
8.4753e-04  
Epoch 11/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0810 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0805 - val\_mean\_squared\_error: 0.0013  
Epoch 12/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0799 -  
mean\_squared\_error: 9.8843e-04 - val\_loss: 0.0793 - val\_mean\_squared\_error:  
9.2994e-04  
Epoch 13/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0788 -  
mean\_squared\_error: 9.8294e-04 - val\_loss: 0.0781 - val\_mean\_squared\_error:  
0.0011  
Epoch 14/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0775 -  
mean\_squared\_error: 9.8256e-04 - val\_loss: 0.0769 - val\_mean\_squared\_error:  
0.0011  
Epoch 15/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0759 -  
mean\_squared\_error: 9.9057e-04 - val\_loss: 0.0746 - val\_mean\_squared\_error:  
8.8222e-04  
Epoch 16/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0742 -  
mean\_squared\_error: 9.8294e-04 - val\_loss: 0.0732 - val\_mean\_squared\_error:  
7.6673e-04  
Epoch 17/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0727 -  
mean\_squared\_error: 9.5648e-04 - val\_loss: 0.0723 - val\_mean\_squared\_error:  
8.7215e-04  
Epoch 18/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0719 -  
mean\_squared\_error: 9.4531e-04 - val\_loss: 0.0714 - val\_mean\_squared\_error:  
8.4945e-04  
Epoch 19/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0709 -  
mean\_squared\_error: 9.0838e-04 - val\_loss: 0.0704 - val\_mean\_squared\_error:  
7.6477e-04  
Epoch 20/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0700 -  
mean\_squared\_error: 8.9048e-04 - val\_loss: 0.0694 - val\_mean\_squared\_error:  
7.9038e-04



Epoch 21/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0688 -  
 mean\_squared\_error: 8.9029e-04 - val\_loss: 0.0678 - val\_mean\_squared\_error:  
 7.7143e-04

Epoch 22/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0673 -  
 mean\_squared\_error: 8.3625e-04 - val\_loss: 0.0668 - val\_mean\_squared\_error:  
 8.1486e-04

Epoch 23/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0663 -  
 mean\_squared\_error: 8.0778e-04 - val\_loss: 0.0657 - val\_mean\_squared\_error:  
 7.3139e-04

Epoch 24/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0655 -  
 mean\_squared\_error: 8.2595e-04 - val\_loss: 0.0649 - val\_mean\_squared\_error:  
 7.1591e-04

Epoch 25/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0647 -  
 mean\_squared\_error: 8.0382e-04 - val\_loss: 0.0647 - val\_mean\_squared\_error:  
 7.3816e-04

Epoch 26/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0646 -  
 mean\_squared\_error: 8.3270e-04 - val\_loss: 0.0646 - val\_mean\_squared\_error:  
 7.5453e-04

Epoch 27/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0644 -  
 mean\_squared\_error: 8.2858e-04 - val\_loss: 0.0642 - val\_mean\_squared\_error:  
 8.0925e-04

Epoch 28/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0643 -  
 mean\_squared\_error: 8.4370e-04 - val\_loss: 0.0639 - val\_mean\_squared\_error:  
 7.0287e-04

Epoch 29/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0642 -  
 mean\_squared\_error: 8.3197e-04 - val\_loss: 0.0639 - val\_mean\_squared\_error:  
 7.0053e-04

Epoch 30/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0640 -  
 mean\_squared\_error: 8.2943e-04 - val\_loss: 0.0644 - val\_mean\_squared\_error:  
 8.5973e-04

Epoch 31/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0639 -  
 mean\_squared\_error: 8.4376e-04 - val\_loss: 0.0637 - val\_mean\_squared\_error:  
 7.4042e-04

Epoch 32/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0637 -  
 mean\_squared\_error: 7.9653e-04 - val\_loss: 0.0636 - val\_mean\_squared\_error:  
 7.0302e-04

Epoch 33/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0637 -  
 mean\_squared\_error: 8.6351e-04 - val\_loss: 0.0633 - val\_mean\_squared\_error:  
 6.8926e-04

Epoch 34/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0634 -  
 mean\_squared\_error: 8.0367e-04 - val\_loss: 0.0633 - val\_mean\_squared\_error:  
 7.3087e-04

Epoch 35/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0633 -  
 mean\_squared\_error: 7.9714e-04 - val\_loss: 0.0631 - val\_mean\_squared\_error:  
 8.3742e-04

Epoch 36/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0633 -  
 mean\_squared\_error: 8.1805e-04 - val\_loss: 0.0631 - val\_mean\_squared\_error:  
 6.9430e-04

Epoch 37/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0631 -  
 mean\_squared\_error: 7.7683e-04 - val\_loss: 0.0630 - val\_mean\_squared\_error:  
 7.6747e-04

Epoch 38/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0630 -  
 mean\_squared\_error: 7.8138e-04 - val\_loss: 0.0628 - val\_mean\_squared\_error:  
 7.1577e-04

Epoch 39/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0630 -  
 mean\_squared\_error: 8.5460e-04 - val\_loss: 0.0627 - val\_mean\_squared\_error:  
 7.3559e-04

Epoch 40/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0628 -  
 mean\_squared\_error: 8.0188e-04 - val\_loss: 0.0627 - val\_mean\_squared\_error:  
 7.1969e-04

Epoch 41/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0627 -  
 mean\_squared\_error: 8.1977e-04 - val\_loss: 0.0631 - val\_mean\_squared\_error:  
 9.9769e-04

Epoch 42/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0626 -  
 mean\_squared\_error: 7.8323e-04 - val\_loss: 0.0627 - val\_mean\_squared\_error:  
 6.9471e-04

Epoch 43/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0626 -  
 mean\_squared\_error: 8.5058e-04 - val\_loss: 0.0627 - val\_mean\_squared\_error:  
 7.1187e-04

Epoch 44/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0625 -  
 mean\_squared\_error: 8.4141e-04 - val\_loss: 0.0632 - val\_mean\_squared\_error:  
 0.0013

Epoch 45/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0624 -  
mean\_squared\_error: 8.4909e-04 - val\_loss: 0.0620 - val\_mean\_squared\_error:  
7.6127e-04  
Epoch 46/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0622 -  
mean\_squared\_error: 7.9771e-04 - val\_loss: 0.0619 - val\_mean\_squared\_error:  
7.4541e-04  
Epoch 47/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0623 -  
mean\_squared\_error: 8.8859e-04 - val\_loss: 0.0619 - val\_mean\_squared\_error:  
8.2490e-04  
Epoch 48/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0621 -  
mean\_squared\_error: 8.1442e-04 - val\_loss: 0.0620 - val\_mean\_squared\_error:  
6.9062e-04  
Epoch 49/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0620 -  
mean\_squared\_error: 7.9881e-04 - val\_loss: 0.0621 - val\_mean\_squared\_error:  
8.5215e-04  
Epoch 50/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0621 -  
mean\_squared\_error: 9.2395e-04 - val\_loss: 0.0623 - val\_mean\_squared\_error:  
0.0011  
Epoch 51/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0618 -  
mean\_squared\_error: 8.0311e-04 - val\_loss: 0.0618 - val\_mean\_squared\_error:  
8.5392e-04  
Epoch 52/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0618 -  
mean\_squared\_error: 8.5788e-04 - val\_loss: 0.0617 - val\_mean\_squared\_error:  
7.3052e-04  
Epoch 53/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0618 -  
mean\_squared\_error: 8.5138e-04 - val\_loss: 0.0618 - val\_mean\_squared\_error:  
0.0011  
Epoch 54/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0617 -  
mean\_squared\_error: 8.3620e-04 - val\_loss: 0.0618 - val\_mean\_squared\_error:  
7.3590e-04  
Epoch 55/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0616 -  
mean\_squared\_error: 8.6990e-04 - val\_loss: 0.0618 - val\_mean\_squared\_error:  
7.0386e-04  
Epoch 56/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0615 -  
mean\_squared\_error: 8.0939e-04 - val\_loss: 0.0614 - val\_mean\_squared\_error:  
7.4789e-04

Epoch 57/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0614 -  
 mean\_squared\_error: 8.2566e-04 - val\_loss: 0.0613 - val\_mean\_squared\_error:  
 7.6602e-04

Epoch 58/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0614 -  
 mean\_squared\_error: 8.6670e-04 - val\_loss: 0.0612 - val\_mean\_squared\_error:  
 9.5267e-04

Epoch 59/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0613 -  
 mean\_squared\_error: 8.5332e-04 - val\_loss: 0.0613 - val\_mean\_squared\_error:  
 9.3917e-04

Epoch 60/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0614 -  
 mean\_squared\_error: 9.5601e-04 - val\_loss: 0.0613 - val\_mean\_squared\_error:  
 8.2042e-04

Epoch 61/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0612 -  
 mean\_squared\_error: 8.4711e-04 - val\_loss: 0.0631 - val\_mean\_squared\_error:  
 0.0025

Epoch 62/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0612 -  
 mean\_squared\_error: 9.0463e-04 - val\_loss: 0.0612 - val\_mean\_squared\_error:  
 8.6022e-04

Epoch 63/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0611 -  
 mean\_squared\_error: 8.1979e-04 - val\_loss: 0.0609 - val\_mean\_squared\_error:  
 9.4111e-04

Epoch 64/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0610 -  
 mean\_squared\_error: 8.4156e-04 - val\_loss: 0.0609 - val\_mean\_squared\_error:  
 9.4281e-04

Epoch 65/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0610 -  
 mean\_squared\_error: 8.6410e-04 - val\_loss: 0.0609 - val\_mean\_squared\_error:  
 8.2586e-04

Epoch 66/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0609 -  
 mean\_squared\_error: 8.2647e-04 - val\_loss: 0.0618 - val\_mean\_squared\_error:  
 0.0013

Epoch 67/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0609 -  
 mean\_squared\_error: 9.1274e-04 - val\_loss: 0.0611 - val\_mean\_squared\_error:  
 9.1160e-04

Epoch 68/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0608 -  
 mean\_squared\_error: 8.5193e-04 - val\_loss: 0.0611 - val\_mean\_squared\_error:  
 9.7561e-04

Epoch 69/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0608 -  
 mean\_squared\_error: 8.5900e-04 - val\_loss: 0.0606 - val\_mean\_squared\_error:  
 7.4742e-04

Epoch 70/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0607 -  
 mean\_squared\_error: 8.6567e-04 - val\_loss: 0.0605 - val\_mean\_squared\_error:  
 8.6936e-04

Epoch 71/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0606 -  
 mean\_squared\_error: 8.4867e-04 - val\_loss: 0.0604 - val\_mean\_squared\_error:  
 7.9037e-04

Epoch 72/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0606 -  
 mean\_squared\_error: 8.5863e-04 - val\_loss: 0.0607 - val\_mean\_squared\_error:  
 7.6843e-04

Epoch 73/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0605 -  
 mean\_squared\_error: 8.6384e-04 - val\_loss: 0.0608 - val\_mean\_squared\_error:  
 9.9880e-04

Epoch 74/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0605 -  
 mean\_squared\_error: 8.3326e-04 - val\_loss: 0.0604 - val\_mean\_squared\_error:  
 8.3538e-04

Epoch 75/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0604 -  
 mean\_squared\_error: 8.8271e-04 - val\_loss: 0.0605 - val\_mean\_squared\_error:  
 9.7623e-04

Epoch 76/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0603 -  
 mean\_squared\_error: 8.4101e-04 - val\_loss: 0.0601 - val\_mean\_squared\_error:  
 7.5234e-04

Epoch 77/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0603 -  
 mean\_squared\_error: 8.5626e-04 - val\_loss: 0.0604 - val\_mean\_squared\_error:  
 0.0012

Epoch 78/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0603 -  
 mean\_squared\_error: 8.8459e-04 - val\_loss: 0.0603 - val\_mean\_squared\_error:  
 7.8470e-04

Epoch 79/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0602 -  
 mean\_squared\_error: 8.9102e-04 - val\_loss: 0.0606 - val\_mean\_squared\_error:  
 8.6372e-04

Epoch 80/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0602 -  
 mean\_squared\_error: 8.4709e-04 - val\_loss: 0.0602 - val\_mean\_squared\_error:  
 7.4294e-04

Epoch 81/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0601 -  
mean\_squared\_error: 8.5183e-04 - val\_loss: 0.0598 - val\_mean\_squared\_error:  
8.5463e-04  
Epoch 82/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0601 -  
mean\_squared\_error: 8.8159e-04 - val\_loss: 0.0600 - val\_mean\_squared\_error:  
9.6857e-04  
Epoch 83/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0600 -  
mean\_squared\_error: 8.2055e-04 - val\_loss: 0.0598 - val\_mean\_squared\_error:  
7.4947e-04  
Epoch 84/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0600 -  
mean\_squared\_error: 8.9362e-04 - val\_loss: 0.0598 - val\_mean\_squared\_error:  
7.3007e-04  
Epoch 85/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0599 -  
mean\_squared\_error: 8.3967e-04 - val\_loss: 0.0601 - val\_mean\_squared\_error:  
8.5161e-04  
Epoch 86/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0599 -  
mean\_squared\_error: 9.0702e-04 - val\_loss: 0.0599 - val\_mean\_squared\_error:  
8.1407e-04  
Epoch 87/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0599 -  
mean\_squared\_error: 8.6647e-04 - val\_loss: 0.0600 - val\_mean\_squared\_error:  
0.0011  
Epoch 88/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0598 -  
mean\_squared\_error: 8.7348e-04 - val\_loss: 0.0597 - val\_mean\_squared\_error:  
8.1828e-04  
Epoch 89/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0598 -  
mean\_squared\_error: 8.9150e-04 - val\_loss: 0.0601 - val\_mean\_squared\_error:  
0.0013  
Epoch 90/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0597 -  
mean\_squared\_error: 8.4389e-04 - val\_loss: 0.0601 - val\_mean\_squared\_error:  
0.0011  
Epoch 91/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0597 -  
mean\_squared\_error: 8.4919e-04 - val\_loss: 0.0601 - val\_mean\_squared\_error:  
9.6578e-04  
Epoch 92/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0597 -  
mean\_squared\_error: 9.0407e-04 - val\_loss: 0.0595 - val\_mean\_squared\_error:  
7.9900e-04

```

Epoch 93/100
215/215 [=====] - 0s 2ms/step - loss: 0.0596 -
mean_squared_error: 8.9673e-04 - val_loss: 0.0598 - val_mean_squared_error:
0.0012
Epoch 94/100
215/215 [=====] - 1s 2ms/step - loss: 0.0595 -
mean_squared_error: 8.5749e-04 - val_loss: 0.0594 - val_mean_squared_error:
7.5380e-04
Epoch 95/100
215/215 [=====] - 1s 2ms/step - loss: 0.0596 -
mean_squared_error: 9.1940e-04 - val_loss: 0.0594 - val_mean_squared_error:
0.0010
Epoch 96/100
215/215 [=====] - 1s 2ms/step - loss: 0.0595 -
mean_squared_error: 8.9953e-04 - val_loss: 0.0595 - val_mean_squared_error:
7.4615e-04
Epoch 97/100
215/215 [=====] - 1s 2ms/step - loss: 0.0594 -
mean_squared_error: 8.4257e-04 - val_loss: 0.0596 - val_mean_squared_error:
7.3995e-04
Epoch 98/100
215/215 [=====] - 0s 2ms/step - loss: 0.0594 -
mean_squared_error: 8.5987e-04 - val_loss: 0.0605 - val_mean_squared_error:
0.0018
Epoch 99/100
215/215 [=====] - 0s 2ms/step - loss: 0.0594 -
mean_squared_error: 8.8091e-04 - val_loss: 0.0592 - val_mean_squared_error:
8.2234e-04
Epoch 100/100
215/215 [=====] - 1s 2ms/step - loss: 0.0592 -
mean_squared_error: 8.2429e-04 - val_loss: 0.0593 - val_mean_squared_error:
8.8268e-04

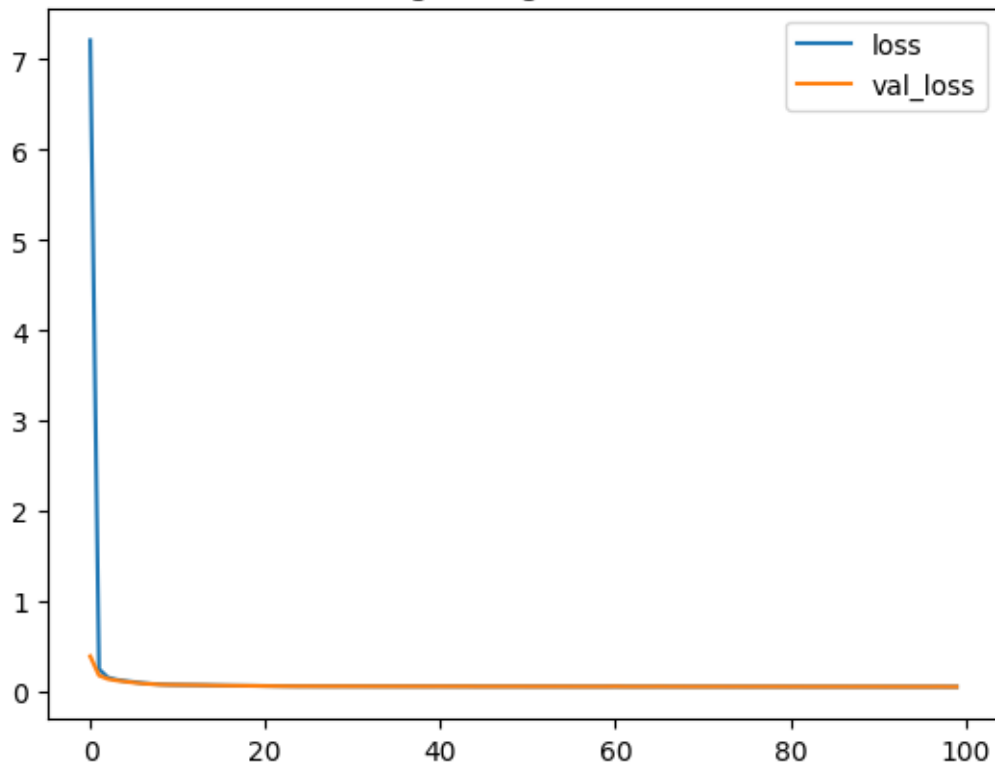
```

```

[58]: plt.plot(l1_history.history['loss'], label='loss')
plt.plot(l1_history.history['val_loss'], label='val_loss')
plt.legend()
plt.title('Using L1 regularization')
plt.show()

```

Using L1 regularization



```
[59]: import tensorflow as tf
from tensorflow.keras.layers import *

class L2Model(tf.keras.models.Model):

    def __init__(self):
        super().__init__()
        self.dense_1 = Dense(128, activation=tf.nn.relu,
↪kernel_regularizer='l2')
        self.dense_2 = Dense(128, activation=tf.nn.relu,
↪kernel_regularizer='l2')
        self.dense_3 = Dense(128, activation=tf.nn.relu,
↪kernel_regularizer='l2')
        self.regression = Dense(1)

    def call(self, input_tensor):
        x = self.dense_1(input_tensor)
        x = self.dense_2(x)
        x = self.dense_3(x)
        return self.regression(x)
```



```

l2_model = L2Model()

l2_model.compile(loss=tf.keras.losses.MeanSquaredError(),
                 optimizer=tf.keras.optimizers.Adam(),
                 metrics=tf.keras.metrics.MeanSquaredError())

l2_history = l2_model.fit(X_train, y_train, epochs=100,
                        validation_data=(X_val, y_val),
                        batch_size=16)

```

```

Epoch 1/100
215/215 [=====] - 2s 4ms/step - loss: 0.9309 -
mean_squared_error: 0.0213 - val_loss: 0.2177 - val_mean_squared_error: 0.0020
Epoch 2/100
215/215 [=====] - 1s 2ms/step - loss: 0.1151 -
mean_squared_error: 0.0022 - val_loss: 0.0603 - val_mean_squared_error: 0.0025
Epoch 3/100
215/215 [=====] - 0s 2ms/step - loss: 0.0435 -
mean_squared_error: 0.0027 - val_loss: 0.0318 - val_mean_squared_error: 0.0019
Epoch 4/100
215/215 [=====] - 0s 2ms/step - loss: 0.0274 -
mean_squared_error: 0.0024 - val_loss: 0.0236 - val_mean_squared_error: 0.0021
Epoch 5/100
215/215 [=====] - 0s 2ms/step - loss: 0.0216 -
mean_squared_error: 0.0019 - val_loss: 0.0216 - val_mean_squared_error: 0.0039
Epoch 6/100
215/215 [=====] - 1s 2ms/step - loss: 0.0196 -
mean_squared_error: 0.0027 - val_loss: 0.0178 - val_mean_squared_error: 0.0018
Epoch 7/100
215/215 [=====] - 1s 2ms/step - loss: 0.0175 -
mean_squared_error: 0.0024 - val_loss: 0.0161 - val_mean_squared_error: 0.0015
Epoch 8/100
215/215 [=====] - 0s 2ms/step - loss: 0.0168 -
mean_squared_error: 0.0029 - val_loss: 0.0172 - val_mean_squared_error: 0.0038
Epoch 9/100
215/215 [=====] - 0s 2ms/step - loss: 0.0152 -
mean_squared_error: 0.0023 - val_loss: 0.0138 - val_mean_squared_error: 0.0013
Epoch 10/100
215/215 [=====] - 1s 2ms/step - loss: 0.0140 -
mean_squared_error: 0.0019 - val_loss: 0.0152 - val_mean_squared_error: 0.0036
Epoch 11/100
215/215 [=====] - 0s 2ms/step - loss: 0.0134 -
mean_squared_error: 0.0020 - val_loss: 0.0123 - val_mean_squared_error: 0.0013
Epoch 12/100
215/215 [=====] - 0s 2ms/step - loss: 0.0126 -
mean_squared_error: 0.0019 - val_loss: 0.0130 - val_mean_squared_error: 0.0028
Epoch 13/100

```

215/215 [=====] - 0s 2ms/step - loss: 0.0118 -  
 mean\_squared\_error: 0.0017 - val\_loss: 0.0122 - val\_mean\_squared\_error: 0.0025  
 Epoch 14/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0110 -  
 mean\_squared\_error: 0.0015 - val\_loss: 0.0151 - val\_mean\_squared\_error: 0.0055  
 Epoch 15/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0113 -  
 mean\_squared\_error: 0.0022 - val\_loss: 0.0134 - val\_mean\_squared\_error: 0.0045  
 Epoch 16/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0103 -  
 mean\_squared\_error: 0.0015 - val\_loss: 0.0100 - val\_mean\_squared\_error: 0.0014  
 Epoch 17/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0109 -  
 mean\_squared\_error: 0.0024 - val\_loss: 0.0096 - val\_mean\_squared\_error: 0.0012  
 Epoch 18/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0099 -  
 mean\_squared\_error: 0.0015 - val\_loss: 0.0101 - val\_mean\_squared\_error: 0.0019  
 Epoch 19/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0098 -  
 mean\_squared\_error: 0.0017 - val\_loss: 0.0090 - val\_mean\_squared\_error:  
 9.6336e-04  
 Epoch 20/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0102 -  
 mean\_squared\_error: 0.0021 - val\_loss: 0.0108 - val\_mean\_squared\_error: 0.0030  
 Epoch 21/100  
 215/215 [=====] - 1s 4ms/step - loss: 0.0098 -  
 mean\_squared\_error: 0.0019 - val\_loss: 0.0093 - val\_mean\_squared\_error: 0.0014  
 Epoch 22/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0095 -  
 mean\_squared\_error: 0.0017 - val\_loss: 0.0096 - val\_mean\_squared\_error: 0.0016  
 Epoch 23/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0092 -  
 mean\_squared\_error: 0.0014 - val\_loss: 0.0092 - val\_mean\_squared\_error: 0.0015  
 Epoch 24/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0097 -  
 mean\_squared\_error: 0.0020 - val\_loss: 0.0098 - val\_mean\_squared\_error: 0.0020  
 Epoch 25/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0094 -  
 mean\_squared\_error: 0.0017 - val\_loss: 0.0086 - val\_mean\_squared\_error: 0.0011  
 Epoch 26/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0092 -  
 mean\_squared\_error: 0.0016 - val\_loss: 0.0090 - val\_mean\_squared\_error: 0.0016  
 Epoch 27/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0092 -  
 mean\_squared\_error: 0.0017 - val\_loss: 0.0089 - val\_mean\_squared\_error: 0.0013  
 Epoch 28/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0090 -  
 mean\_squared\_error: 0.0015 - val\_loss: 0.0088 - val\_mean\_squared\_error: 0.0013

Epoch 29/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0088 -  
mean\_squared\_error: 0.0014 - val\_loss: 0.0095 - val\_mean\_squared\_error: 0.0022  
Epoch 30/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0089 -  
mean\_squared\_error: 0.0016 - val\_loss: 0.0086 - val\_mean\_squared\_error: 0.0013  
Epoch 31/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0085 -  
mean\_squared\_error: 0.0013 - val\_loss: 0.0084 - val\_mean\_squared\_error: 0.0012  
Epoch 32/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0088 -  
mean\_squared\_error: 0.0016 - val\_loss: 0.0082 - val\_mean\_squared\_error:  
9.9968e-04  
Epoch 33/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0087 -  
mean\_squared\_error: 0.0015 - val\_loss: 0.0083 - val\_mean\_squared\_error: 0.0012  
Epoch 34/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0086 -  
mean\_squared\_error: 0.0015 - val\_loss: 0.0080 - val\_mean\_squared\_error:  
9.1257e-04  
Epoch 35/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0086 -  
mean\_squared\_error: 0.0015 - val\_loss: 0.0080 - val\_mean\_squared\_error:  
9.6763e-04  
Epoch 36/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0086 -  
mean\_squared\_error: 0.0016 - val\_loss: 0.0116 - val\_mean\_squared\_error: 0.0046  
Epoch 37/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0090 -  
mean\_squared\_error: 0.0020 - val\_loss: 0.0090 - val\_mean\_squared\_error: 0.0020  
Epoch 38/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0083 -  
mean\_squared\_error: 0.0014 - val\_loss: 0.0079 - val\_mean\_squared\_error: 0.0010  
Epoch 39/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0085 -  
mean\_squared\_error: 0.0016 - val\_loss: 0.0080 - val\_mean\_squared\_error: 0.0010  
Epoch 40/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0085 -  
mean\_squared\_error: 0.0016 - val\_loss: 0.0085 - val\_mean\_squared\_error: 0.0016  
Epoch 41/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0082 -  
mean\_squared\_error: 0.0014 - val\_loss: 0.0079 - val\_mean\_squared\_error: 0.0012  
Epoch 42/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0084 -  
mean\_squared\_error: 0.0016 - val\_loss: 0.0089 - val\_mean\_squared\_error: 0.0022  
Epoch 43/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0084 -  
mean\_squared\_error: 0.0017 - val\_loss: 0.0089 - val\_mean\_squared\_error: 0.0022

Epoch 44/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0081 -  
mean\_squared\_error: 0.0014 - val\_loss: 0.0076 - val\_mean\_squared\_error:  
9.3285e-04

Epoch 45/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0082 -  
mean\_squared\_error: 0.0016 - val\_loss: 0.0076 - val\_mean\_squared\_error:  
9.4305e-04

Epoch 46/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0084 -  
mean\_squared\_error: 0.0017 - val\_loss: 0.0083 - val\_mean\_squared\_error: 0.0016

Epoch 47/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0079 -  
mean\_squared\_error: 0.0013 - val\_loss: 0.0077 - val\_mean\_squared\_error: 0.0011

Epoch 48/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0079 -  
mean\_squared\_error: 0.0014 - val\_loss: 0.0076 - val\_mean\_squared\_error: 0.0011

Epoch 49/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0079 -  
mean\_squared\_error: 0.0014 - val\_loss: 0.0079 - val\_mean\_squared\_error: 0.0013

Epoch 50/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0079 -  
mean\_squared\_error: 0.0015 - val\_loss: 0.0075 - val\_mean\_squared\_error: 0.0011

Epoch 51/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0085 -  
mean\_squared\_error: 0.0020 - val\_loss: 0.0076 - val\_mean\_squared\_error: 0.0012

Epoch 52/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0079 -  
mean\_squared\_error: 0.0014 - val\_loss: 0.0089 - val\_mean\_squared\_error: 0.0025

Epoch 53/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0081 -  
mean\_squared\_error: 0.0017 - val\_loss: 0.0072 - val\_mean\_squared\_error:  
8.4011e-04

Epoch 54/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0077 -  
mean\_squared\_error: 0.0013 - val\_loss: 0.0072 - val\_mean\_squared\_error:  
8.2325e-04

Epoch 55/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0080 -  
mean\_squared\_error: 0.0017 - val\_loss: 0.0090 - val\_mean\_squared\_error: 0.0026

Epoch 56/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0077 -  
mean\_squared\_error: 0.0014 - val\_loss: 0.0073 - val\_mean\_squared\_error: 0.0010

Epoch 57/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0080 -  
mean\_squared\_error: 0.0017 - val\_loss: 0.0080 - val\_mean\_squared\_error: 0.0018

Epoch 58/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0076 -

```

mean_squared_error: 0.0014 - val_loss: 0.0080 - val_mean_squared_error: 0.0018
Epoch 59/100
215/215 [=====] - 1s 2ms/step - loss: 0.0079 -
mean_squared_error: 0.0017 - val_loss: 0.0075 - val_mean_squared_error: 0.0012
Epoch 60/100
215/215 [=====] - 0s 2ms/step - loss: 0.0076 -
mean_squared_error: 0.0014 - val_loss: 0.0073 - val_mean_squared_error: 0.0011
Epoch 61/100
215/215 [=====] - 1s 2ms/step - loss: 0.0077 -
mean_squared_error: 0.0016 - val_loss: 0.0073 - val_mean_squared_error: 0.0012
Epoch 62/100
215/215 [=====] - 0s 2ms/step - loss: 0.0077 -
mean_squared_error: 0.0015 - val_loss: 0.0071 - val_mean_squared_error:
9.1998e-04
Epoch 63/100
215/215 [=====] - 0s 2ms/step - loss: 0.0075 -
mean_squared_error: 0.0014 - val_loss: 0.0071 - val_mean_squared_error:
9.7355e-04
Epoch 64/100
215/215 [=====] - 0s 2ms/step - loss: 0.0076 -
mean_squared_error: 0.0015 - val_loss: 0.0075 - val_mean_squared_error: 0.0014
Epoch 65/100
215/215 [=====] - 0s 2ms/step - loss: 0.0079 -
mean_squared_error: 0.0019 - val_loss: 0.0081 - val_mean_squared_error: 0.0022
Epoch 66/100
215/215 [=====] - 0s 2ms/step - loss: 0.0075 -
mean_squared_error: 0.0014 - val_loss: 0.0080 - val_mean_squared_error: 0.0021
Epoch 67/100
215/215 [=====] - 1s 2ms/step - loss: 0.0073 -
mean_squared_error: 0.0013 - val_loss: 0.0072 - val_mean_squared_error: 0.0012
Epoch 68/100
215/215 [=====] - 0s 2ms/step - loss: 0.0077 -
mean_squared_error: 0.0017 - val_loss: 0.0078 - val_mean_squared_error: 0.0019
Epoch 69/100
215/215 [=====] - 0s 2ms/step - loss: 0.0073 -
mean_squared_error: 0.0013 - val_loss: 0.0071 - val_mean_squared_error: 0.0012
Epoch 70/100
215/215 [=====] - 0s 2ms/step - loss: 0.0076 -
mean_squared_error: 0.0017 - val_loss: 0.0069 - val_mean_squared_error: 0.0010
Epoch 71/100
215/215 [=====] - 0s 2ms/step - loss: 0.0074 -
mean_squared_error: 0.0014 - val_loss: 0.0068 - val_mean_squared_error:
8.5620e-04
Epoch 72/100
215/215 [=====] - 1s 2ms/step - loss: 0.0079 -
mean_squared_error: 0.0020 - val_loss: 0.0069 - val_mean_squared_error: 0.0010
Epoch 73/100
215/215 [=====] - 0s 2ms/step - loss: 0.0072 -

```

```

mean_squared_error: 0.0013 - val_loss: 0.0068 - val_mean_squared_error:
9.7527e-04
Epoch 74/100
215/215 [=====] - 1s 2ms/step - loss: 0.0074 -
mean_squared_error: 0.0015 - val_loss: 0.0081 - val_mean_squared_error: 0.0022
Epoch 75/100
215/215 [=====] - 1s 2ms/step - loss: 0.0076 -
mean_squared_error: 0.0017 - val_loss: 0.0068 - val_mean_squared_error:
9.3896e-04
Epoch 76/100
215/215 [=====] - 0s 2ms/step - loss: 0.0074 -
mean_squared_error: 0.0015 - val_loss: 0.0080 - val_mean_squared_error: 0.0021
Epoch 77/100
215/215 [=====] - 1s 2ms/step - loss: 0.0074 -
mean_squared_error: 0.0016 - val_loss: 0.0068 - val_mean_squared_error:
9.9328e-04
Epoch 78/100
215/215 [=====] - 1s 2ms/step - loss: 0.0071 -
mean_squared_error: 0.0013 - val_loss: 0.0076 - val_mean_squared_error: 0.0017
Epoch 79/100
215/215 [=====] - 0s 2ms/step - loss: 0.0074 -
mean_squared_error: 0.0016 - val_loss: 0.0069 - val_mean_squared_error: 0.0012
Epoch 80/100
215/215 [=====] - 0s 2ms/step - loss: 0.0071 -
mean_squared_error: 0.0014 - val_loss: 0.0071 - val_mean_squared_error: 0.0013
Epoch 81/100
215/215 [=====] - 0s 2ms/step - loss: 0.0070 -
mean_squared_error: 0.0013 - val_loss: 0.0066 - val_mean_squared_error:
9.5890e-04
Epoch 82/100
215/215 [=====] - 0s 2ms/step - loss: 0.0073 -
mean_squared_error: 0.0016 - val_loss: 0.0066 - val_mean_squared_error:
9.0516e-04
Epoch 83/100
215/215 [=====] - 0s 2ms/step - loss: 0.0073 -
mean_squared_error: 0.0016 - val_loss: 0.0067 - val_mean_squared_error: 0.0010
Epoch 84/100
215/215 [=====] - 1s 2ms/step - loss: 0.0071 -
mean_squared_error: 0.0014 - val_loss: 0.0076 - val_mean_squared_error: 0.0019
Epoch 85/100
215/215 [=====] - 0s 2ms/step - loss: 0.0073 -
mean_squared_error: 0.0017 - val_loss: 0.0065 - val_mean_squared_error:
8.5429e-04
Epoch 86/100
215/215 [=====] - 0s 2ms/step - loss: 0.0070 -
mean_squared_error: 0.0013 - val_loss: 0.0066 - val_mean_squared_error:
9.4299e-04
Epoch 87/100

```

```

215/215 [=====] - 0s 2ms/step - loss: 0.0072 -
mean_squared_error: 0.0016 - val_loss: 0.0065 - val_mean_squared_error:
8.7774e-04
Epoch 88/100
215/215 [=====] - 0s 2ms/step - loss: 0.0075 -
mean_squared_error: 0.0019 - val_loss: 0.0065 - val_mean_squared_error:
8.6753e-04
Epoch 89/100
215/215 [=====] - 0s 2ms/step - loss: 0.0069 -
mean_squared_error: 0.0013 - val_loss: 0.0071 - val_mean_squared_error: 0.0016
Epoch 90/100
215/215 [=====] - 1s 2ms/step - loss: 0.0073 -
mean_squared_error: 0.0017 - val_loss: 0.0066 - val_mean_squared_error: 0.0010
Epoch 91/100
215/215 [=====] - 0s 2ms/step - loss: 0.0069 -
mean_squared_error: 0.0014 - val_loss: 0.0067 - val_mean_squared_error: 0.0012
Epoch 92/100
215/215 [=====] - 0s 2ms/step - loss: 0.0068 -
mean_squared_error: 0.0013 - val_loss: 0.0073 - val_mean_squared_error: 0.0018
Epoch 93/100
215/215 [=====] - 0s 2ms/step - loss: 0.0073 -
mean_squared_error: 0.0018 - val_loss: 0.0085 - val_mean_squared_error: 0.0029
Epoch 94/100
215/215 [=====] - 0s 2ms/step - loss: 0.0073 -
mean_squared_error: 0.0018 - val_loss: 0.0066 - val_mean_squared_error: 0.0012
Epoch 95/100
215/215 [=====] - 0s 2ms/step - loss: 0.0069 -
mean_squared_error: 0.0014 - val_loss: 0.0069 - val_mean_squared_error: 0.0015
Epoch 96/100
215/215 [=====] - 0s 2ms/step - loss: 0.0070 -
mean_squared_error: 0.0015 - val_loss: 0.0064 - val_mean_squared_error:
8.6344e-04
Epoch 97/100
215/215 [=====] - 0s 2ms/step - loss: 0.0069 -
mean_squared_error: 0.0014 - val_loss: 0.0074 - val_mean_squared_error: 0.0021
Epoch 98/100
215/215 [=====] - 0s 2ms/step - loss: 0.0070 -
mean_squared_error: 0.0016 - val_loss: 0.0080 - val_mean_squared_error: 0.0024
Epoch 99/100
215/215 [=====] - 0s 2ms/step - loss: 0.0068 -
mean_squared_error: 0.0014 - val_loss: 0.0070 - val_mean_squared_error: 0.0016
Epoch 100/100
215/215 [=====] - 0s 2ms/step - loss: 0.0072 -
mean_squared_error: 0.0018 - val_loss: 0.0069 - val_mean_squared_error: 0.0015

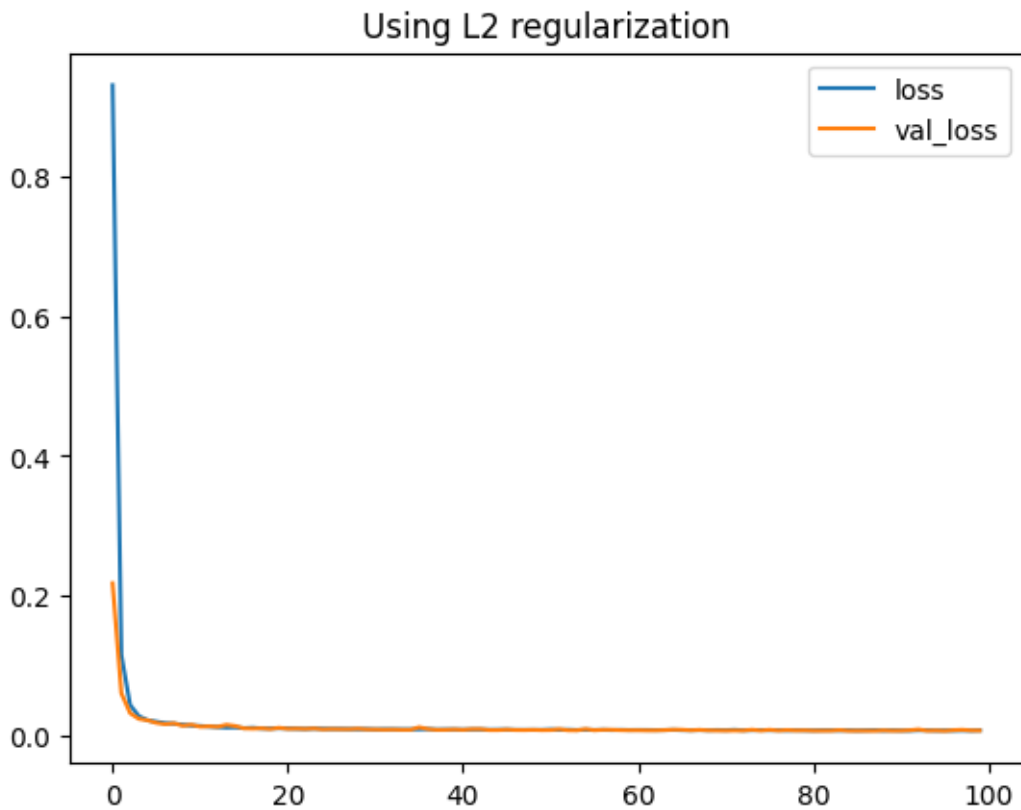
```

```

[60]: plt.plot(l2_history.history['loss'], label='loss')
      plt.plot(l2_history.history['val_loss'], label='val_loss')

```

```
plt.legend()
plt.title('Using L2 regularization')
plt.show()
```



```
[62]: import tensorflow as tf
from tensorflow.keras.layers import *

class ElasticModel(tf.keras.models.Model):

    def __init__(self):
        super().__init__()
        self.dense_1 = Dense(128, activation=tf.nn.relu,
↪kernel_regularizer='l1_l2')
        self.dense_2 = Dense(128, activation=tf.nn.relu,
↪kernel_regularizer='l1_l2')
        self.dense_3 = Dense(128, activation=tf.nn.relu,
↪kernel_regularizer='l1_l2')
        self.regression = Dense(1)

    def call(self, input_tensor):
```



```

        x = self.dense_1(input_tensor)
        x = self.dense_2(x)
        x = self.dense_3(x)
        return self.regression(x)

elastic_model = ElasticModel()

elastic_model.compile(loss=tf.keras.losses.MeanSquaredError(),
                      optimizer=tf.keras.optimizers.Adam(),
                      metrics=tf.keras.metrics.MeanSquaredError())

elastic_history = elastic_model.fit(X_train, y_train, epochs=100,
                                   validation_data=(X_val, y_val),
                                   batch_size=16)

```

```

Epoch 1/100
215/215 [=====] - 2s 3ms/step - loss: 7.6227 -
mean_squared_error: 0.0392 - val_loss: 0.3149 - val_mean_squared_error: 0.0098
Epoch 2/100
215/215 [=====] - 0s 2ms/step - loss: 0.2195 -
mean_squared_error: 0.0088 - val_loss: 0.1720 - val_mean_squared_error: 0.0069
Epoch 3/100
215/215 [=====] - 0s 2ms/step - loss: 0.1479 -
mean_squared_error: 0.0039 - val_loss: 0.1317 - val_mean_squared_error: 0.0024
Epoch 4/100
215/215 [=====] - 0s 2ms/step - loss: 0.1243 -
mean_squared_error: 0.0021 - val_loss: 0.1174 - val_mean_squared_error: 0.0014
Epoch 5/100
215/215 [=====] - 0s 2ms/step - loss: 0.1128 -
mean_squared_error: 0.0014 - val_loss: 0.1090 - val_mean_squared_error: 0.0014
Epoch 6/100
215/215 [=====] - 0s 2ms/step - loss: 0.1068 -
mean_squared_error: 0.0013 - val_loss: 0.1044 - val_mean_squared_error: 0.0010
Epoch 7/100
215/215 [=====] - 1s 3ms/step - loss: 0.1029 -
mean_squared_error: 0.0013 - val_loss: 0.1005 - val_mean_squared_error: 0.0011
Epoch 8/100
215/215 [=====] - 0s 2ms/step - loss: 0.0988 -
mean_squared_error: 0.0012 - val_loss: 0.0969 - val_mean_squared_error:
8.5473e-04
Epoch 9/100
215/215 [=====] - 0s 2ms/step - loss: 0.0962 -
mean_squared_error: 0.0012 - val_loss: 0.0950 - val_mean_squared_error: 0.0010
Epoch 10/100
215/215 [=====] - 1s 2ms/step - loss: 0.0943 -
mean_squared_error: 0.0012 - val_loss: 0.0927 - val_mean_squared_error: 0.0011
Epoch 11/100
215/215 [=====] - 1s 2ms/step - loss: 0.0918 -

```

```

mean_squared_error: 0.0012 - val_loss: 0.0903 - val_mean_squared_error:
9.2670e-04
Epoch 12/100
215/215 [=====] - 1s 2ms/step - loss: 0.0896 -
mean_squared_error: 0.0012 - val_loss: 0.0886 - val_mean_squared_error: 0.0011
Epoch 13/100
215/215 [=====] - 0s 2ms/step - loss: 0.0880 -
mean_squared_error: 0.0012 - val_loss: 0.0868 - val_mean_squared_error: 0.0011
Epoch 14/100
215/215 [=====] - 1s 2ms/step - loss: 0.0861 -
mean_squared_error: 0.0012 - val_loss: 0.0848 - val_mean_squared_error: 0.0012
Epoch 15/100
215/215 [=====] - 1s 2ms/step - loss: 0.0840 -
mean_squared_error: 0.0012 - val_loss: 0.0838 - val_mean_squared_error: 0.0015
Epoch 16/100
215/215 [=====] - 1s 2ms/step - loss: 0.0824 -
mean_squared_error: 0.0012 - val_loss: 0.0812 - val_mean_squared_error:
9.0415e-04
Epoch 17/100
215/215 [=====] - 1s 2ms/step - loss: 0.0810 -
mean_squared_error: 0.0011 - val_loss: 0.0800 - val_mean_squared_error:
9.3448e-04
Epoch 18/100
215/215 [=====] - 1s 2ms/step - loss: 0.0801 -
mean_squared_error: 0.0012 - val_loss: 0.0793 - val_mean_squared_error:
9.7038e-04
Epoch 19/100
215/215 [=====] - 1s 2ms/step - loss: 0.0791 -
mean_squared_error: 0.0014 - val_loss: 0.0779 - val_mean_squared_error: 0.0013
Epoch 20/100
215/215 [=====] - 0s 2ms/step - loss: 0.0771 -
mean_squared_error: 0.0011 - val_loss: 0.0763 - val_mean_squared_error:
9.9006e-04
Epoch 21/100
215/215 [=====] - 0s 2ms/step - loss: 0.0757 -
mean_squared_error: 0.0011 - val_loss: 0.0751 - val_mean_squared_error: 0.0011
Epoch 22/100
215/215 [=====] - 1s 2ms/step - loss: 0.0746 -
mean_squared_error: 0.0011 - val_loss: 0.0742 - val_mean_squared_error: 0.0015
Epoch 23/100
215/215 [=====] - 1s 2ms/step - loss: 0.0736 -
mean_squared_error: 0.0010 - val_loss: 0.0730 - val_mean_squared_error:
8.9859e-04
Epoch 24/100
215/215 [=====] - 1s 2ms/step - loss: 0.0733 -
mean_squared_error: 0.0011 - val_loss: 0.0735 - val_mean_squared_error: 0.0014
Epoch 25/100
215/215 [=====] - 1s 2ms/step - loss: 0.0729 -

```

```

mean_squared_error: 0.0010 - val_loss: 0.0724 - val_mean_squared_error:
8.6496e-04
Epoch 26/100
215/215 [=====] - 0s 2ms/step - loss: 0.0726 -
mean_squared_error: 0.0010 - val_loss: 0.0723 - val_mean_squared_error:
8.5884e-04
Epoch 27/100
215/215 [=====] - 0s 2ms/step - loss: 0.0724 -
mean_squared_error: 0.0011 - val_loss: 0.0722 - val_mean_squared_error:
9.0559e-04
Epoch 28/100
215/215 [=====] - 1s 2ms/step - loss: 0.0723 -
mean_squared_error: 0.0011 - val_loss: 0.0715 - val_mean_squared_error:
8.5457e-04
Epoch 29/100
215/215 [=====] - 0s 2ms/step - loss: 0.0719 -
mean_squared_error: 0.0011 - val_loss: 0.0715 - val_mean_squared_error:
9.1511e-04
Epoch 30/100
215/215 [=====] - 0s 2ms/step - loss: 0.0716 -
mean_squared_error: 0.0010 - val_loss: 0.0726 - val_mean_squared_error: 0.0019
Epoch 31/100
215/215 [=====] - 0s 2ms/step - loss: 0.0715 -
mean_squared_error: 0.0010 - val_loss: 0.0716 - val_mean_squared_error: 0.0012
Epoch 32/100
215/215 [=====] - 0s 2ms/step - loss: 0.0714 -
mean_squared_error: 0.0011 - val_loss: 0.0710 - val_mean_squared_error:
7.9894e-04
Epoch 33/100
215/215 [=====] - 0s 2ms/step - loss: 0.0710 -
mean_squared_error: 0.0010 - val_loss: 0.0708 - val_mean_squared_error:
9.6077e-04
Epoch 34/100
215/215 [=====] - 1s 2ms/step - loss: 0.0708 -
mean_squared_error: 0.0010 - val_loss: 0.0704 - val_mean_squared_error:
8.1780e-04
Epoch 35/100
215/215 [=====] - 1s 2ms/step - loss: 0.0707 -
mean_squared_error: 0.0010 - val_loss: 0.0706 - val_mean_squared_error:
9.4292e-04
Epoch 36/100
215/215 [=====] - 1s 2ms/step - loss: 0.0705 -
mean_squared_error: 9.9827e-04 - val_loss: 0.0704 - val_mean_squared_error:
7.9835e-04
Epoch 37/100
215/215 [=====] - 0s 2ms/step - loss: 0.0705 -
mean_squared_error: 0.0011 - val_loss: 0.0698 - val_mean_squared_error:
8.0782e-04

```

Epoch 38/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0702 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0699 - val\_mean\_squared\_error:  
9.0186e-04  
Epoch 39/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0701 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0696 - val\_mean\_squared\_error:  
8.9147e-04  
Epoch 40/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0699 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0696 - val\_mean\_squared\_error:  
9.8281e-04  
Epoch 41/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0699 -  
mean\_squared\_error: 0.0012 - val\_loss: 0.0694 - val\_mean\_squared\_error:  
9.2146e-04  
Epoch 42/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0696 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0693 - val\_mean\_squared\_error:  
7.7901e-04  
Epoch 43/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0695 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0689 - val\_mean\_squared\_error:  
8.0510e-04  
Epoch 44/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0693 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0694 - val\_mean\_squared\_error: 0.0012  
Epoch 45/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0692 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0691 - val\_mean\_squared\_error: 0.0011  
Epoch 46/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0692 -  
mean\_squared\_error: 0.0011 - val\_loss: 0.0686 - val\_mean\_squared\_error:  
8.3241e-04  
Epoch 47/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0689 -  
mean\_squared\_error: 9.9855e-04 - val\_loss: 0.0686 - val\_mean\_squared\_error:  
8.4821e-04  
Epoch 48/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0688 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0686 - val\_mean\_squared\_error:  
7.9295e-04  
Epoch 49/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0688 -  
mean\_squared\_error: 0.0011 - val\_loss: 0.0686 - val\_mean\_squared\_error:  
9.1737e-04  
Epoch 50/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0686 -

```

mean_squared_error: 0.0010 - val_loss: 0.0683 - val_mean_squared_error:
8.4879e-04
Epoch 51/100
215/215 [=====] - 0s 2ms/step - loss: 0.0685 -
mean_squared_error: 0.0011 - val_loss: 0.0681 - val_mean_squared_error:
7.9372e-04
Epoch 52/100
215/215 [=====] - 0s 2ms/step - loss: 0.0684 -
mean_squared_error: 0.0011 - val_loss: 0.0686 - val_mean_squared_error: 0.0013
Epoch 53/100
215/215 [=====] - 0s 2ms/step - loss: 0.0682 -
mean_squared_error: 0.0010 - val_loss: 0.0680 - val_mean_squared_error:
8.7712e-04
Epoch 54/100
215/215 [=====] - 0s 2ms/step - loss: 0.0682 -
mean_squared_error: 0.0011 - val_loss: 0.0680 - val_mean_squared_error:
8.9879e-04
Epoch 55/100
215/215 [=====] - 0s 2ms/step - loss: 0.0681 -
mean_squared_error: 0.0011 - val_loss: 0.0676 - val_mean_squared_error:
7.7041e-04
Epoch 56/100
215/215 [=====] - 0s 2ms/step - loss: 0.0679 -
mean_squared_error: 0.0010 - val_loss: 0.0680 - val_mean_squared_error: 0.0012
Epoch 57/100
215/215 [=====] - 0s 2ms/step - loss: 0.0678 -
mean_squared_error: 9.8125e-04 - val_loss: 0.0680 - val_mean_squared_error:
0.0014
Epoch 58/100
215/215 [=====] - 0s 2ms/step - loss: 0.0678 -
mean_squared_error: 0.0011 - val_loss: 0.0675 - val_mean_squared_error: 0.0012
Epoch 59/100
215/215 [=====] - 0s 2ms/step - loss: 0.0676 -
mean_squared_error: 9.9584e-04 - val_loss: 0.0675 - val_mean_squared_error:
9.5467e-04
Epoch 60/100
215/215 [=====] - 0s 2ms/step - loss: 0.0675 -
mean_squared_error: 9.9797e-04 - val_loss: 0.0680 - val_mean_squared_error:
0.0014
Epoch 61/100
215/215 [=====] - 0s 2ms/step - loss: 0.0675 -
mean_squared_error: 0.0010 - val_loss: 0.0669 - val_mean_squared_error:
8.1982e-04
Epoch 62/100
215/215 [=====] - 0s 2ms/step - loss: 0.0673 -
mean_squared_error: 9.7734e-04 - val_loss: 0.0671 - val_mean_squared_error:
9.3491e-04
Epoch 63/100

```

215/215 [=====] - 0s 2ms/step - loss: 0.0673 -  
 mean\_squared\_error: 0.0011 - val\_loss: 0.0675 - val\_mean\_squared\_error: 0.0013  
 Epoch 64/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0671 -  
 mean\_squared\_error: 0.0010 - val\_loss: 0.0667 - val\_mean\_squared\_error:  
 7.9192e-04  
 Epoch 65/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0671 -  
 mean\_squared\_error: 0.0011 - val\_loss: 0.0667 - val\_mean\_squared\_error:  
 8.9222e-04  
 Epoch 66/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0669 -  
 mean\_squared\_error: 0.0010 - val\_loss: 0.0668 - val\_mean\_squared\_error:  
 8.9230e-04  
 Epoch 67/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0669 -  
 mean\_squared\_error: 0.0011 - val\_loss: 0.0667 - val\_mean\_squared\_error:  
 8.1660e-04  
 Epoch 68/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0668 -  
 mean\_squared\_error: 0.0010 - val\_loss: 0.0667 - val\_mean\_squared\_error:  
 8.2893e-04  
 Epoch 69/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0667 -  
 mean\_squared\_error: 9.9408e-04 - val\_loss: 0.0664 - val\_mean\_squared\_error:  
 7.7674e-04  
 Epoch 70/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0666 -  
 mean\_squared\_error: 0.0010 - val\_loss: 0.0662 - val\_mean\_squared\_error:  
 7.6538e-04  
 Epoch 71/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0666 -  
 mean\_squared\_error: 0.0010 - val\_loss: 0.0665 - val\_mean\_squared\_error: 0.0011  
 Epoch 72/100  
 215/215 [=====] - 1s 2ms/step - loss: 0.0665 -  
 mean\_squared\_error: 0.0010 - val\_loss: 0.0666 - val\_mean\_squared\_error: 0.0012  
 Epoch 73/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0664 -  
 mean\_squared\_error: 0.0010 - val\_loss: 0.0662 - val\_mean\_squared\_error:  
 8.6475e-04  
 Epoch 74/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0665 -  
 mean\_squared\_error: 0.0012 - val\_loss: 0.0659 - val\_mean\_squared\_error:  
 8.3473e-04  
 Epoch 75/100  
 215/215 [=====] - 0s 2ms/step - loss: 0.0663 -  
 mean\_squared\_error: 0.0010 - val\_loss: 0.0659 - val\_mean\_squared\_error:  
 8.0830e-04

Epoch 76/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0661 -  
mean\_squared\_error: 9.6055e-04 - val\_loss: 0.0657 - val\_mean\_squared\_error:  
7.9631e-04

Epoch 77/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0661 -  
mean\_squared\_error: 9.9801e-04 - val\_loss: 0.0660 - val\_mean\_squared\_error:  
0.0012

Epoch 78/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0660 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0657 - val\_mean\_squared\_error:  
8.5133e-04

Epoch 79/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0660 -  
mean\_squared\_error: 0.0011 - val\_loss: 0.0656 - val\_mean\_squared\_error:  
8.4341e-04

Epoch 80/100  
215/215 [=====] - 0s 2ms/step - loss: 0.0659 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0656 - val\_mean\_squared\_error:  
8.3981e-04

Epoch 81/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0659 -  
mean\_squared\_error: 0.0011 - val\_loss: 0.0660 - val\_mean\_squared\_error: 0.0012

Epoch 82/100  
215/215 [=====] - 1s 3ms/step - loss: 0.0658 -  
mean\_squared\_error: 0.0011 - val\_loss: 0.0656 - val\_mean\_squared\_error: 0.0011

Epoch 83/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0657 -  
mean\_squared\_error: 9.9021e-04 - val\_loss: 0.0655 - val\_mean\_squared\_error:  
8.7387e-04

Epoch 84/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0657 -  
mean\_squared\_error: 0.0011 - val\_loss: 0.0656 - val\_mean\_squared\_error:  
9.6323e-04

Epoch 85/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0656 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0653 - val\_mean\_squared\_error:  
8.2906e-04

Epoch 86/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0655 -  
mean\_squared\_error: 0.0010 - val\_loss: 0.0653 - val\_mean\_squared\_error:  
8.8048e-04

Epoch 87/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0656 -  
mean\_squared\_error: 0.0011 - val\_loss: 0.0650 - val\_mean\_squared\_error:  
8.2338e-04

Epoch 88/100  
215/215 [=====] - 1s 2ms/step - loss: 0.0655 -

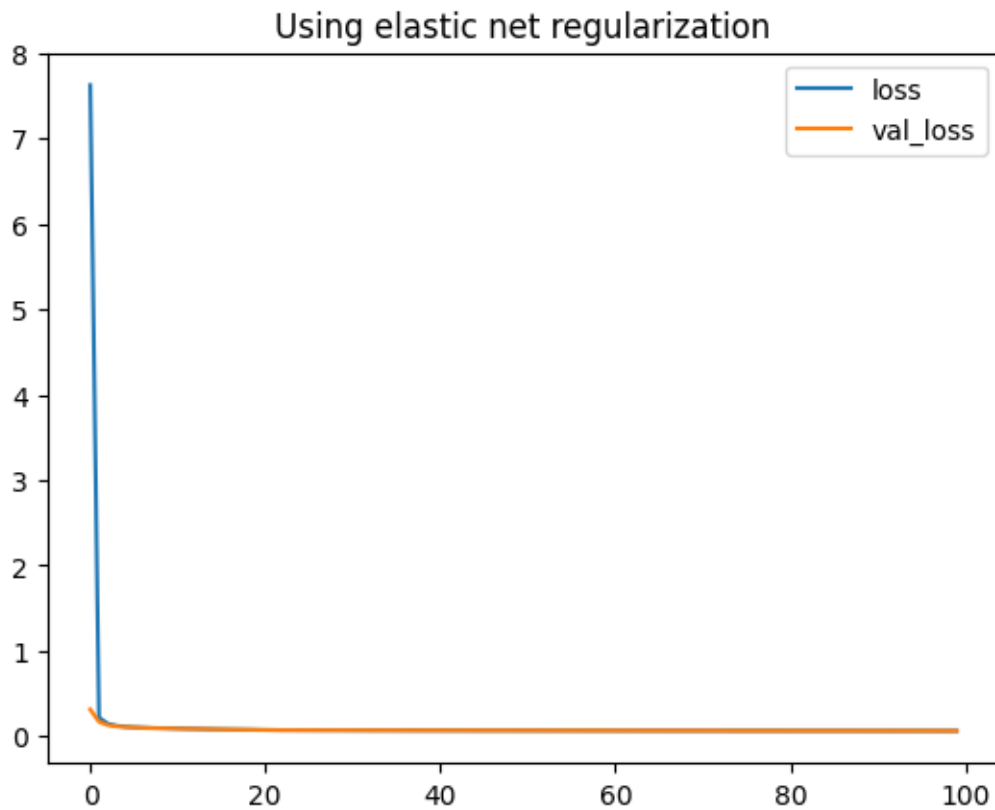
```

mean_squared_error: 0.0011 - val_loss: 0.0657 - val_mean_squared_error: 0.0015
Epoch 89/100
215/215 [=====] - 1s 2ms/step - loss: 0.0655 -
mean_squared_error: 0.0011 - val_loss: 0.0651 - val_mean_squared_error:
8.7206e-04
Epoch 90/100
215/215 [=====] - 1s 2ms/step - loss: 0.0653 -
mean_squared_error: 0.0010 - val_loss: 0.0652 - val_mean_squared_error:
9.4304e-04
Epoch 91/100
215/215 [=====] - 1s 2ms/step - loss: 0.0652 -
mean_squared_error: 0.0010 - val_loss: 0.0651 - val_mean_squared_error: 0.0011
Epoch 92/100
215/215 [=====] - 1s 2ms/step - loss: 0.0652 -
mean_squared_error: 9.9383e-04 - val_loss: 0.0656 - val_mean_squared_error:
0.0014
Epoch 93/100
215/215 [=====] - 0s 2ms/step - loss: 0.0651 -
mean_squared_error: 0.0010 - val_loss: 0.0647 - val_mean_squared_error:
8.1710e-04
Epoch 94/100
215/215 [=====] - 0s 2ms/step - loss: 0.0652 -
mean_squared_error: 0.0011 - val_loss: 0.0646 - val_mean_squared_error:
7.6389e-04
Epoch 95/100
215/215 [=====] - 0s 2ms/step - loss: 0.0650 -
mean_squared_error: 0.0010 - val_loss: 0.0649 - val_mean_squared_error:
9.3929e-04
Epoch 96/100
215/215 [=====] - 0s 2ms/step - loss: 0.0650 -
mean_squared_error: 0.0010 - val_loss: 0.0651 - val_mean_squared_error: 0.0011
Epoch 97/100
215/215 [=====] - 0s 2ms/step - loss: 0.0649 -
mean_squared_error: 9.9785e-04 - val_loss: 0.0646 - val_mean_squared_error:
8.8434e-04
Epoch 98/100
215/215 [=====] - 1s 2ms/step - loss: 0.0649 -
mean_squared_error: 0.0011 - val_loss: 0.0645 - val_mean_squared_error:
8.0498e-04
Epoch 99/100
215/215 [=====] - 0s 2ms/step - loss: 0.0648 -
mean_squared_error: 9.8291e-04 - val_loss: 0.0644 - val_mean_squared_error:
8.4214e-04
Epoch 100/100
215/215 [=====] - 0s 2ms/step - loss: 0.0648 -
mean_squared_error: 0.0010 - val_loss: 0.0642 - val_mean_squared_error:
7.6755e-04

```

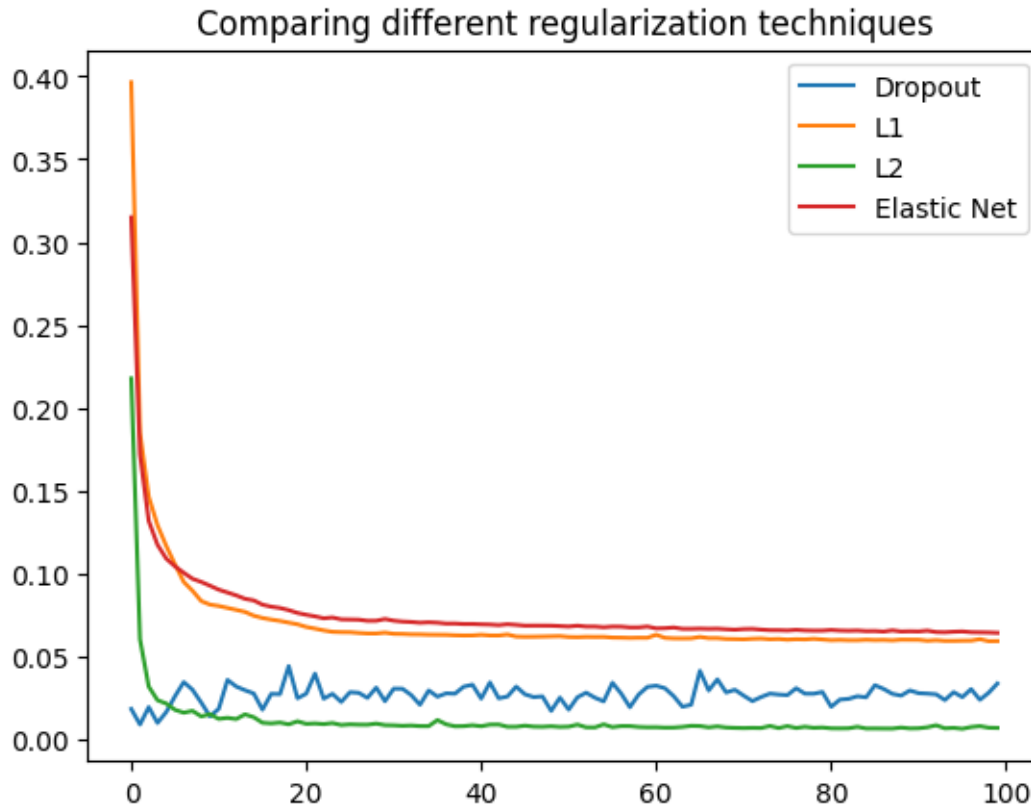


```
[63]: plt.plot(elastic_history.history['loss'], label='loss')
plt.plot(elastic_history.history['val_loss'], label='val_loss')
plt.legend()
plt.title('Using elastic net regularization')
plt.show()
```



### 1.0.7 Comparing Regularization Techniques

```
[64]: plt.plot(history.history['val_loss'], label='Dropout')
plt.plot(l1_history.history['val_loss'], label='L1')
plt.plot(l2_history.history['val_loss'], label='L2')
plt.plot(elastic_history.history['val_loss'], label='Elastic Net')
plt.legend()
plt.title('Comparing different regularization techniques')
plt.show()
```



L2 regularization appears to obtain the best results. This makes sense as with L2 regularization, the network becomes less sensitive to small changes in the input data which is usually the case with stock markets.

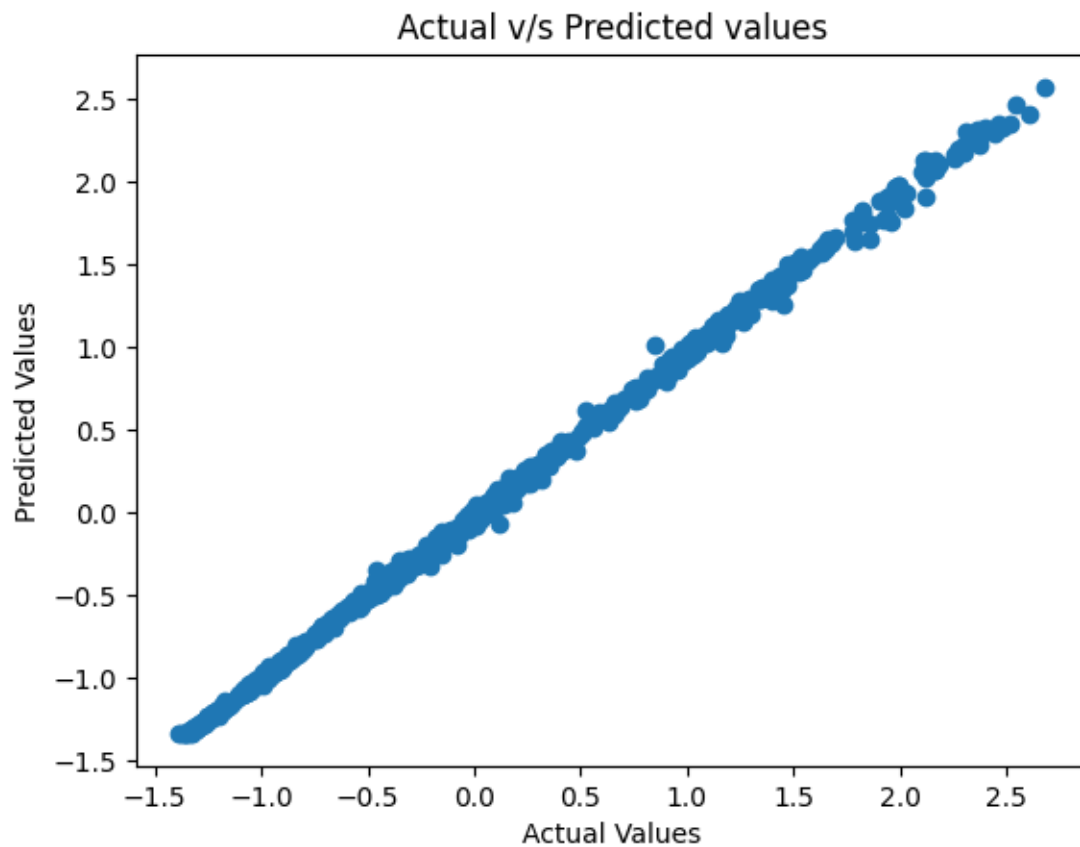
### 1.0.8 Looking at the predictions

We will use the model created with L2 regularization.

```
[74]: # Predicting the values
y_pred = l2_model.predict(X_test)

# Plotting the predictions
plt.scatter(y_test, y_pred)
plt.title("Actual v/s Predicted values")
plt.xlabel("Actual Values")
plt.ylabel("Predicted Values")
plt.show()
```

30/30 [=====] - 0s 2ms/step



Wonderful! We get a linear curve which signifies that the predictions are very close to the actual values.

**2 End**