Cognizant ml modelling

October 17, 2023

1 COGNIZANT VIRTUAL INTERSHIP-TASK 4(ML MOD-ELLING)

2 IMPORTING REQUIRED LIBRARIES

```
[52]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
pd.set_option('display.max_columns',25)
pd.set_option('display.max_rows',300)
plt.rcParams['axes.facecolor']='#99ffcc'
import warnings
warnings.filterwarnings('ignore')
```

3 IMPORTING DATASETS AND MERGING INTO SINGLE DATAFRAME

```
[54]: sales.sample(4)
```

```
[54]: Unnamed: 0 transaction_id timestamp \
5436 5436 e07caa0e-050d-4fe6-a401-b99f229aba1e 2022-03-06 18:28:52
6859 6859 6c79d5e9-28e5-4338-8d8e-3c6a5b889952 2022-03-03 10:16:15
960 960 47ae603c-4864-46e4-aa3f-a120618d2dd1 2022-03-07 15:54:39
4457 4457 96b14f29-7002-40de-814e-4bf90023e1e0 2022-03-04 15:09:18
```

product_id category customer_type \

```
5436 508c8cb7-6447-4c2f-8843-84aeac9b0aad
                                                    baked goods
                                                                         gold
      6859 d6aec25d-8af7-4661-9917-4d71d1587a6f
                                                  personal care
                                                                   non-member
      960
            14736243-d346-438f-9535-d80fcb9f3882
                                                          fruit
                                                                        basic
           951aba8f-b7fc-4fba-b188-b9978cb96304
      4457
                                                         cheese
                                                                   non-member
            unit_price quantity total payment_type
                  4.19
                               4 16.76
                                          debit card
      5436
                  5.19
                               2 10.38
      6859
                                                cash
      960
                  1.49
                               2
                                   2.98
                                         credit card
      4457
                  5.49
                               3 16.47
                                                cash
[55]: ssl.head()
      #['id', 'timestamp', 'product_id', 'estimated_stock_pct']
[55]:
         Unnamed: 0
                                                       id
                                                                     timestamp \
                  0
                    4220e505-c247-478d-9831-6b9f87a4488a
                                                           2022-03-07 12:13:02
      0
      1
                  1 f2612b26-fc82-49ea-8940-0751fdd4d9ef
                                                           2022-03-07 16:39:46
                  2 989a287f-67e6-4478-aa49-c3a35dac0e2e 2022-03-01 18:17:43
      2
      3
                     af8e5683-d247-46ac-9909-1a77bdebefb2 2022-03-02 14:29:09
                     08a32247-3f44-4002-85fb-c198434dd4bb 2022-03-02 13:46:18
                                   product_id estimated_stock_pct
       f658605e-75f3-4fed-a655-c0903f344427
                                                              0.75
      1 de06083a-f5c0-451d-b2f4-9ab88b52609d
                                                              0.48
      2 ce8f3a04-d1a4-43b1-a7c2-fa1b8e7674c8
                                                              0.58
      3 c21e3ba9-92a3-4745-92c2-6faef73223f7
                                                              0.79
      4 7f478817-aa5b-44e9-9059-8045228c9eb0
                                                              0.22
[56]: sst.head()
      #['id', 'timestamp', 'temperature']
         Unnamed: 0
[56]:
                                                       id
                                                                     timestamp
      0
                    d1ca1ef8-0eac-42fc-af80-97106efc7b13
                                                           2022-03-07 15:55:20
      1
                  1 4b8a66c4-0f3a-4f16-826f-8cf9397e9d18
                                                           2022-03-01 09:18:22
                  2 3d47a0c7-1e72-4512-812f-b6b5d8428cf3 2022-03-04 15:12:26
      2
                  3 9500357b-ce15-424a-837a-7677b386f471 2022-03-02 12:30:42
      3
      4
                  4 c4b61fec-99c2-4c6d-8e5d-4edd8c9632fa 2022-03-05 09:09:33
         temperature
      0
                2.96
      1
                1.88
      2
                1.78
      3
                2.18
      4
                1.38
[57]: m1=pd.merge(sales,ssl,on=['product_id'])
      m1.sample(5)
```

```
#['id', 'timestamp', 'product_id', 'estimated_stock_pct']
[57]:
             Unnamed: 0 x
                                                  transaction id \
      404673
                      7504 b3c5386a-e0df-4d15-98f0-909d9a12a135
      380903
                      6797
                           d998a270-e65d-4bcd-8b1a-03721a8d2ed7
      346105
                      6141
                           6a0ccd7e-dc36-4738-8235-3ed15a21af4d
      362671
                      6483
                           659a75dd-ff33-4dbc-95aa-10e572ac7fe2
      347546
                      6126
                           db4e7c9b-e4b5-4d2a-a78b-9168807a19e0
                      timestamp_x
                                                             product_id \
             2022-03-05 13:51:32 0f493675-d59d-45dd-9406-084932bf41c1
      404673
      380903 2022-03-03 16:59:38
                                  75d576b8-dab5-4663-9523-ec5f07819535
      346105 2022-03-01 10:28:52
                                   32a2ea10-8010-4046-bf34-2117692ebd4a
      362671 2022-03-07 17:16:47
                                   3ff2b6a7-3ceb-44e5-bed4-bd96bed752c4
      347546 2022-03-07 15:09:49 3fbf5c52-e83a-4476-8475-07ce81e1bdb8
                   category customer_type unit_price quantity total payment_type \
      404673
                   kitchen
                              non-member
                                                13.19
                                                              1 13.19 credit card
      380903
            personal care
                                 standard
                                                 6.99
                                                              3 20.97
                                                                         debit card
      346105
                     snacks
                                  premium
                                                 3.19
                                                              1
                                                                  3.19
                                                                               cash
      362671
             baby products
                                     gold
                                                15.19
                                                              4 60.76
                                                                               cash
      347546
                                                 3.99
                                                                  7.98
                     snacks
                                 standard
                                                                               cash
             Unnamed: 0_y
                                                              id
      404673
                      2007 32d7bb99-874b-477f-83b4-926e82175409
      380903
                           b140c344-1084-43f2-98d8-ce59bce8f9a5
                      4952
      346105
                     11977
                           f9b04faa-bb60-44f4-a56a-59c09b4d01b1
                      6444
                           234c50ef-4dba-4be5-b6f9-64f223756fd3
      362671
                      7490 dd8c53a1-2a7e-4f86-b503-8f7111f87664
      347546
                      timestamp_y
                                  estimated stock pct
      404673
             2022-03-06 19:21:15
                                                  0.13
      380903 2022-03-01 14:52:48
                                                  0.47
      346105 2022-03-03 13:21:22
                                                  0.51
      362671 2022-03-07 10:27:55
                                                  0.35
      347546 2022-03-05 11:47:33
                                                  0.52
[58]: sst.columns=['Unnamed: 0', 'id', 'timestamp_x', 'temperature']
[59]: fdf=pd.merge(m1,sst,on='timestamp x')
      fdf.head()
[59]:
        Unnamed: 0 x
                                             transaction_id
                                                                     timestamp_x \
      0
                    5 b5b3c8b9-f496-484d-aa30-4f2efb5ed56c
                                                             2022-03-07 17:59:47
      1
                    5 b5b3c8b9-f496-484d-aa30-4f2efb5ed56c
                                                             2022-03-07 17:59:47
      2
                    5 b5b3c8b9-f496-484d-aa30-4f2efb5ed56c
                                                             2022-03-07 17:59:47
      3
                      b5b3c8b9-f496-484d-aa30-4f2efb5ed56c
                                                             2022-03-07 17:59:47
```

```
product_id category customer_type
                                                                     unit_price \
     0 3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                fruit
                                                           standard
                                                                           3.99
     1 3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                fruit
                                                           standard
                                                                           3.99
     2 3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                fruit
                                                           standard
                                                                           3.99
     3 3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                fruit
                                                           standard
                                                                           3.99
     4 3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                fruit
                                                           standard
                                                                           3.99
        quantity total payment_type
                                     Unnamed: 0 y \
     0
                  15.96
                                cash
                                               164
     1
                 15.96
                                cash
                                               172
     2
               4 15.96
                                cash
                                               242
     3
               4 15.96
                                cash
                                               343
               4 15.96
                                cash
                                               465
                                        id_x
                                                      timestamp_y \
     0 5540da96-885e-4d17-aa33-5720ca9b57d5
                                              2022-03-03 09:16:02
     1 e65514e8-2935-4921-baf8-2a9cf849a7f2
                                              2022-03-04 17:52:19
     2 1faae309-bd2c-412c-b988-42e3086e2e86
                                              2022-03-07 12:44:46
     3 d383beb7-6dd0-46b3-aaa3-a1141e0c94dd
                                              2022-03-02 17:36:38
     4 9c0d4cbe-c00d-42c2-8489-263188ae0b17
                                              2022-03-07 09:44:25
        estimated_stock_pct Unnamed: 0
                                                                         id y \
     0
                       0.25
                                  15854
                                         38808b9d-712c-4703-a5ae-920817960fa6
     1
                       0.23
                                  15854 38808b9d-712c-4703-a5ae-920817960fa6
     2
                       0.80
                                  15854 38808b9d-712c-4703-a5ae-920817960fa6
     3
                       0.79
                                  15854 38808b9d-712c-4703-a5ae-920817960fa6
     4
                       0.86
                                  15854 38808b9d-712c-4703-a5ae-920817960fa6
        temperature
     0
              -2.84
     1
              -2.84
     2
              -2.84
     3
              -2.84
     4
              -2.84
[60]: dr=['transaction_id','Unnamed: 0_x', 'id_x','Unnamed: 0_y', __
       fdf.drop(columns=dr,inplace=True,axis=1)
[61]: fdf.head()
[61]:
                                                       product_id category \
                timestamp_x
     0 2022-03-07 17:59:47
                             3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                     fruit
     1 2022-03-07 17:59:47
                             3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                     fruit
     2 2022-03-07 17:59:47
                             3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                     fruit
```

5 b5b3c8b9-f496-484d-aa30-4f2efb5ed56c 2022-03-07 17:59:47

4

```
3 2022-03-07 17:59:47 3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                       fruit
      4 2022-03-07 17:59:47 3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                       fruit
        customer_type unit_price quantity total payment_type
      0
             standard
                             3.99
                                              15.96
                                                            cash
             standard
                             3.99
                                           4
                                             15.96
      1
                                                            cash
                                           4 15.96
      2
             standard
                             3.99
                                                            cash
             standard
                             3.99
                                           4 15.96
      3
                                                            cash
                             3.99
             standard
                                           4 15.96
                                                            cash
         estimated_stock_pct temperature
      0
                        0.25
                                    -2.84
                        0.23
                                    -2.84
      1
      2
                                    -2.84
                        0.80
      3
                        0.79
                                    -2.84
      4
                        0.86
                                    -2.84
[62]: fdf.columns=['timestamp', 'prd_id', 'category', 'customer_type', 'unit_price',
             'quantity', 'total', 'payment_type', 'avg_stk_prc',
             'temperature']
[63]:
     fdf.head()
[63]:
                   timestamp
                                                             prd_id category \
                                                                       fruit
      0 2022-03-07 17:59:47
                              3bc6c1ea-0198-46de-9ffd-514ae3338713
      1 2022-03-07 17:59:47
                              3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                       fruit
      2 2022-03-07 17:59:47
                              3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                       fruit
      3 2022-03-07 17:59:47
                              3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                       fruit
      4 2022-03-07 17:59:47
                              3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                       fruit
        customer_type
                       unit_price quantity total payment_type avg_stk_prc \
             standard
                             3.99
                                           4 15.96
                                                                         0.25
      0
                                                            cash
                             3.99
                                           4 15.96
                                                                         0.23
      1
             standard
                                                            cash
      2
             standard
                             3.99
                                           4 15.96
                                                            cash
                                                                         0.80
             standard
                             3.99
                                           4 15.96
                                                                         0.79
                                                            cash
             standard
                             3.99
                                           4 15.96
                                                                         0.86
                                                            cash
         temperature
      0
               -2.84
               -2.84
      1
      2
               -2.84
               -2.84
      3
               -2.84
```

4 DATA PREPROCESSING

```
[64]: fdf.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 38942 entries, 0 to 38941
     Data columns (total 10 columns):
                         Non-Null Count Dtype
          Column
          -----
      0
          timestamp
                         38942 non-null
                                        object
      1
          prd_id
                         38942 non-null object
      2
          category
                         38942 non-null
                                        object
      3
          customer_type 38942 non-null
                                         object
      4
          unit_price
                         38942 non-null float64
      5
          quantity
                         38942 non-null int64
      6
          total
                         38942 non-null float64
      7
          payment_type
                         38942 non-null
                                        object
          avg_stk_prc
                         38942 non-null
                                        float64
          temperature
                         38942 non-null
                                        float64
     dtypes: float64(4), int64(1), object(5)
     memory usage: 3.0+ MB
[65]: fdf['timestamp']=pd.to_datetime(fdf['timestamp'])
     fdf.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 38942 entries, 0 to 38941
     Data columns (total 10 columns):
      #
          Column
                         Non-Null Count Dtype
                         -----
          _____
      0
          timestamp
                         38942 non-null datetime64[ns]
      1
          prd_id
                         38942 non-null object
      2
          category
                         38942 non-null
                                        object
      3
          customer_type 38942 non-null
                                        object
      4
          unit_price
                         38942 non-null float64
      5
                         38942 non-null int64
          quantity
      6
          total
                         38942 non-null float64
      7
                         38942 non-null object
          payment_type
          avg_stk_prc
                         38942 non-null float64
          temperature
                         38942 non-null float64
     dtypes: datetime64[ns](1), float64(4), int64(1), object(4)
     memory usage: 3.0+ MB
[66]: fdf['week']=fdf['timestamp'].dt.dayofweek
     fdf['day']=fdf['timestamp'].dt.day
     fdf['hour']=fdf['timestamp'].dt.hour
[67]: sum(fdf.duplicated())
```

[67]: 9237

```
[68]: dups=fdf.duplicated()
  ndf=fdf[~dups]
  ndf
```

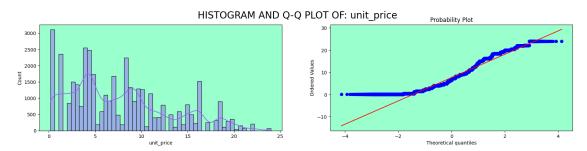
[68]:	38938 38939 38940	timestamy 2022-03-07 17:59:4' 2022-03-07 17:59:4' 2022-03-07 17:59:4' 2022-03-07 17:59:4' 2022-03-07 17:59:4' 2022-03-07 11:08:4' 2022-03-07 11:08:4' 2022-03-07 11:08:4' 2022-03-07 11:08:4' 2022-03-07 11:08:4'	7 3bc6c1ea-01 7 3bc6c1ea-01 7 3bc6c1ea-01 7 3bc6c1ea-01 7 3bc6c1ea-01 7 bc6187a9-d5 7 bc6187a9-d5 7 bc6187a9-d5 7 bc6187a9-d5	98-46de 98-46de 98-46de 98-46de 08-482b 08-482b 08-482b	-9ffd- -9ffd- -9ffd- -9ca6- -9ca6- -9ca6- -9ca6-	514ae 514ae 514ae 514ae 590d1 590d1 590d1	33387: 33387: 33387: 33387: cc7524 cc7524 cc7524	13 13 13 13 13 4f 4f 4f	
	0 1 2		customer_type standard standard standard	unit_			tity 4 4 4	total 15.96 15.96 15.96	\
	3 4 	fruit fruit 	standard standard 		3.99		4 4	15.96 15.96	
	38936 38938 38939 38940 38941	cleaning products cleaning products cleaning products cleaning products cleaning products	basic basic basic basic		14.19 14.19 14.19 14.19 14.19		2 2 2 2 2	28.38 28.38 28.38 28.38 28.38	
	0 1 2 3 4	payment_type avg_s cash cash cash cash cash cash	0.25 0.23 0.80 0.79 0.86	rature -2.84 -2.84 -2.84 -2.84 -2.84	week 0 0 0 0	day 7 7 7 7 7	hour 17 17 17 17		
	38936 38938 38939 38940 38941	debit card debit card debit card debit card debit card debit card	 0.68 0.67 0.92 0.21 0.14	-1.33 -1.33 -1.33 -1.33 -1.33	 0 0 0 0	7 7 7 7 7	11 11 11 11		

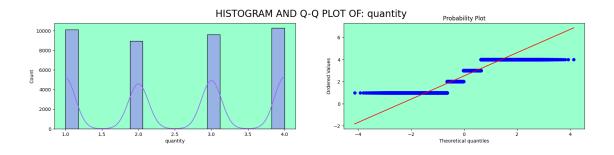
[29705 rows x 13 columns]

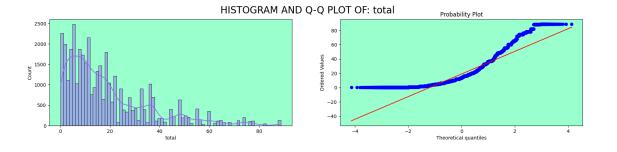
[71]: for i in ifc: histandbox(i,fdf)

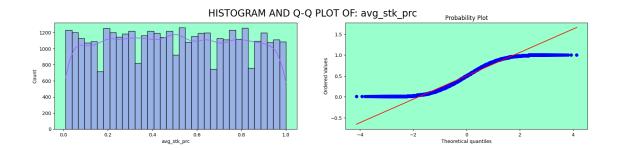
plt.show()

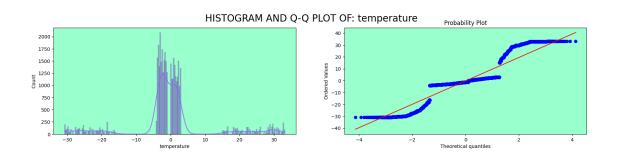
fig.suptitle(f"HISTOGRAM AND Q-Q PLOT OF: {a}", fontsize=20)

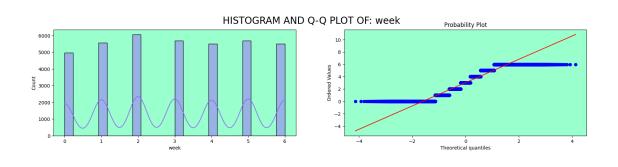


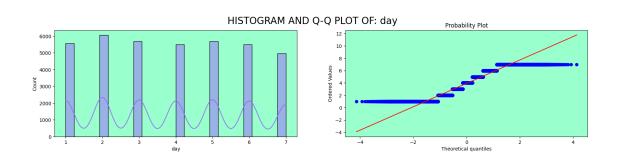


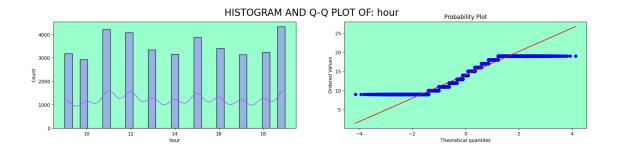












5 OUTLIER DETECTION AND REMOVAL OF INTER-QUARTILE RANGE METHOD

coefficient calculation towards sigma value

```
• = Q1 - 1.722 * IQR
```

- $\bullet = Q1 1.722 * (Q3 Q1)$
- $\bullet = -0.675 1.722 * (0.675 [-0.675])$
- $\bullet = -0.675 1.722*1.35$

29.73

29.73

1

1

1

1

14

14

 $\bullet = -2.99$

[72]: #outliers

69

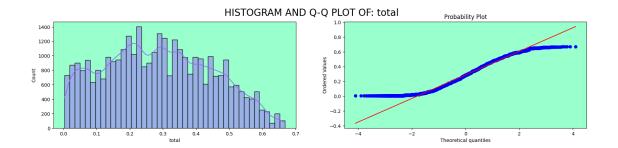
70

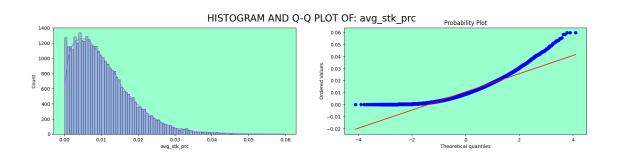
• ~=3sigma

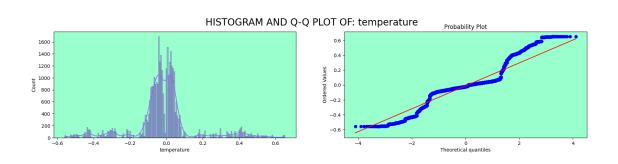
```
c=fdf.select_dtypes(include=['int','float']).columns
      q1=fdf[c].quantile(.25)
      q3=fdf[c].quantile(.75)
      iqr=q3-q1
      ll=q1-1.722*iqr
      ul=q3+1.722*iqr
      mask=fdf[~(fdf[c]>ul)|(fdf[c]<ll)].all(axis=1)
      cdf=fdf[mask]
      cdf.head(3)
[72]:
                   timestamp
                                                            prd_id category \
      69 2022-03-01 14:58:07 3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                      fruit
      70 2022-03-01 14:58:07 3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                      fruit
      71 2022-03-01 14:58:07 3bc6c1ea-0198-46de-9ffd-514ae3338713
                                                                      fruit
         customer_type
                        unit_price quantity total payment_type avg_stk_prc \
               premium
                                           4 15.96
                                                        e-wallet
                                                                         0.25
      69
                              3.99
      70
                              3.99
                                           4 15.96
                                                        e-wallet
                                                                         0.23
               premium
                                           4 15.96
      71
               premium
                              3.99
                                                        e-wallet
                                                                         0.80
          temperature week
                             day hour
```

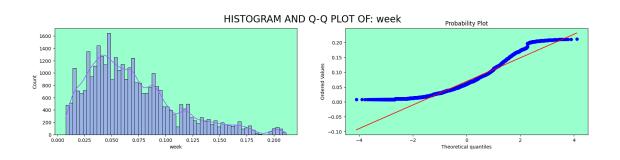
```
71
                 29.73
                           1
                                 1
                                      14
[73]: cdf.drop(columns=['timestamp', 'prd_id'],axis=1,inplace=True)
[74]: cdf.select_dtypes(include=['int','float']).columns
[74]: Index(['unit_price', 'quantity', 'total', 'avg_stk_prc', 'temperature', 'week',
              'day', 'hour'],
             dtype='object')
[75]: from sklearn.preprocessing import PowerTransformer,QuantileTransformer
      f=['unit_price', 'quantity', 'total', 'avg_stk_prc', 'temperature', 'week',
              'day', 'hour']
         DISTRIBUTION – FEATURE TRANSOFORMATION
[76]: from sklearn.preprocessing import MaxAbsScaler,Normalizer
[77]: n=Normalizer(norm='11',)
      cdf[f]=n.fit_transform(cdf[f])
      for i in f:
          histandbox(i,cdf)
                                    HISTOGRAM AND Q-Q PLOT OF: unit price
                                                                     Probability Plot
           1400
           1200
           1000
                                                     0.2
                                                     0.1
                                                                    0
Theoretical quantiles
                                    HISTOGRAM AND Q-Q PLOT OF: quantity
                                                                     Probability Plot
                                                    0.10
                                                    0.05
                                                     0.00
```

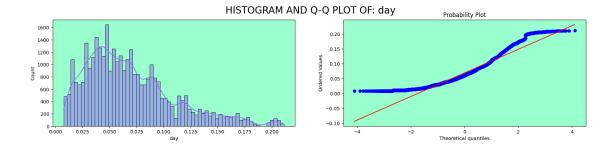
0 Theoretical quantiles

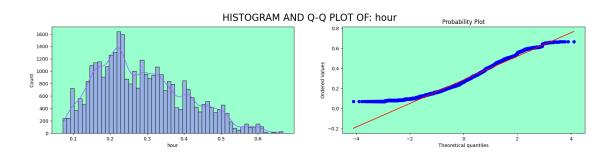












```
[78]: cat=cdf.select_dtypes(include='object').columns
cat
```

[78]: Index(['category', 'customer_type', 'payment_type'], dtype='object')

category 22

```
[79]: for i in cat:
    print("******************")
    print(f'Nuniques of : {i}',cdf[i].nunique())
    print()
    print(cdf[i].value_counts().sort_values(ascending=False))
```

Nuniques of :

-	
category	
fruit	6081
vegetables	4287
dairy	2503
meat	2066
canned foods	2016
refrigerated items	1873
packaged foods	1480
kitchen	1372
seafood	1305
cheese	1195

```
baked goods
                        1165
baby products
                        1156
medicine
                        1132
pets
                        1001
beverages
                         915
cleaning products
                         790
personal care
                         772
baking
                         748
snacks
                         710
frozen
                         677
condiments and sauces
                         388
spices and herbs
                         292
Name: count, dtype: int64
********
Nuniques of :
                     customer_type 5
customer_type
non-member
             7801
premium
             7273
gold
             7046
basic
             6000
standard
             5804
Name: count, dtype: int64
********
Nuniques of :
                     payment_type 4
payment_type
              9112
cash
              8787
credit card
e-wallet
              8640
              7385
debit card
Name: count, dtype: int64
```

7 CATEGORICAL FEATURES ENCODING: LABELEN-CODER & ONEHOTENCODER

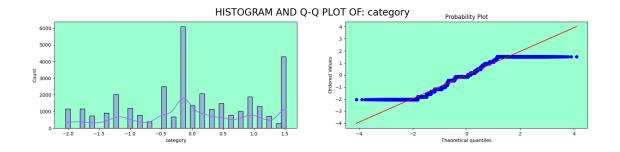
```
[80]: from sklearn.preprocessing import LabelEncoder,OneHotEncoder
le=LabelEncoder()
cdf['category']=le.fit_transform(cdf['category'])

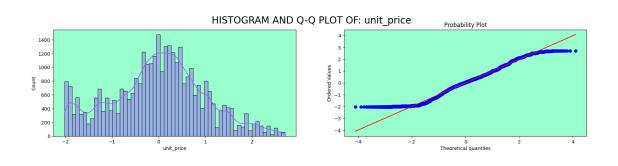
[81]: for i in ['customer_type','payment_type']:
    cdf[i]=cdf[i].astype('category')

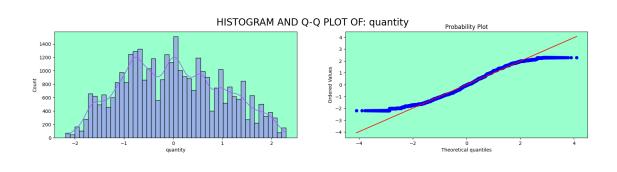
[82]: cdf=pd.get_dummies(cdf,columns=['customer_type','payment_type'])
cdf
```

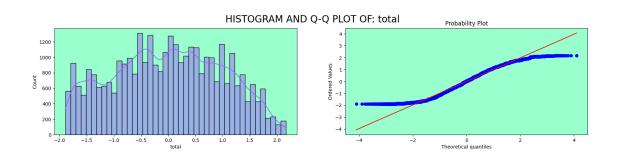
```
[82]:
                        unit_price quantity
                                                   total avg_stk_prc
             category
                                                                       temperature
                          0.057057
      69
                    10
                                     0.057200
                                                0.228228
                                                              0.003575
                                                                            0.425139
      70
                    10
                          0.057073
                                     0.057216
                                                0.228294
                                                              0.003290
                                                                            0.425261
      71
                    10
                          0.056612
                                     0.056754
                                                0.226447
                                                              0.011351
                                                                            0.421822
      72
                    10
                          0.056620
                                     0.056762
                                                0.226479
                                                              0.011210
                                                                            0.421882
      73
                    10
                          0.056564 0.056705
                                                0.226255
                                                                            0.421463
                                                              0.012192
      38891
                     6
                          0.224350 0.026410
                                                0.448699
                                                              0.004490
                                                                           -0.031956
                                                                           -0.031960
      38892
                     6
                          0.224379
                                    0.026413
                                                0.448759
                                                              0.004358
      38893
                     6
                          0.224320
                                     0.026406
                                                0.448640
                                                              0.004621
                                                                           -0.031951
                     6
                                     0.026438
                                                0.449174
      38894
                          0.224587
                                                              0.003437
                                                                           -0.031989
                     6
                          0.222586 0.026202
                                                0.445172
                                                              0.012315
                                                                           -0.031704
      38895
                  week
                              day
                                       hour
                                              customer_type_basic
                                                                    customer_type_gold
      69
             0.014300
                        0.014300
                                   0.200200
                                                             False
                                                                                  False
      70
             0.014304
                        0.014304
                                   0.200257
                                                             False
                                                                                  False
      71
             0.014188
                        0.014188
                                   0.198638
                                                             False
                                                                                  False
             0.014190
                                                                                  False
      72
                        0.014190
                                                             False
                                   0.198666
      73
             0.014176
                        0.014176
                                   0.198469
                                                             False
                                                                                  False
             0.052819
      38891
                        0.052819 0.158458
                                                             False
                                                                                  False
                                                             False
                                                                                  False
      38892
             0.052826
                        0.052826
                                   0.158479
      38893
             0.052812
                        0.052812
                                   0.158437
                                                             False
                                                                                  False
      38894
             0.052875
                        0.052875
                                   0.158625
                                                             False
                                                                                  False
      38895
             0.052404
                        0.052404
                                   0.157212
                                                             False
                                                                                  False
              customer_type_non-member
                                         customer_type_premium
      69
                                  False
                                                            True
      70
                                  False
                                                            True
      71
                                  False
                                                            True
      72
                                  False
                                                            True
      73
                                  False
                                                            True
      38891
                                  False
                                                            True
                                  False
                                                            True
      38892
      38893
                                  False
                                                            True
                                  False
      38894
                                                            True
      38895
                                  False
                                                            True
             \verb"customer_type_standard"
                                                            payment_type_credit card
                                       payment_type_cash
      69
                                                    False
                                False
                                                                                False
      70
                                False
                                                    False
                                                                                False
      71
                                False
                                                    False
                                                                                False
      72
                                False
                                                    False
                                                                                False
      73
                                False
                                                    False
                                                                                False
      38891
                                False
                                                                                False
                                                     True
```

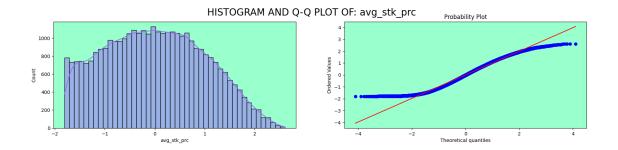
```
38892
                              False
                                                   True
                                                                             False
      38893
                              False
                                                   True
                                                                             False
      38894
                              False
                                                   True
                                                                             False
      38895
                              False
                                                                             False
                                                   True
             payment_type_debit card payment_type_e-wallet
      69
                               False
      70
                               False
                                                        True
      71
                               False
                                                        True
      72
                               False
                                                        True
      73
                               False
                                                        True
      38891
                               False
                                                       False
                               False
                                                       False
      38892
      38893
                               False
                                                       False
      38894
                               False
                                                       False
      38895
                               False
                                                       False
      [33924 rows x 18 columns]
[83]: cdf.columns
[83]: Index(['category', 'unit_price', 'quantity', 'total', 'avg_stk_prc',
             'temperature', 'week', 'day', 'hour', 'customer_type_basic',
             'customer type gold', 'customer type non-member',
             'customer_type_premium', 'customer_type_standard', 'payment_type_cash',
             'payment_type_credit card', 'payment_type_debit card',
             'payment_type_e-wallet'],
            dtype='object')
[84]: cols=['category', 'unit_price', 'quantity', 'total', 'avg_stk_prc',
             'temperature', 'week', 'day', 'hour', 'customer_type_basic',
             'customer_type_gold', 'customer_type_non-member',
             'customer type premium', 'customer type standard', 'payment type cash',
             'payment_type_credit card', 'payment_type_debit card',
             'payment type e-wallet']
      ss=PowerTransformer()
      cdf[cols]=ss.fit transform(cdf[cols])
      for i in cols:
          histandbox(i,cdf)
```

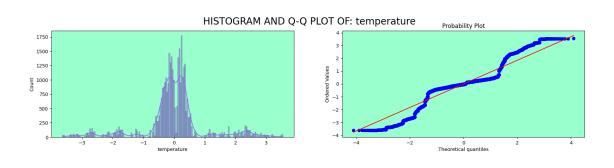


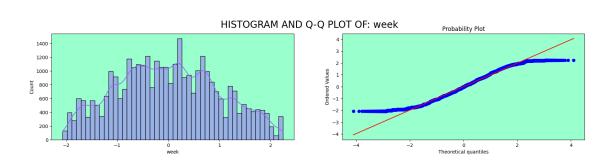


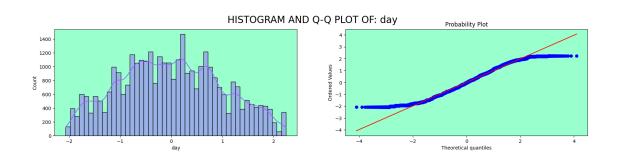


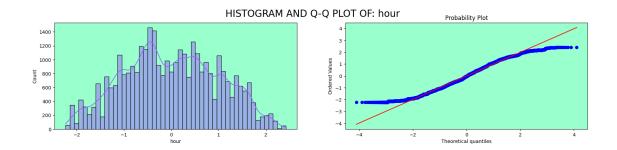


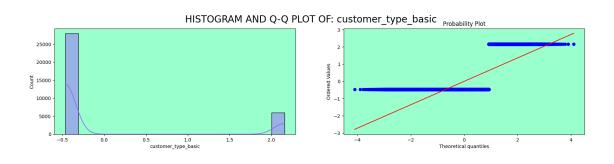


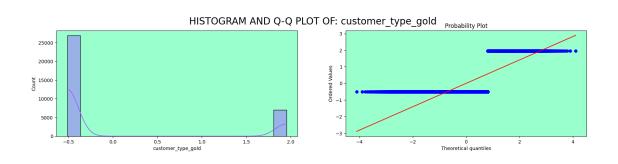


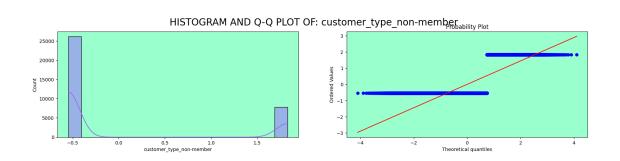


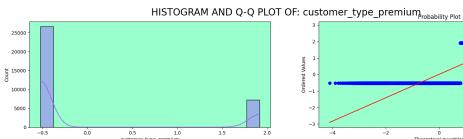


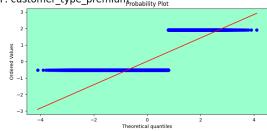


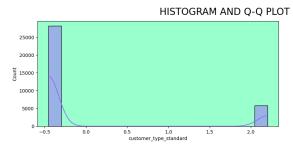


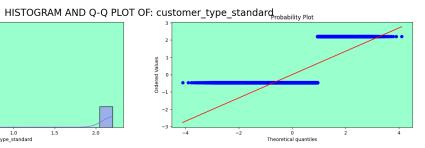


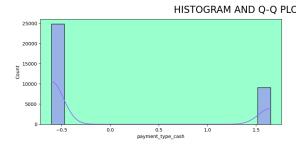


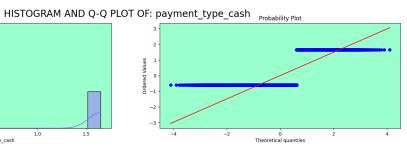


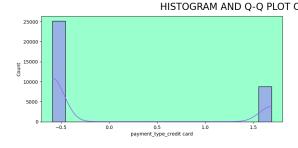


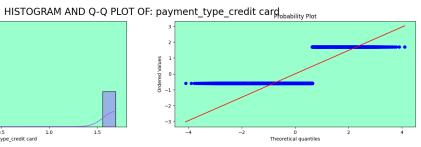


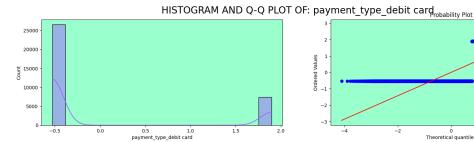


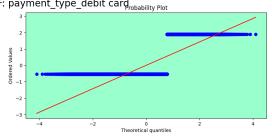


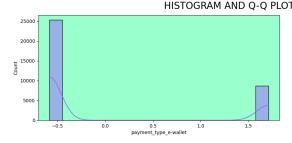


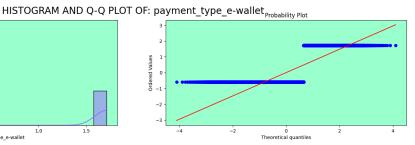












```
[85]: cc=['customer_type_basic',
             'customer_type_gold', 'customer_type_non-member',
             'customer_type_premium', 'customer_type_standard', 'payment_type_cash',
             'payment_type_credit card', 'payment_type_debit card',
             'payment_type_e-wallet']
      for i in cc:
          cdf[cc]=cdf[cc].astype('category')
```

[86]: cdf[cc].info()

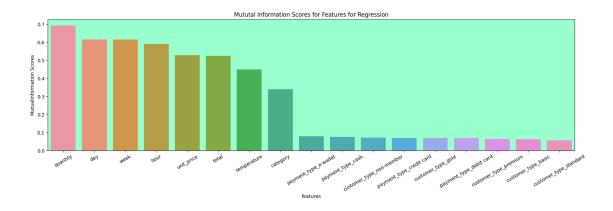
<class 'pandas.core.frame.DataFrame'> Index: 33924 entries, 69 to 38895 Data columns (total 9 columns):

#	Column	Non-Null Count	Dtype
0	customer_type_basic	33924 non-null	category
1	customer_type_gold	33924 non-null	category
2	customer_type_non-member	33924 non-null	category
3	customer_type_premium	33924 non-null	category
4	customer_type_standard	33924 non-null	category
5	payment_type_cash	33924 non-null	category
6	<pre>payment_type_credit card</pre>	33924 non-null	category
7	<pre>payment_type_debit card</pre>	33924 non-null	category
8	<pre>payment_type_e-wallet</pre>	33924 non-null	category

dtypes: category(9) memory usage: 564.3 KB

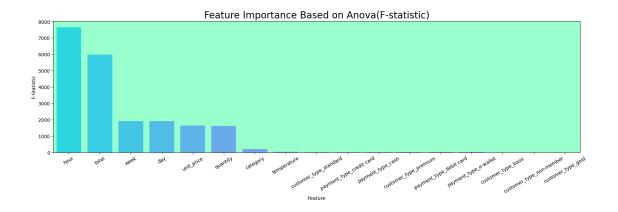
8 FEATURE IMPORTANCE: MUTUAL INFORMATION & F-STATISTIC for REGRESSION

```
[87]: y=cdf.avg_stk_prc
       x=cdf.drop(columns='avg_stk_prc')
[229]: from sklearn.feature_selection import mutual_info_regression
       d=['cus_gold', 'cus_non-member',
              'cus_premium', 'cus_standard', 'pymode_credit card',
              'pymode_debit card', 'pymode_e-wallet']
       mi=mutual_info_regression(x,y,discrete_features='auto')
[230]: miscores=mi.tolist()
       ms=pd.DataFrame({'features':x.columns,'MutualInformation Scores':miscores})
[230]:
                           features
                                     MutualInformation Scores
                                                      0.338898
       0
                           category
       1
                         unit_price
                                                      0.528565
       2
                                                      0.693032
                           quantity
       3
                                                      0.523817
                              total
       4
                        temperature
                                                      0.448207
       5
                                                      0.615157
                               week
       6
                                                      0.615265
                                day
       7
                                                      0.591473
                               hour
       8
                customer_type_basic
                                                      0.061260
       9
                 customer_type_gold
                                                      0.068757
       10
           customer_type_non-member
                                                      0.071831
                                                      0.064185
       11
              customer_type_premium
       12
             customer_type_standard
                                                      0.055076
       13
                  payment_type_cash
                                                      0.075531
       14
                                                      0.069140
          payment_type_credit card
       15
            payment_type_debit card
                                                      0.068745
       16
              payment_type_e-wallet
                                                      0.077961
[231]: ms=ms.sort values(by='MutualInformation Scores',ascending=False)
       plt.figure(figsize=(20,5))
       sns.barplot(data=ms,x='features',y='MutualInformation Scores')
       plt.xticks(rotation=30)
       plt.title('Mututal Information Scores for Features for Regression')
       plt.show()
```



```
[232]:
                             Feature F-Statistic P value
       7
                                 hour
                                           7653.04
                                                        0.00
       3
                                           5970.87
                                                        0.00
                                total
       5
                                 week
                                           1915.42
                                                        0.00
       6
                                  day
                                           1915.42
                                                        0.00
       1
                          unit_price
                                           1649.66
                                                        0.00
       2
                                           1620.00
                                                        0.00
                            quantity
       0
                            category
                                            196.28
                                                        0.00
       4
                                                        0.00
                         temperature
                                              30.03
       12
                                              19.75
                                                        0.00
             customer_type_standard
       14
           payment_type_credit card
                                              19.55
                                                        0.00
       13
                   payment_type_cash
                                              19.18
                                                        0.00
                                              18.99
                                                        0.00
       11
               customer_type_premium
       15
            payment_type_debit card
                                              11.29
                                                        0.00
       16
              payment_type_e-wallet
                                              10.07
                                                        0.00
       8
                 customer_type_basic
                                               0.20
                                                        0.66
       10
           customer_type_non-member
                                               0.18
                                                        0.68
       9
                  customer_type_gold
                                               0.09
                                                        0.76
```

```
[233]: plt.figure(figsize=(20,5))
    sns.barplot(data=anova,x='Feature',y='F-Statistic',palette='cool')
    plt.xticks(rotation=30)
    plt.title('Feature Importance Based on Anova(F-statistic)',fontsize=20)
    plt.show()
```



9 TRAIN-TEST SPLIT

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

[89]: xtrain,xtest,ytrain,ytest=train_test_split(x,y,shuffle=True,train_size=.80)
    print(f"xtrain: {xtrain.shape} and xtest: {xtest.shape}")
    print(f"ytrain: {ytrain.shape} and ytest: {ytest.shape}")

xtrain: (27139, 17) and xtest: (6785, 17)
    ytrain: (27139,) and ytest: (6785,)
```

10 PREDICTIVE MODEL BUILDING AND VALIDATION

11 LINEAR REGRESSION

```
[47]: lr=LinearRegression(n_jobs=-1,)
      g={'positive':[True,False],'fit_intercept':[True,False]
     p(g,lr,'linearreg')
     LinearRegression(n_jobs=-1)
     Best Parameters {'positive': False, 'fit_intercept': False}
     Best Score 22.65 %
[47]:
                      Metric_Name Values
      0
                          R.sq(\%) 23.110
      1
               Mean-Squared-Error
                                   0.757
      2 Rooted-Mean-Squared-Error
                                   0.850
             Mean-Absoulute-Error
      3
                                   0.723
            Median-Absoulte-Error
                                    0.666
```

12 HIST-GRADIENT-BOOSTING-REGRESSOR

```
[48]: from sklearn.ensemble import AdaBoostRegressor, HistGradientBoostingRegressor from sklearn.tree import DecisionTreeRegressor import warnings warnings ('ignore')
```

```
[49]: h=HistGradientBoostingRegressor(random_state=34563,max_bins=244,max_depth=30)
     g={'learning_rate':[0.1,0.01],
        'max_iter': [100,200,500,600,800,900],
        'max_leaf_nodes': [20,30],
       '12_regularization':[1,0.01],'tol':[1e-7,1e-8]}
     p(g,h,'histgrdbstreg')
     HistGradientBoostingRegressor(max_bins=244, max_depth=30, random_state=34563)
     Best Parameters {'tol': 1e-08, 'max_leaf_nodes': 30, 'max_iter': 800,
     'learning_rate': 0.1, 'l2_regularization': 1}
     Best Score 94.08 %
[49]:
                      Metric_Name Values
     0
                          R.sq(%) 94.810
               Mean-Squared-Error
                                   0.051
     1
     2 Rooted-Mean-Squared-Error
                                    0.408
             Mean-Absoulute-Error
     3
                                    0.166
            Median-Absoulte-Error
                                    0.126
          ADAPTIVE-BOOSTING-REGRESSOR
     13
[50]: ada=AdaBoostRegressor(estimator=DecisionTreeRegressor(max_depth=16))
     grid={'n_estimators':[7,8,10],
               'learning_rate': [1.2,1.6,2],
               'loss':['linear', 'square', 'exponential']}
     p(grid,ada, 'adabstreg')
     AdaBoostRegressor(estimator=DecisionTreeRegressor(max_depth=16))
     Best Parameters {'n_estimators': 8, 'loss': 'linear', 'learning_rate': 1.6}
     Best Score 97.39 %
[50]:
                      Metric_Name Values
                                  96.790
     0
                          R.sq(\%)
               Mean-Squared-Error
                                  0.032
     1
     2 Rooted-Mean-Squared-Error
                                    0.338
     3
             Mean-Absoulute-Error
                                    0.114
            Median-Absoulte-Error
                                    0.069
          BAGGINGREGRESSOR
     14
```

[51]: from sklearn.ensemble import BaggingRegressor

```
[52]: bag=BaggingRegressor(estimator=DecisionTreeRegressor(max_depth=17),oob_score=False,n_jobs=-1)
     bgrid={'n_estimators':[10,13,16]}
     p(bgrid,bag,'bagreg')
     BaggingRegressor(estimator=DecisionTreeRegressor(max_depth=17), n_jobs=-1)
     Best Parameters {'n_estimators': 16}
     Best Score 94.52 %
[52]:
                      Metric_Name Values
     0
                          R.sq(\%)
                                  94.540
               Mean-Squared-Error
     1
                                   0.054
     2 Rooted-Mean-Squared-Error
                                    0.384
             Mean-Absoulute-Error
     3
                                    0.148
            Median-Absoulte-Error
                                    0.083
     15
          RANDOMFORESTREGESSOR
[53]: from sklearn.ensemble import RandomForestRegressor
[54]: r=RandomForestRegressor(n_jobs=-1,oob_score=True)
     rgrid={'max_depth':[170,190,200,210],
             'max_features':['sqrt', 'log2'],
             'max_samples': [30,100,150,200],
             'max_leaf_nodes':[20,40,60,100]}
```

```
RandomForestRegressor(n_jobs=-1, oob_score=True)
```

```
Best Parameters {'max_samples': 200, 'max_leaf_nodes': 40, 'max_features':
'log2', 'max_depth': 170}
Best Score 25.39 %
```

```
[54]:
                       Metric_Name Values
      0
                           R.sq(\%)
                                    25.350
                Mean-Squared-Error
                                   0.735
      1
      2
       Rooted-Mean-Squared-Error
                                   0.844
              Mean-Absoulute-Error
      3
                                     0.713
            Median-Absoulte-Error
                                     0.662
```

p(rgrid,r,'randfrstreg')

16 RADIUS NEIGHBOR REGRESSOR

```
[55]: from sklearn.neighbors import RadiusNeighborsRegressor
```

```
[56]: rnr=RadiusNeighborsRegressor(algorithm='auto',n_jobs=-1)
      rng={'weights':['uniform', 'distance'],
          'p':[1,2], 'leaf_size':[30,40,50]}
      p(rng,rnr,'Radiusneighreg')
     RadiusNeighborsRegressor(n_jobs=-1)
     Best Parameters {'weights': 'distance', 'p': 1, 'leaf_size': 30}
     Best Score 93.77 %
[56]:
                      Metric_Name Values
                          R.sq(%) 94.760
      0
               Mean-Squared-Error
      1
                                   0.052
      2 Rooted-Mean-Squared-Error
                                   0.362
             Mean-Absoulute-Error
      3
                                   0.131
            Median-Absoulte-Error
                                    0.064
```

17 STOCHASTIC-GRADIENT-DESCENT-REGRESSOR

```
[57]: from sklearn.linear_model import SGDRegressor
[58]: sgd=SGDRegressor()
      sgdg={'penalty':['12', '11', 'elasticnet', None],
           'max_iter':[100,400,800], 'tol':[1e-3,1e-5,1e-8],'alpha':[0.1,.001,0.
       40001,1], 'learning_rate': ['constant', 'optimal', 'invscaling', 'adaptive']
      p(sgdg,sgd,'sgdreg')
     SGDRegressor()
     Best Parameters {'tol': 0.001, 'penalty': 'l1', 'max_iter': 100,
     'learning_rate': 'adaptive', 'alpha': 0.0001}
     Best Score 22.65 %
[58]:
                      Metric_Name Values
      0
                           R.sq(%) 23.110
      1
               Mean-Squared-Error 0.757
      2 Rooted-Mean-Squared-Error 0.850
              Mean-Absoulute-Error 0.723
      3
            Median-Absoulte-Error
                                    0.666
```

18 *ARTIFICIAL NEURAL NETWORK MODEL-MULTIPLE LINEAR REGRESSION PROBLEM**

```
[98]: import tensorflow as tf
       from tensorflow import keras
       from keras.layers import Dense
       from tensorflow.keras.losses import MeanSquaredError
       from keras.optimizers import SGD, Adam
[99]: regNN=tf.keras.Sequential([Dense(units=256,input_dim=xtrain.
        ⇔shape[1],activation='relu',name='l1'),
                                  Dense(units=128,activation='relu',name='12'),
                                  Dense(units=64, activation='relu', name='13'),
                                  Dense(units=32,activation='relu',name='14'),
                                  Dense(units=16,activation='relu',name='15'),
                                  Dense(units=8,activation='relu',name='16'),#2
                                  Dense(units=4,activation='relu',name='17'),
                                  Dense(units=1,activation='linear',name='18')
                                 ])
[100]: ms=tf.keras.losses.MeanSquaredError()
       sgd=tf.keras.optimizers.SGD(learning_rate=0.001,nesterov=True,momentum=0.
        ⇔001,ema_momentum=1)#epsilon=1e-03
       regNN.compile(optimizer=sgd,loss=ms,metrics=['mean_absolute_error',_
        ⇔'mean_squared_error'],)
[101]: regNN.summary()
```

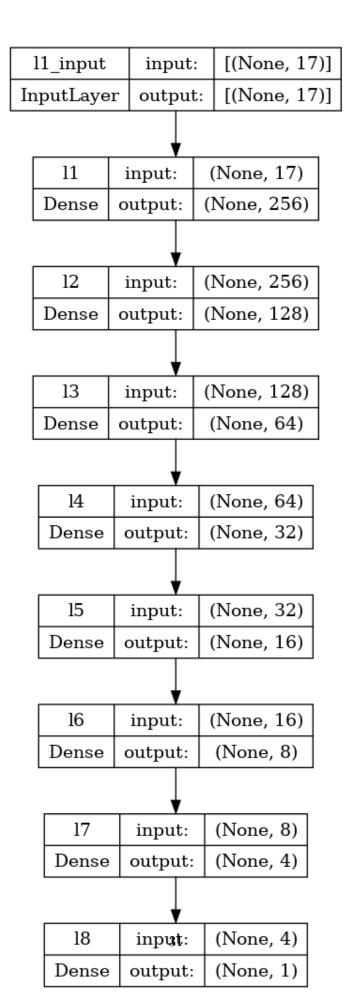
Model: "sequential_5"

Layer (type)	Output Shape	Param #
l1 (Dense)	(None, 256)	4608
12 (Dense)	(None, 128)	32896
13 (Dense)	(None, 64)	8256
14 (Dense)	(None, 32)	2080
15 (Dense)	(None, 16)	528
16 (Dense)	(None, 8)	136
17 (Dense)	(None, 4)	36
18 (Dense)	(None, 1)	5

Total params: 48,545 Trainable params: 48,545 Non-trainable params: 0

[102]: from tensorflow.keras.utils import plot_model
plot_model(regNN,show_shapes=True,to_file='regNN.png')

[102]:



```
→2,steps_per_epoch=250)
his=history.history
Epoch 1/350
mean_absolute_error: 0.8262 - mean_squared_error: 0.9670 - val_loss: 0.9567 -
val_mean_absolute_error: 0.8247 - val_mean_squared_error: 0.9567
Epoch 2/350
mean_absolute_error: 0.8041 - mean_squared_error: 0.9177 - val_loss: 0.9176 -
val_mean_absolute_error: 0.8065 - val_mean_squared_error: 0.9176
Epoch 3/350
mean_absolute_error: 0.7879 - mean_squared_error: 0.8843 - val_loss: 0.8887 -
val_mean_absolute_error: 0.7919 - val_mean_squared_error: 0.8887
Epoch 4/350
250/250 [============= ] - 1s 4ms/step - loss: 0.8601 -
mean_absolute_error: 0.7755 - mean_squared_error: 0.8601 - val_loss: 0.8689 -
val mean absolute error: 0.7826 - val mean squared error: 0.8689
Epoch 5/350
mean_absolute_error: 0.7663 - mean_squared_error: 0.8418 - val_loss: 0.8534 -
val_mean_absolute_error: 0.7751 - val_mean_squared_error: 0.8534
Epoch 6/350
mean_absolute_error: 0.7594 - mean_squared_error: 0.8276 - val_loss: 0.8404 -
val_mean_absolute_error: 0.7674 - val_mean_squared_error: 0.8404
Epoch 7/350
250/250 [============= ] - 1s 3ms/step - loss: 0.8166 -
mean_absolute_error: 0.7535 - mean_squared_error: 0.8166 - val_loss: 0.8314 -
val_mean_absolute_error: 0.7643 - val_mean_squared_error: 0.8314
Epoch 8/350
mean_absolute_error: 0.7493 - mean_squared_error: 0.8080 - val_loss: 0.8239 -
val_mean_absolute_error: 0.7613 - val_mean_squared_error: 0.8239
Epoch 9/350
mean_absolute_error: 0.7461 - mean_squared_error: 0.8009 - val_loss: 0.8163 -
val_mean_absolute_error: 0.7572 - val_mean_squared_error: 0.8163
Epoch 10/350
mean absolute error: 0.7431 - mean squared error: 0.7947 - val loss: 0.8105 -
val_mean_absolute_error: 0.7549 - val_mean_squared_error: 0.8105
```

[103]: history=regNN.fit(xtrain,ytrain,shuffle=True,epochs=350,validation_split=.

Epoch 11/350

```
mean_absolute_error: 0.7404 - mean_squared_error: 0.7894 - val_loss: 0.8067 -
val mean absolute error: 0.7544 - val mean squared error: 0.8067
Epoch 12/350
mean_absolute_error: 0.7388 - mean_squared_error: 0.7852 - val_loss: 0.8019 -
val_mean_absolute_error: 0.7513 - val_mean_squared_error: 0.8019
Epoch 13/350
mean_absolute_error: 0.7365 - mean_squared_error: 0.7813 - val_loss: 0.7994 -
val_mean_absolute_error: 0.7517 - val_mean_squared_error: 0.7994
Epoch 14/350
mean_absolute_error: 0.7356 - mean_squared_error: 0.7784 - val_loss: 0.7953 -
val_mean_absolute_error: 0.7482 - val_mean_squared_error: 0.7953
Epoch 15/350
250/250 [============ ] - 1s 3ms/step - loss: 0.7756 -
mean_absolute_error: 0.7340 - mean_squared_error: 0.7756 - val_loss: 0.7924 -
val_mean_absolute_error: 0.7461 - val_mean_squared_error: 0.7924
Epoch 16/350
mean_absolute_error: 0.7329 - mean_squared_error: 0.7731 - val_loss: 0.7903 -
val_mean_absolute_error: 0.7442 - val_mean_squared_error: 0.7903
Epoch 17/350
mean_absolute_error: 0.7315 - mean_squared_error: 0.7709 - val_loss: 0.7890 -
val_mean_absolute_error: 0.7461 - val_mean_squared_error: 0.7890
Epoch 18/350
mean_absolute_error: 0.7308 - mean_squared_error: 0.7688 - val_loss: 0.7876 -
val_mean_absolute_error: 0.7458 - val_mean_squared_error: 0.7876
Epoch 19/350
mean_absolute_error: 0.7299 - mean_squared_error: 0.7670 - val_loss: 0.7855 -
val_mean_absolute_error: 0.7445 - val_mean_squared_error: 0.7855
Epoch 20/350
250/250 [============= ] - 1s 4ms/step - loss: 0.7652 -
mean_absolute_error: 0.7292 - mean_squared_error: 0.7652 - val_loss: 0.7837 -
val_mean_absolute_error: 0.7427 - val_mean_squared_error: 0.7837
Epoch 21/350
mean_absolute_error: 0.7282 - mean_squared_error: 0.7637 - val_loss: 0.7826 -
val_mean_absolute_error: 0.7428 - val_mean_squared_error: 0.7826
Epoch 22/350
mean_absolute_error: 0.7276 - mean_squared_error: 0.7620 - val_loss: 0.7808 -
val_mean_absolute_error: 0.7409 - val_mean_squared_error: 0.7808
Epoch 23/350
```

```
mean_absolute_error: 0.7267 - mean_squared_error: 0.7606 - val_loss: 0.7797 -
val_mean_absolute_error: 0.7416 - val_mean_squared_error: 0.7797
Epoch 24/350
mean_absolute_error: 0.7261 - mean_squared_error: 0.7591 - val_loss: 0.7782 -
val_mean_absolute_error: 0.7405 - val_mean_squared_error: 0.7782
Epoch 25/350
mean_absolute_error: 0.7253 - mean_squared_error: 0.7577 - val_loss: 0.7772 -
val_mean_absolute_error: 0.7403 - val_mean_squared_error: 0.7772
Epoch 26/350
mean_absolute_error: 0.7250 - mean_squared_error: 0.7564 - val_loss: 0.7756 -
val_mean_absolute_error: 0.7382 - val_mean_squared_error: 0.7756
Epoch 27/350
250/250 [=========== ] - 1s 3ms/step - loss: 0.7551 -
mean_absolute_error: 0.7239 - mean_squared_error: 0.7551 - val_loss: 0.7756 -
val_mean_absolute_error: 0.7407 - val_mean_squared_error: 0.7756
Epoch 28/350
mean_absolute_error: 0.7239 - mean_squared_error: 0.7537 - val_loss: 0.7734 -
val_mean_absolute_error: 0.7368 - val_mean_squared_error: 0.7734
Epoch 29/350
mean_absolute_error: 0.7231 - mean_squared_error: 0.7527 - val_loss: 0.7723 -
val_mean_absolute_error: 0.7371 - val_mean_squared_error: 0.7723
Epoch 30/350
mean_absolute_error: 0.7224 - mean_squared_error: 0.7512 - val_loss: 0.7723 -
val_mean_absolute_error: 0.7393 - val_mean_squared_error: 0.7723
Epoch 31/350
250/250 [============= ] - 1s 3ms/step - loss: 0.7501 -
mean_absolute_error: 0.7220 - mean_squared_error: 0.7501 - val_loss: 0.7718 -
val_mean_absolute_error: 0.7393 - val_mean_squared_error: 0.7718
Epoch 32/350
250/250 [============ ] - 1s 4ms/step - loss: 0.7490 -
mean_absolute_error: 0.7214 - mean_squared_error: 0.7490 - val_loss: 0.7692 -
val_mean_absolute_error: 0.7365 - val_mean_squared_error: 0.7692
Epoch 33/350
mean_absolute_error: 0.7209 - mean_squared_error: 0.7478 - val_loss: 0.7681 -
val_mean_absolute_error: 0.7351 - val_mean_squared_error: 0.7681
Epoch 34/350
mean_absolute_error: 0.7201 - mean_squared_error: 0.7465 - val_loss: 0.7675 -
val_mean_absolute_error: 0.7365 - val_mean_squared_error: 0.7675
Epoch 35/350
```

```
mean_absolute_error: 0.7204 - mean_squared_error: 0.7456 - val_loss: 0.7661 -
val_mean_absolute_error: 0.7331 - val_mean_squared_error: 0.7661
Epoch 36/350
mean_absolute_error: 0.7193 - mean_squared_error: 0.7445 - val_loss: 0.7658 -
val_mean_absolute_error: 0.7358 - val_mean_squared_error: 0.7658
Epoch 37/350
mean_absolute_error: 0.7191 - mean_squared_error: 0.7434 - val_loss: 0.7643 -
val mean absolute error: 0.7338 - val mean squared error: 0.7643
Epoch 38/350
mean_absolute_error: 0.7184 - mean_squared_error: 0.7424 - val_loss: 0.7633 -
val_mean_absolute_error: 0.7328 - val_mean_squared_error: 0.7633
Epoch 39/350
250/250 [============ ] - 1s 3ms/step - loss: 0.7412 -
mean_absolute_error: 0.7177 - mean_squared_error: 0.7412 - val_loss: 0.7632 -
val_mean_absolute_error: 0.7351 - val_mean_squared_error: 0.7632
Epoch 40/350
mean_absolute_error: 0.7174 - mean_squared_error: 0.7399 - val_loss: 0.7619 -
val_mean_absolute_error: 0.7340 - val_mean_squared_error: 0.7619
Epoch 41/350
mean_absolute_error: 0.7171 - mean_squared_error: 0.7392 - val_loss: 0.7605 -
val_mean_absolute_error: 0.7317 - val_mean_squared_error: 0.7605
Epoch 42/350
mean_absolute_error: 0.7163 - mean_squared_error: 0.7380 - val_loss: 0.7596 -
val_mean_absolute_error: 0.7314 - val_mean_squared_error: 0.7596
Epoch 43/350
250/250 [============= ] - 1s 3ms/step - loss: 0.7370 -
mean_absolute_error: 0.7158 - mean_squared_error: 0.7370 - val_loss: 0.7600 -
val_mean_absolute_error: 0.7343 - val_mean_squared_error: 0.7600
Epoch 44/350
250/250 [============= ] - 1s 3ms/step - loss: 0.7359 -
mean_absolute_error: 0.7155 - mean_squared_error: 0.7359 - val_loss: 0.7579 -
val_mean_absolute_error: 0.7309 - val_mean_squared_error: 0.7579
Epoch 45/350
mean_absolute_error: 0.7149 - mean_squared_error: 0.7348 - val_loss: 0.7568 -
val_mean_absolute_error: 0.7305 - val_mean_squared_error: 0.7568
Epoch 46/350
mean_absolute_error: 0.7145 - mean_squared_error: 0.7338 - val_loss: 0.7561 -
val_mean_absolute_error: 0.7303 - val_mean_squared_error: 0.7561
Epoch 47/350
```

```
mean_absolute_error: 0.7139 - mean_squared_error: 0.7329 - val_loss: 0.7557 -
val_mean_absolute_error: 0.7307 - val_mean_squared_error: 0.7557
Epoch 48/350
mean_absolute_error: 0.7136 - mean_squared_error: 0.7318 - val_loss: 0.7545 -
val_mean_absolute_error: 0.7282 - val_mean_squared_error: 0.7545
Epoch 49/350
mean_absolute_error: 0.7127 - mean_squared_error: 0.7306 - val_loss: 0.7542 -
val_mean_absolute_error: 0.7303 - val_mean_squared_error: 0.7542
Epoch 50/350
mean_absolute_error: 0.7125 - mean_squared_error: 0.7299 - val_loss: 0.7561 -
val_mean_absolute_error: 0.7343 - val_mean_squared_error: 0.7561
Epoch 51/350
250/250 [=========== ] - 1s 3ms/step - loss: 0.7291 -
mean_absolute_error: 0.7122 - mean_squared_error: 0.7291 - val_loss: 0.7525 -
val_mean_absolute_error: 0.7292 - val_mean_squared_error: 0.7525
Epoch 52/350
mean_absolute_error: 0.7115 - mean_squared_error: 0.7279 - val_loss: 0.7520 -
val_mean_absolute_error: 0.7300 - val_mean_squared_error: 0.7520
Epoch 53/350
mean_absolute_error: 0.7111 - mean_squared_error: 0.7268 - val_loss: 0.7530 -
val_mean_absolute_error: 0.7330 - val_mean_squared_error: 0.7530
Epoch 54/350
mean_absolute_error: 0.7104 - mean_squared_error: 0.7259 - val_loss: 0.7515 -
val_mean_absolute_error: 0.7319 - val_mean_squared_error: 0.7515
Epoch 55/350
250/250 [============ ] - 1s 3ms/step - loss: 0.7247 -
mean_absolute_error: 0.7103 - mean_squared_error: 0.7247 - val_loss: 0.7503 -
val_mean_absolute_error: 0.7307 - val_mean_squared_error: 0.7503
Epoch 56/350
250/250 [============ ] - 1s 3ms/step - loss: 0.7238 -
mean_absolute_error: 0.7096 - mean_squared_error: 0.7238 - val_loss: 0.7478 -
val_mean_absolute_error: 0.7277 - val_mean_squared_error: 0.7478
Epoch 57/350
mean_absolute_error: 0.7093 - mean_squared_error: 0.7229 - val_loss: 0.7470 -
val_mean_absolute_error: 0.7260 - val_mean_squared_error: 0.7470
Epoch 58/350
250/250 [============ ] - 1s 4ms/step - loss: 0.7218 -
mean_absolute_error: 0.7084 - mean_squared_error: 0.7218 - val_loss: 0.7465 -
val_mean_absolute_error: 0.7278 - val_mean_squared_error: 0.7465
Epoch 59/350
```

```
mean_absolute_error: 0.7082 - mean_squared_error: 0.7209 - val_loss: 0.7462 -
val_mean_absolute_error: 0.7272 - val_mean_squared_error: 0.7462
Epoch 60/350
mean_absolute_error: 0.7078 - mean_squared_error: 0.7200 - val_loss: 0.7456 -
val_mean_absolute_error: 0.7234 - val_mean_squared_error: 0.7456
Epoch 61/350
mean_absolute_error: 0.7069 - mean_squared_error: 0.7188 - val_loss: 0.7463 -
val_mean_absolute_error: 0.7296 - val_mean_squared_error: 0.7463
Epoch 62/350
mean_absolute_error: 0.7067 - mean_squared_error: 0.7180 - val_loss: 0.7429 -
val_mean_absolute_error: 0.7236 - val_mean_squared_error: 0.7429
Epoch 63/350
250/250 [============ ] - 1s 3ms/step - loss: 0.7172 -
mean_absolute_error: 0.7065 - mean_squared_error: 0.7172 - val_loss: 0.7434 -
val_mean_absolute_error: 0.7203 - val_mean_squared_error: 0.7434
Epoch 64/350
mean_absolute_error: 0.7057 - mean_squared_error: 0.7162 - val_loss: 0.7418 -
val_mean_absolute_error: 0.7213 - val_mean_squared_error: 0.7418
Epoch 65/350
mean_absolute_error: 0.7050 - mean_squared_error: 0.7149 - val_loss: 0.7413 -
val_mean_absolute_error: 0.7249 - val_mean_squared_error: 0.7413
Epoch 66/350
mean_absolute_error: 0.7048 - mean_squared_error: 0.7142 - val_loss: 0.7399 -
val_mean_absolute_error: 0.7211 - val_mean_squared_error: 0.7399
Epoch 67/350
mean_absolute_error: 0.7043 - mean_squared_error: 0.7133 - val_loss: 0.7398 -
val_mean_absolute_error: 0.7210 - val_mean_squared_error: 0.7398
Epoch 68/350
250/250 [============= ] - 1s 3ms/step - loss: 0.7127 -
mean_absolute_error: 0.7041 - mean_squared_error: 0.7127 - val_loss: 0.7389 -
val_mean_absolute_error: 0.7228 - val_mean_squared_error: 0.7389
Epoch 69/350
mean_absolute_error: 0.7032 - mean_squared_error: 0.7113 - val_loss: 0.7377 -
val_mean_absolute_error: 0.7205 - val_mean_squared_error: 0.7377
Epoch 70/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.7104 -
mean_absolute_error: 0.7031 - mean_squared_error: 0.7104 - val_loss: 0.7370 -
val_mean_absolute_error: 0.7188 - val_mean_squared_error: 0.7370
Epoch 71/350
```

```
mean_absolute_error: 0.7024 - mean_squared_error: 0.7093 - val_loss: 0.7365 -
val mean absolute error: 0.7209 - val mean squared error: 0.7365
Epoch 72/350
mean_absolute_error: 0.7018 - mean_squared_error: 0.7083 - val_loss: 0.7355 -
val_mean_absolute_error: 0.7192 - val_mean_squared_error: 0.7355
Epoch 73/350
mean_absolute_error: 0.7015 - mean_squared_error: 0.7074 - val_loss: 0.7348 -
val mean absolute error: 0.7208 - val mean squared error: 0.7348
Epoch 74/350
mean_absolute_error: 0.7005 - mean_squared_error: 0.7061 - val_loss: 0.7334 -
val_mean_absolute_error: 0.7198 - val_mean_squared_error: 0.7334
Epoch 75/350
250/250 [============ ] - 1s 3ms/step - loss: 0.7052 -
mean_absolute_error: 0.7004 - mean_squared_error: 0.7052 - val_loss: 0.7333 -
val_mean_absolute_error: 0.7205 - val_mean_squared_error: 0.7333
Epoch 76/350
mean_absolute_error: 0.6994 - mean_squared_error: 0.7040 - val_loss: 0.7388 -
val_mean_absolute_error: 0.7287 - val_mean_squared_error: 0.7388
Epoch 77/350
mean_absolute_error: 0.6993 - mean_squared_error: 0.7028 - val_loss: 0.7309 -
val_mean_absolute_error: 0.7164 - val_mean_squared_error: 0.7309
Epoch 78/350
mean_absolute_error: 0.6984 - mean_squared_error: 0.7019 - val_loss: 0.7308 -
val_mean_absolute_error: 0.7149 - val_mean_squared_error: 0.7308
Epoch 79/350
250/250 [============= ] - 1s 3ms/step - loss: 0.7008 -
mean_absolute_error: 0.6979 - mean_squared_error: 0.7008 - val_loss: 0.7299 -
val_mean_absolute_error: 0.7187 - val_mean_squared_error: 0.7299
Epoch 80/350
250/250 [============= ] - 1s 4ms/step - loss: 0.6997 -
mean_absolute_error: 0.6973 - mean_squared_error: 0.6997 - val_loss: 0.7289 -
val_mean_absolute_error: 0.7168 - val_mean_squared_error: 0.7289
Epoch 81/350
mean_absolute_error: 0.6966 - mean_squared_error: 0.6981 - val_loss: 0.7303 -
val_mean_absolute_error: 0.7219 - val_mean_squared_error: 0.7303
Epoch 82/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.6970 -
mean_absolute_error: 0.6958 - mean_squared_error: 0.6970 - val_loss: 0.7309 -
val_mean_absolute_error: 0.7237 - val_mean_squared_error: 0.7309
Epoch 83/350
```

```
mean_absolute_error: 0.6955 - mean_squared_error: 0.6957 - val_loss: 0.7257 -
val_mean_absolute_error: 0.7141 - val_mean_squared_error: 0.7257
Epoch 84/350
mean_absolute_error: 0.6945 - mean_squared_error: 0.6944 - val_loss: 0.7241 -
val_mean_absolute_error: 0.7161 - val_mean_squared_error: 0.7241
Epoch 85/350
mean_absolute_error: 0.6938 - mean_squared_error: 0.6932 - val_loss: 0.7254 -
val mean absolute error: 0.7203 - val mean squared error: 0.7254
Epoch 86/350
mean_absolute_error: 0.6931 - mean_squared_error: 0.6916 - val_loss: 0.7213 -
val_mean_absolute_error: 0.7126 - val_mean_squared_error: 0.7213
Epoch 87/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.6909 -
mean_absolute_error: 0.6926 - mean_squared_error: 0.6909 - val_loss: 0.7219 -
val_mean_absolute_error: 0.7171 - val_mean_squared_error: 0.7219
Epoch 88/350
mean_absolute_error: 0.6916 - mean_squared_error: 0.6891 - val_loss: 0.7236 -
val_mean_absolute_error: 0.7190 - val_mean_squared_error: 0.7236
Epoch 89/350
mean_absolute_error: 0.6913 - mean_squared_error: 0.6873 - val_loss: 0.7198 -
val_mean_absolute_error: 0.7071 - val_mean_squared_error: 0.7198
Epoch 90/350
mean_absolute_error: 0.6893 - mean_squared_error: 0.6853 - val_loss: 0.7293 -
val_mean_absolute_error: 0.7263 - val_mean_squared_error: 0.7293
Epoch 91/350
250/250 [============ ] - 1s 3ms/step - loss: 0.6844 -
mean_absolute_error: 0.6893 - mean_squared_error: 0.6844 - val_loss: 0.7163 -
val_mean_absolute_error: 0.7066 - val_mean_squared_error: 0.7163
Epoch 92/350
250/250 [============ ] - 1s 4ms/step - loss: 0.6828 -
mean_absolute_error: 0.6877 - mean_squared_error: 0.6828 - val_loss: 0.7126 -
val_mean_absolute_error: 0.7086 - val_mean_squared_error: 0.7126
Epoch 93/350
mean_absolute_error: 0.6868 - mean_squared_error: 0.6804 - val_loss: 0.7187 -
val_mean_absolute_error: 0.7195 - val_mean_squared_error: 0.7187
Epoch 94/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.6796 -
mean_absolute_error: 0.6868 - mean_squared_error: 0.6796 - val_loss: 0.7198 -
val_mean_absolute_error: 0.7205 - val_mean_squared_error: 0.7198
Epoch 95/350
```

```
mean_absolute_error: 0.6856 - mean_squared_error: 0.6774 - val_loss: 0.7085 -
val_mean_absolute_error: 0.7022 - val_mean_squared_error: 0.7085
Epoch 96/350
mean_absolute_error: 0.6845 - mean_squared_error: 0.6750 - val_loss: 0.7083 -
val_mean_absolute_error: 0.7015 - val_mean_squared_error: 0.7083
Epoch 97/350
mean_absolute_error: 0.6837 - mean_squared_error: 0.6735 - val_loss: 0.7093 -
val mean absolute error: 0.6973 - val mean squared error: 0.7093
Epoch 98/350
mean_absolute_error: 0.6820 - mean_squared_error: 0.6711 - val_loss: 0.7015 -
val_mean_absolute_error: 0.7035 - val_mean_squared_error: 0.7015
Epoch 99/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.6685 -
mean_absolute_error: 0.6810 - mean_squared_error: 0.6685 - val_loss: 0.7013 -
val_mean_absolute_error: 0.6968 - val_mean_squared_error: 0.7013
Epoch 100/350
mean_absolute_error: 0.6797 - mean_squared_error: 0.6667 - val_loss: 0.7095 -
val_mean_absolute_error: 0.6925 - val_mean_squared_error: 0.7095
Epoch 101/350
mean_absolute_error: 0.6784 - mean_squared_error: 0.6645 - val_loss: 0.6959 -
val_mean_absolute_error: 0.6929 - val_mean_squared_error: 0.6959
Epoch 102/350
mean_absolute_error: 0.6769 - mean_squared_error: 0.6616 - val_loss: 0.6944 -
val_mean_absolute_error: 0.6905 - val_mean_squared_error: 0.6944
Epoch 103/350
250/250 [============= ] - 1s 4ms/step - loss: 0.6587 -
mean_absolute_error: 0.6756 - mean_squared_error: 0.6587 - val_loss: 0.6886 -
val_mean_absolute_error: 0.6974 - val_mean_squared_error: 0.6886
Epoch 104/350
250/250 [============ ] - 1s 3ms/step - loss: 0.6556 -
mean_absolute_error: 0.6734 - mean_squared_error: 0.6556 - val_loss: 0.6893 -
val_mean_absolute_error: 0.6893 - val_mean_squared_error: 0.6893
Epoch 105/350
mean_absolute_error: 0.6714 - mean_squared_error: 0.6522 - val_loss: 0.6842 -
val_mean_absolute_error: 0.6988 - val_mean_squared_error: 0.6842
Epoch 106/350
mean_absolute_error: 0.6697 - mean_squared_error: 0.6487 - val_loss: 0.6782 -
val_mean_absolute_error: 0.6941 - val_mean_squared_error: 0.6782
Epoch 107/350
```

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mean_absolute_error: 0.6689 - mean_squared_error: 0.6461 - val_loss: 0.6745 -
val_mean_absolute_error: 0.6863 - val_mean_squared_error: 0.6745
Epoch 108/350
mean_absolute_error: 0.6663 - mean_squared_error: 0.6414 - val_loss: 0.6768 -
val_mean_absolute_error: 0.6969 - val_mean_squared_error: 0.6768
Epoch 109/350
mean_absolute_error: 0.6643 - mean_squared_error: 0.6385 - val_loss: 0.6659 -
val mean absolute error: 0.6773 - val mean squared error: 0.6659
Epoch 110/350
mean_absolute_error: 0.6618 - mean_squared_error: 0.6332 - val_loss: 0.6646 -
val_mean_absolute_error: 0.6784 - val_mean_squared_error: 0.6646
Epoch 111/350
250/250 [=========== ] - 1s 3ms/step - loss: 0.6285 -
mean_absolute_error: 0.6591 - mean_squared_error: 0.6285 - val_loss: 0.6714 -
val_mean_absolute_error: 0.6666 - val_mean_squared_error: 0.6714
Epoch 112/350
mean_absolute_error: 0.6560 - mean_squared_error: 0.6235 - val_loss: 0.6526 -
val_mean_absolute_error: 0.6808 - val_mean_squared_error: 0.6526
Epoch 113/350
mean_absolute_error: 0.6535 - mean_squared_error: 0.6187 - val_loss: 0.6736 -
val_mean_absolute_error: 0.7012 - val_mean_squared_error: 0.6736
Epoch 114/350
mean_absolute_error: 0.6495 - mean_squared_error: 0.6123 - val_loss: 0.6438 -
val_mean_absolute_error: 0.6596 - val_mean_squared_error: 0.6438
Epoch 115/350
250/250 [============ ] - 1s 3ms/step - loss: 0.6043 -
mean_absolute_error: 0.6454 - mean_squared_error: 0.6043 - val_loss: 0.6241 -
val_mean_absolute_error: 0.6596 - val_mean_squared_error: 0.6241
Epoch 116/350
250/250 [============= ] - 1s 3ms/step - loss: 0.5958 -
mean_absolute_error: 0.6398 - mean_squared_error: 0.5958 - val_loss: 0.6746 -
val_mean_absolute_error: 0.7031 - val_mean_squared_error: 0.6746
Epoch 117/350
mean_absolute_error: 0.6372 - mean_squared_error: 0.5908 - val_loss: 0.6154 -
val_mean_absolute_error: 0.6458 - val_mean_squared_error: 0.6154
Epoch 118/350
mean_absolute_error: 0.6328 - mean_squared_error: 0.5835 - val_loss: 0.6061 -
val_mean_absolute_error: 0.6453 - val_mean_squared_error: 0.6061
Epoch 119/350
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mean_absolute_error: 0.6274 - mean_squared_error: 0.5728 - val_loss: 0.6388 -
val_mean_absolute_error: 0.6836 - val_mean_squared_error: 0.6388
Epoch 120/350
mean_absolute_error: 0.6213 - mean_squared_error: 0.5638 - val_loss: 0.5731 -
val_mean_absolute_error: 0.6252 - val_mean_squared_error: 0.5731
Epoch 121/350
mean_absolute_error: 0.6195 - mean_squared_error: 0.5621 - val_loss: 0.5678 -
val_mean_absolute_error: 0.6169 - val_mean_squared_error: 0.5678
Epoch 122/350
mean_absolute_error: 0.6118 - mean_squared_error: 0.5498 - val_loss: 0.5575 -
val_mean_absolute_error: 0.6091 - val_mean_squared_error: 0.5575
Epoch 123/350
250/250 [=========== ] - 1s 3ms/step - loss: 0.5457 -
mean_absolute_error: 0.6085 - mean_squared_error: 0.5457 - val_loss: 0.5408 -
val_mean_absolute_error: 0.6077 - val_mean_squared_error: 0.5408
Epoch 124/350
mean_absolute_error: 0.6109 - mean_squared_error: 0.5526 - val_loss: 0.5551 -
val_mean_absolute_error: 0.6379 - val_mean_squared_error: 0.5551
Epoch 125/350
mean_absolute_error: 0.6030 - mean_squared_error: 0.5415 - val_loss: 0.5478 -
val_mean_absolute_error: 0.5986 - val_mean_squared_error: 0.5478
Epoch 126/350
mean_absolute_error: 0.6038 - mean_squared_error: 0.5470 - val_loss: 0.5041 -
val_mean_absolute_error: 0.5858 - val_mean_squared_error: 0.5041
Epoch 127/350
250/250 [============= ] - 1s 4ms/step - loss: 0.5412 -
mean_absolute_error: 0.6010 - mean_squared_error: 0.5412 - val_loss: 0.5158 -
val_mean_absolute_error: 0.6073 - val_mean_squared_error: 0.5158
Epoch 128/350
250/250 [============ ] - 1s 3ms/step - loss: 0.5414 -
mean_absolute_error: 0.6004 - mean_squared_error: 0.5414 - val_loss: 0.5738 -
val_mean_absolute_error: 0.6514 - val_mean_squared_error: 0.5738
Epoch 129/350
mean_absolute_error: 0.5925 - mean_squared_error: 0.5254 - val_loss: 0.5754 -
val_mean_absolute_error: 0.6516 - val_mean_squared_error: 0.5754
Epoch 130/350
mean_absolute_error: 0.6003 - mean_squared_error: 0.5404 - val_loss: 0.5209 -
val_mean_absolute_error: 0.6209 - val_mean_squared_error: 0.5209
Epoch 131/350
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mean_absolute_error: 0.5962 - mean_squared_error: 0.5364 - val_loss: 0.5760 -
val mean absolute error: 0.5867 - val mean squared error: 0.5760
Epoch 132/350
mean_absolute_error: 0.5841 - mean_squared_error: 0.5118 - val_loss: 0.4600 -
val_mean_absolute_error: 0.5785 - val_mean_squared_error: 0.4600
Epoch 133/350
mean_absolute_error: 0.5907 - mean_squared_error: 0.5281 - val_loss: 0.6208 -
val mean absolute error: 0.5962 - val mean squared error: 0.6208
Epoch 134/350
mean_absolute_error: 0.5851 - mean_squared_error: 0.5166 - val_loss: 0.7309 -
val_mean_absolute_error: 0.6485 - val_mean_squared_error: 0.7309
Epoch 135/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.5077 -
mean_absolute_error: 0.5781 - mean_squared_error: 0.5077 - val_loss: 0.4248 -
val_mean_absolute_error: 0.5409 - val_mean_squared_error: 0.4248
Epoch 136/350
mean_absolute_error: 0.5823 - mean_squared_error: 0.5152 - val_loss: 0.8753 -
val_mean_absolute_error: 0.7162 - val_mean_squared_error: 0.8753
Epoch 137/350
mean_absolute_error: 0.5734 - mean_squared_error: 0.5016 - val_loss: 0.5821 -
val_mean_absolute_error: 0.6546 - val_mean_squared_error: 0.5821
Epoch 138/350
mean_absolute_error: 0.5599 - mean_squared_error: 0.4741 - val_loss: 0.7032 -
val_mean_absolute_error: 0.6346 - val_mean_squared_error: 0.7032
Epoch 139/350
250/250 [============= ] - 1s 3ms/step - loss: 0.5079 -
mean_absolute_error: 0.5766 - mean_squared_error: 0.5079 - val_loss: 0.5110 -
val_mean_absolute_error: 0.5471 - val_mean_squared_error: 0.5110
Epoch 140/350
250/250 [============= ] - 1s 3ms/step - loss: 0.4893 -
mean_absolute_error: 0.5650 - mean_squared_error: 0.4893 - val_loss: 0.7646 -
val_mean_absolute_error: 0.6708 - val_mean_squared_error: 0.7646
Epoch 141/350
mean_absolute_error: 0.5504 - mean_squared_error: 0.4652 - val_loss: 0.3903 -
val_mean_absolute_error: 0.5070 - val_mean_squared_error: 0.3903
Epoch 142/350
mean_absolute_error: 0.5686 - mean_squared_error: 0.4946 - val_loss: 0.4126 -
val_mean_absolute_error: 0.5104 - val_mean_squared_error: 0.4126
Epoch 143/350
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mean_absolute_error: 0.5538 - mean_squared_error: 0.4699 - val_loss: 0.3922 -
val mean absolute error: 0.4941 - val mean squared error: 0.3922
Epoch 144/350
mean_absolute_error: 0.5502 - mean_squared_error: 0.4675 - val_loss: 0.3979 -
val_mean_absolute_error: 0.4913 - val_mean_squared_error: 0.3979
Epoch 145/350
mean_absolute_error: 0.5533 - mean_squared_error: 0.4741 - val_loss: 0.4688 -
val_mean_absolute_error: 0.5813 - val_mean_squared_error: 0.4688
Epoch 146/350
mean_absolute_error: 0.5483 - mean_squared_error: 0.4605 - val_loss: 0.5311 -
val_mean_absolute_error: 0.5537 - val_mean_squared_error: 0.5311
Epoch 147/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.4435 -
mean_absolute_error: 0.5344 - mean_squared_error: 0.4435 - val_loss: 0.3713 -
val_mean_absolute_error: 0.5009 - val_mean_squared_error: 0.3713
Epoch 148/350
mean_absolute_error: 0.5417 - mean_squared_error: 0.4521 - val_loss: 0.3375 -
val_mean_absolute_error: 0.4819 - val_mean_squared_error: 0.3375
Epoch 149/350
mean_absolute_error: 0.5375 - mean_squared_error: 0.4433 - val_loss: 0.4895 -
val_mean_absolute_error: 0.5290 - val_mean_squared_error: 0.4895
Epoch 150/350
mean_absolute_error: 0.5400 - mean_squared_error: 0.4502 - val_loss: 0.3861 -
val_mean_absolute_error: 0.5324 - val_mean_squared_error: 0.3861
Epoch 151/350
mean_absolute_error: 0.5234 - mean_squared_error: 0.4246 - val_loss: 0.3988 -
val_mean_absolute_error: 0.4829 - val_mean_squared_error: 0.3988
Epoch 152/350
250/250 [============ ] - 1s 4ms/step - loss: 0.4432 -
mean_absolute_error: 0.5362 - mean_squared_error: 0.4432 - val_loss: 0.3141 -
val_mean_absolute_error: 0.4696 - val_mean_squared_error: 0.3141
Epoch 153/350
mean_absolute_error: 0.5181 - mean_squared_error: 0.4139 - val_loss: 0.4216 -
val_mean_absolute_error: 0.5586 - val_mean_squared_error: 0.4216
Epoch 154/350
mean_absolute_error: 0.5211 - mean_squared_error: 0.4202 - val_loss: 0.3185 -
val_mean_absolute_error: 0.4351 - val_mean_squared_error: 0.3185
Epoch 155/350
```

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mean_absolute_error: 0.5180 - mean_squared_error: 0.4174 - val_loss: 0.5576 -
val_mean_absolute_error: 0.5777 - val_mean_squared_error: 0.5576
Epoch 156/350
mean_absolute_error: 0.5178 - mean_squared_error: 0.4138 - val_loss: 0.3668 -
val_mean_absolute_error: 0.5201 - val_mean_squared_error: 0.3668
Epoch 157/350
mean_absolute_error: 0.5045 - mean_squared_error: 0.3931 - val_loss: 0.4985 -
val_mean_absolute_error: 0.5406 - val_mean_squared_error: 0.4985
Epoch 158/350
mean_absolute_error: 0.5017 - mean_squared_error: 0.3899 - val_loss: 0.4606 -
val_mean_absolute_error: 0.5944 - val_mean_squared_error: 0.4606
Epoch 159/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.3983 -
mean_absolute_error: 0.5060 - mean_squared_error: 0.3983 - val_loss: 0.5730 -
val_mean_absolute_error: 0.6559 - val_mean_squared_error: 0.5730
Epoch 160/350
mean_absolute_error: 0.5012 - mean_squared_error: 0.3889 - val_loss: 0.2995 -
val_mean_absolute_error: 0.4177 - val_mean_squared_error: 0.2995
Epoch 161/350
mean_absolute_error: 0.4944 - mean_squared_error: 0.3802 - val_loss: 0.3699 -
val_mean_absolute_error: 0.4668 - val_mean_squared_error: 0.3699
Epoch 162/350
mean_absolute_error: 0.4869 - mean_squared_error: 0.3711 - val_loss: 0.2473 -
val_mean_absolute_error: 0.3938 - val_mean_squared_error: 0.2473
Epoch 163/350
250/250 [============= ] - 1s 3ms/step - loss: 0.3702 -
mean_absolute_error: 0.4870 - mean_squared_error: 0.3702 - val_loss: 0.2505 -
val_mean_absolute_error: 0.3998 - val_mean_squared_error: 0.2505
Epoch 164/350
250/250 [============ ] - 1s 3ms/step - loss: 0.3536 -
mean_absolute_error: 0.4744 - mean_squared_error: 0.3536 - val_loss: 0.3640 -
val_mean_absolute_error: 0.5108 - val_mean_squared_error: 0.3640
Epoch 165/350
mean_absolute_error: 0.4772 - mean_squared_error: 0.3570 - val_loss: 0.4506 -
val_mean_absolute_error: 0.5773 - val_mean_squared_error: 0.4506
Epoch 166/350
mean_absolute_error: 0.4797 - mean_squared_error: 0.3578 - val_loss: 0.2967 -
val_mean_absolute_error: 0.4161 - val_mean_squared_error: 0.2967
Epoch 167/350
```

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mean_absolute_error: 0.4756 - mean_squared_error: 0.3524 - val_loss: 0.2721 -
val_mean_absolute_error: 0.4311 - val_mean_squared_error: 0.2721
Epoch 168/350
mean_absolute_error: 0.4843 - mean_squared_error: 0.3627 - val_loss: 0.2927 -
val_mean_absolute_error: 0.4508 - val_mean_squared_error: 0.2927
Epoch 169/350
mean_absolute_error: 0.4797 - mean_squared_error: 0.3602 - val_loss: 0.4613 -
val mean absolute error: 0.5920 - val mean squared error: 0.4613
Epoch 170/350
mean_absolute_error: 0.4679 - mean_squared_error: 0.3443 - val_loss: 0.2683 -
val_mean_absolute_error: 0.3915 - val_mean_squared_error: 0.2683
Epoch 171/350
250/250 [=========== ] - 1s 3ms/step - loss: 0.3303 -
mean_absolute_error: 0.4586 - mean_squared_error: 0.3303 - val_loss: 0.2326 -
val_mean_absolute_error: 0.3846 - val_mean_squared_error: 0.2326
Epoch 172/350
mean_absolute_error: 0.4662 - mean_squared_error: 0.3381 - val_loss: 0.3395 -
val_mean_absolute_error: 0.4425 - val_mean_squared_error: 0.3395
Epoch 173/350
mean_absolute_error: 0.4643 - mean_squared_error: 0.3378 - val_loss: 0.4840 -
val_mean_absolute_error: 0.5869 - val_mean_squared_error: 0.4840
Epoch 174/350
mean_absolute_error: 0.4494 - mean_squared_error: 0.3179 - val_loss: 0.6881 -
val_mean_absolute_error: 0.6746 - val_mean_squared_error: 0.6881
Epoch 175/350
mean_absolute_error: 0.4526 - mean_squared_error: 0.3224 - val_loss: 0.2447 -
val_mean_absolute_error: 0.3711 - val_mean_squared_error: 0.2447
Epoch 176/350
250/250 [============ ] - 1s 3ms/step - loss: 0.3145 -
mean_absolute_error: 0.4462 - mean_squared_error: 0.3145 - val_loss: 0.3081 -
val_mean_absolute_error: 0.4766 - val_mean_squared_error: 0.3081
Epoch 177/350
mean_absolute_error: 0.4424 - mean_squared_error: 0.3094 - val_loss: 0.3432 -
val_mean_absolute_error: 0.5139 - val_mean_squared_error: 0.3432
Epoch 178/350
mean_absolute_error: 0.4494 - mean_squared_error: 0.3150 - val_loss: 0.2402 -
val_mean_absolute_error: 0.4139 - val_mean_squared_error: 0.2402
Epoch 179/350
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mean_absolute_error: 0.4388 - mean_squared_error: 0.3073 - val_loss: 0.3543 -
val_mean_absolute_error: 0.5110 - val_mean_squared_error: 0.3543
Epoch 180/350
mean_absolute_error: 0.4407 - mean_squared_error: 0.3048 - val_loss: 0.2938 -
val_mean_absolute_error: 0.4554 - val_mean_squared_error: 0.2938
Epoch 181/350
mean_absolute_error: 0.4469 - mean_squared_error: 0.3113 - val_loss: 0.2732 -
val_mean_absolute_error: 0.4519 - val_mean_squared_error: 0.2732
Epoch 182/350
mean_absolute_error: 0.4235 - mean_squared_error: 0.2857 - val_loss: 0.2033 -
val_mean_absolute_error: 0.3676 - val_mean_squared_error: 0.2033
Epoch 183/350
250/250 [============ ] - 1s 4ms/step - loss: 0.2878 -
mean_absolute_error: 0.4284 - mean_squared_error: 0.2878 - val_loss: 0.2495 -
val_mean_absolute_error: 0.4199 - val_mean_squared_error: 0.2495
Epoch 184/350
mean_absolute_error: 0.4307 - mean_squared_error: 0.2930 - val_loss: 0.3249 -
val_mean_absolute_error: 0.4924 - val_mean_squared_error: 0.3249
Epoch 185/350
mean_absolute_error: 0.4210 - mean_squared_error: 0.2806 - val_loss: 0.4564 -
val_mean_absolute_error: 0.5852 - val_mean_squared_error: 0.4564
Epoch 186/350
mean_absolute_error: 0.4255 - mean_squared_error: 0.2854 - val_loss: 0.3296 -
val_mean_absolute_error: 0.4538 - val_mean_squared_error: 0.3296
Epoch 187/350
mean_absolute_error: 0.4184 - mean_squared_error: 0.2749 - val_loss: 0.2299 -
val_mean_absolute_error: 0.3597 - val_mean_squared_error: 0.2299
Epoch 188/350
250/250 [============ ] - 1s 3ms/step - loss: 0.2629 -
mean_absolute_error: 0.4057 - mean_squared_error: 0.2629 - val_loss: 0.1608 -
val_mean_absolute_error: 0.3315 - val_mean_squared_error: 0.1608
Epoch 189/350
mean_absolute_error: 0.4218 - mean_squared_error: 0.2780 - val_loss: 0.1715 -
val_mean_absolute_error: 0.3387 - val_mean_squared_error: 0.1715
Epoch 190/350
mean_absolute_error: 0.4081 - mean_squared_error: 0.2636 - val_loss: 0.1988 -
val_mean_absolute_error: 0.3277 - val_mean_squared_error: 0.1988
Epoch 191/350
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mean_absolute_error: 0.4162 - mean_squared_error: 0.2715 - val_loss: 0.6950 -
val mean absolute error: 0.7179 - val mean squared error: 0.6950
Epoch 192/350
mean_absolute_error: 0.4114 - mean_squared_error: 0.2684 - val_loss: 0.6306 -
val_mean_absolute_error: 0.6833 - val_mean_squared_error: 0.6306
Epoch 193/350
mean_absolute_error: 0.4048 - mean_squared_error: 0.2611 - val_loss: 0.3486 -
val_mean_absolute_error: 0.4886 - val_mean_squared_error: 0.3486
Epoch 194/350
mean_absolute_error: 0.4003 - mean_squared_error: 0.2530 - val_loss: 0.2427 -
val_mean_absolute_error: 0.4205 - val_mean_squared_error: 0.2427
Epoch 195/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.2591 -
mean_absolute_error: 0.4053 - mean_squared_error: 0.2591 - val_loss: 0.4544 -
val_mean_absolute_error: 0.5458 - val_mean_squared_error: 0.4544
Epoch 196/350
mean_absolute_error: 0.3948 - mean_squared_error: 0.2474 - val_loss: 0.6565 -
val_mean_absolute_error: 0.6710 - val_mean_squared_error: 0.6565
Epoch 197/350
mean_absolute_error: 0.3962 - mean_squared_error: 0.2478 - val_loss: 0.2508 -
val_mean_absolute_error: 0.4280 - val_mean_squared_error: 0.2508
Epoch 198/350
mean_absolute_error: 0.3819 - mean_squared_error: 0.2347 - val_loss: 0.2688 -
val_mean_absolute_error: 0.3965 - val_mean_squared_error: 0.2688
Epoch 199/350
mean_absolute_error: 0.3915 - mean_squared_error: 0.2428 - val_loss: 0.4642 -
val_mean_absolute_error: 0.5827 - val_mean_squared_error: 0.4642
Epoch 200/350
250/250 [============ ] - 1s 3ms/step - loss: 0.2444 -
mean_absolute_error: 0.3945 - mean_squared_error: 0.2444 - val_loss: 0.2132 -
val_mean_absolute_error: 0.3511 - val_mean_squared_error: 0.2132
Epoch 201/350
mean_absolute_error: 0.3695 - mean_squared_error: 0.2192 - val_loss: 0.2057 -
val_mean_absolute_error: 0.3553 - val_mean_squared_error: 0.2057
Epoch 202/350
mean_absolute_error: 0.3938 - mean_squared_error: 0.2466 - val_loss: 0.1688 -
val_mean_absolute_error: 0.3393 - val_mean_squared_error: 0.1688
Epoch 203/350
```

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mean_absolute_error: 0.3672 - mean_squared_error: 0.2188 - val_loss: 0.2764 -
val_mean_absolute_error: 0.4036 - val_mean_squared_error: 0.2764
Epoch 204/350
mean_absolute_error: 0.3831 - mean_squared_error: 0.2366 - val_loss: 0.1506 -
val_mean_absolute_error: 0.2813 - val_mean_squared_error: 0.1506
Epoch 205/350
mean_absolute_error: 0.3765 - mean_squared_error: 0.2293 - val_loss: 0.3052 -
val_mean_absolute_error: 0.4733 - val_mean_squared_error: 0.3052
Epoch 206/350
mean_absolute_error: 0.3657 - mean_squared_error: 0.2150 - val_loss: 0.1251 -
val_mean_absolute_error: 0.2746 - val_mean_squared_error: 0.1251
Epoch 207/350
250/250 [=========== ] - 1s 3ms/step - loss: 0.2255 -
mean_absolute_error: 0.3783 - mean_squared_error: 0.2255 - val_loss: 0.2624 -
val_mean_absolute_error: 0.4477 - val_mean_squared_error: 0.2624
Epoch 208/350
mean_absolute_error: 0.3544 - mean_squared_error: 0.2031 - val_loss: 0.1129 -
val_mean_absolute_error: 0.2405 - val_mean_squared_error: 0.1129
Epoch 209/350
mean_absolute_error: 0.3676 - mean_squared_error: 0.2144 - val_loss: 0.1171 -
val_mean_absolute_error: 0.2645 - val_mean_squared_error: 0.1171
Epoch 210/350
mean_absolute_error: 0.3663 - mean_squared_error: 0.2143 - val_loss: 0.1272 -
val_mean_absolute_error: 0.2811 - val_mean_squared_error: 0.1272
Epoch 211/350
250/250 [============= ] - 1s 3ms/step - loss: 0.2115 -
mean_absolute_error: 0.3603 - mean_squared_error: 0.2115 - val_loss: 0.1679 -
val_mean_absolute_error: 0.3340 - val_mean_squared_error: 0.1679
Epoch 212/350
250/250 [============ ] - 1s 4ms/step - loss: 0.1986 -
mean_absolute_error: 0.3478 - mean_squared_error: 0.1986 - val_loss: 0.1036 -
val_mean_absolute_error: 0.2482 - val_mean_squared_error: 0.1036
Epoch 213/350
mean_absolute_error: 0.3494 - mean_squared_error: 0.2013 - val_loss: 0.0946 -
val_mean_absolute_error: 0.2228 - val_mean_squared_error: 0.0946
Epoch 214/350
mean_absolute_error: 0.3384 - mean_squared_error: 0.1866 - val_loss: 0.2614 -
val_mean_absolute_error: 0.4416 - val_mean_squared_error: 0.2614
Epoch 215/350
```

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mean_absolute_error: 0.3554 - mean_squared_error: 0.2027 - val_loss: 0.2304 -
val mean absolute error: 0.4191 - val mean squared error: 0.2304
Epoch 216/350
mean_absolute_error: 0.3563 - mean_squared_error: 0.2015 - val_loss: 0.1266 -
val_mean_absolute_error: 0.2846 - val_mean_squared_error: 0.1266
Epoch 217/350
mean_absolute_error: 0.3394 - mean_squared_error: 0.1844 - val_loss: 0.1492 -
val_mean_absolute_error: 0.2868 - val_mean_squared_error: 0.1492
Epoch 218/350
mean_absolute_error: 0.3486 - mean_squared_error: 0.1936 - val_loss: 0.1292 -
val_mean_absolute_error: 0.2966 - val_mean_squared_error: 0.1292
Epoch 219/350
250/250 [=========== ] - 1s 3ms/step - loss: 0.1799 -
mean_absolute_error: 0.3343 - mean_squared_error: 0.1799 - val_loss: 0.1445 -
val_mean_absolute_error: 0.3117 - val_mean_squared_error: 0.1445
Epoch 220/350
mean_absolute_error: 0.3409 - mean_squared_error: 0.1876 - val_loss: 0.1081 -
val_mean_absolute_error: 0.2418 - val_mean_squared_error: 0.1081
Epoch 221/350
mean_absolute_error: 0.3288 - mean_squared_error: 0.1745 - val_loss: 0.2456 -
val_mean_absolute_error: 0.4309 - val_mean_squared_error: 0.2456
Epoch 222/350
mean_absolute_error: 0.3374 - mean_squared_error: 0.1835 - val_loss: 0.1234 -
val_mean_absolute_error: 0.2821 - val_mean_squared_error: 0.1234
Epoch 223/350
mean_absolute_error: 0.3420 - mean_squared_error: 0.1886 - val_loss: 0.0929 -
val_mean_absolute_error: 0.2225 - val_mean_squared_error: 0.0929
Epoch 224/350
250/250 [============ ] - 1s 3ms/step - loss: 0.1839 -
mean_absolute_error: 0.3397 - mean_squared_error: 0.1839 - val_loss: 0.6567 -
val_mean_absolute_error: 0.6478 - val_mean_squared_error: 0.6567
Epoch 225/350
mean_absolute_error: 0.3166 - mean_squared_error: 0.1660 - val_loss: 0.2630 -
val_mean_absolute_error: 0.4163 - val_mean_squared_error: 0.2630
Epoch 226/350
mean_absolute_error: 0.3226 - mean_squared_error: 0.1699 - val_loss: 0.3183 -
val_mean_absolute_error: 0.4599 - val_mean_squared_error: 0.3183
Epoch 227/350
```

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mean_absolute_error: 0.3339 - mean_squared_error: 0.1776 - val_loss: 0.1970 -
val mean absolute error: 0.3407 - val mean squared error: 0.1970
Epoch 228/350
mean_absolute_error: 0.3101 - mean_squared_error: 0.1576 - val_loss: 0.0752 -
val_mean_absolute_error: 0.1966 - val_mean_squared_error: 0.0752
Epoch 229/350
mean_absolute_error: 0.3021 - mean_squared_error: 0.1503 - val_loss: 0.1437 -
val_mean_absolute_error: 0.2878 - val_mean_squared_error: 0.1437
Epoch 230/350
mean_absolute_error: 0.3181 - mean_squared_error: 0.1674 - val_loss: 0.2741 -
val_mean_absolute_error: 0.4117 - val_mean_squared_error: 0.2741
Epoch 231/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.1662 -
mean_absolute_error: 0.3210 - mean_squared_error: 0.1662 - val_loss: 0.1291 -
val_mean_absolute_error: 0.2759 - val_mean_squared_error: 0.1291
Epoch 232/350
mean_absolute_error: 0.3106 - mean_squared_error: 0.1585 - val_loss: 0.2100 -
val_mean_absolute_error: 0.3970 - val_mean_squared_error: 0.2100
Epoch 233/350
mean_absolute_error: 0.3086 - mean_squared_error: 0.1559 - val_loss: 0.1185 -
val_mean_absolute_error: 0.2861 - val_mean_squared_error: 0.1185
Epoch 234/350
mean_absolute_error: 0.2963 - mean_squared_error: 0.1452 - val_loss: 0.2507 -
val_mean_absolute_error: 0.4251 - val_mean_squared_error: 0.2507
Epoch 235/350
mean_absolute_error: 0.3175 - mean_squared_error: 0.1641 - val_loss: 0.1751 -
val_mean_absolute_error: 0.3228 - val_mean_squared_error: 0.1751
Epoch 236/350
250/250 [============ ] - 1s 4ms/step - loss: 0.1371 -
mean_absolute_error: 0.2816 - mean_squared_error: 0.1371 - val_loss: 0.1149 -
val_mean_absolute_error: 0.2738 - val_mean_squared_error: 0.1149
Epoch 237/350
mean_absolute_error: 0.3190 - mean_squared_error: 0.1689 - val_loss: 0.0931 -
val_mean_absolute_error: 0.2411 - val_mean_squared_error: 0.0931
Epoch 238/350
mean_absolute_error: 0.2898 - mean_squared_error: 0.1386 - val_loss: 0.1879 -
val_mean_absolute_error: 0.3274 - val_mean_squared_error: 0.1879
Epoch 239/350
```

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mean_absolute_error: 0.3032 - mean_squared_error: 0.1527 - val_loss: 0.2227 -
val mean absolute error: 0.3503 - val mean squared error: 0.2227
Epoch 240/350
mean_absolute_error: 0.3240 - mean_squared_error: 0.1657 - val_loss: 0.1198 -
val_mean_absolute_error: 0.2975 - val_mean_squared_error: 0.1198
Epoch 241/350
mean_absolute_error: 0.2877 - mean_squared_error: 0.1364 - val_loss: 0.0950 -
val_mean_absolute_error: 0.2388 - val_mean_squared_error: 0.0950
Epoch 242/350
mean_absolute_error: 0.2996 - mean_squared_error: 0.1492 - val_loss: 0.1544 -
val_mean_absolute_error: 0.3277 - val_mean_squared_error: 0.1544
Epoch 243/350
250/250 [============ ] - 1s 4ms/step - loss: 0.1275 -
mean_absolute_error: 0.2746 - mean_squared_error: 0.1275 - val_loss: 0.1956 -
val_mean_absolute_error: 0.3775 - val_mean_squared_error: 0.1956
Epoch 244/350
mean_absolute_error: 0.2942 - mean_squared_error: 0.1440 - val_loss: 0.1743 -
val_mean_absolute_error: 0.3237 - val_mean_squared_error: 0.1743
Epoch 245/350
mean_absolute_error: 0.2921 - mean_squared_error: 0.1404 - val_loss: 0.1398 -
val_mean_absolute_error: 0.2925 - val_mean_squared_error: 0.1398
Epoch 246/350
mean_absolute_error: 0.3030 - mean_squared_error: 0.1477 - val_loss: 0.1892 -
val_mean_absolute_error: 0.3752 - val_mean_squared_error: 0.1892
Epoch 247/350
mean_absolute_error: 0.2861 - mean_squared_error: 0.1360 - val_loss: 0.1938 -
val_mean_absolute_error: 0.3597 - val_mean_squared_error: 0.1938
Epoch 248/350
250/250 [============ ] - 1s 4ms/step - loss: 0.1295 -
mean_absolute_error: 0.2816 - mean_squared_error: 0.1295 - val_loss: 0.1080 -
val_mean_absolute_error: 0.2564 - val_mean_squared_error: 0.1080
Epoch 249/350
mean_absolute_error: 0.2707 - mean_squared_error: 0.1242 - val_loss: 0.1255 -
val_mean_absolute_error: 0.2954 - val_mean_squared_error: 0.1255
Epoch 250/350
mean_absolute_error: 0.2845 - mean_squared_error: 0.1366 - val_loss: 0.1284 -
val_mean_absolute_error: 0.3004 - val_mean_squared_error: 0.1284
Epoch 251/350
```

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mean_absolute_error: 0.2777 - mean_squared_error: 0.1276 - val_loss: 0.1500 -
val mean absolute error: 0.3047 - val mean squared error: 0.1500
Epoch 252/350
mean_absolute_error: 0.2703 - mean_squared_error: 0.1231 - val_loss: 0.3724 -
val_mean_absolute_error: 0.5070 - val_mean_squared_error: 0.3724
Epoch 253/350
mean_absolute_error: 0.2958 - mean_squared_error: 0.1426 - val_loss: 0.0859 -
val mean absolute error: 0.2141 - val mean squared error: 0.0859
Epoch 254/350
mean_absolute_error: 0.2623 - mean_squared_error: 0.1195 - val_loss: 0.1233 -
val_mean_absolute_error: 0.2849 - val_mean_squared_error: 0.1233
Epoch 255/350
250/250 [============ ] - 1s 4ms/step - loss: 0.1272 -
mean_absolute_error: 0.2796 - mean_squared_error: 0.1272 - val_loss: 0.2445 -
val_mean_absolute_error: 0.4005 - val_mean_squared_error: 0.2445
Epoch 256/350
mean_absolute_error: 0.2685 - mean_squared_error: 0.1191 - val_loss: 0.0531 -
val_mean_absolute_error: 0.1772 - val_mean_squared_error: 0.0531
Epoch 257/350
mean_absolute_error: 0.2518 - mean_squared_error: 0.1115 - val_loss: 0.2598 -
val_mean_absolute_error: 0.4370 - val_mean_squared_error: 0.2598
Epoch 258/350
mean_absolute_error: 0.2588 - mean_squared_error: 0.1145 - val_loss: 0.1184 -
val_mean_absolute_error: 0.2665 - val_mean_squared_error: 0.1184
Epoch 259/350
mean_absolute_error: 0.2583 - mean_squared_error: 0.1121 - val_loss: 0.0603 -
val_mean_absolute_error: 0.1780 - val_mean_squared_error: 0.0603
Epoch 260/350
mean_absolute_error: 0.2691 - mean_squared_error: 0.1179 - val_loss: 0.1802 -
val_mean_absolute_error: 0.3401 - val_mean_squared_error: 0.1802
Epoch 261/350
mean_absolute_error: 0.2612 - mean_squared_error: 0.1146 - val_loss: 0.0658 -
val_mean_absolute_error: 0.1842 - val_mean_squared_error: 0.0658
Epoch 262/350
mean_absolute_error: 0.2554 - mean_squared_error: 0.1111 - val_loss: 0.0551 -
val_mean_absolute_error: 0.1650 - val_mean_squared_error: 0.0551
Epoch 263/350
```

```
mean_absolute_error: 0.2513 - mean_squared_error: 0.1069 - val_loss: 0.0438 -
val_mean_absolute_error: 0.1608 - val_mean_squared_error: 0.0438
Epoch 264/350
mean_absolute_error: 0.2428 - mean_squared_error: 0.1005 - val_loss: 0.0420 -
val_mean_absolute_error: 0.1470 - val_mean_squared_error: 0.0420
Epoch 265/350
mean_absolute_error: 0.2604 - mean_squared_error: 0.1141 - val_loss: 0.0599 -
val_mean_absolute_error: 0.1783 - val_mean_squared_error: 0.0599
Epoch 266/350
mean_absolute_error: 0.2551 - mean_squared_error: 0.1084 - val_loss: 0.2232 -
val_mean_absolute_error: 0.4033 - val_mean_squared_error: 0.2232
Epoch 267/350
250/250 [=========== ] - 1s 3ms/step - loss: 0.1120 -
mean_absolute_error: 0.2551 - mean_squared_error: 0.1120 - val_loss: 0.2587 -
val_mean_absolute_error: 0.4150 - val_mean_squared_error: 0.2587
Epoch 268/350
mean_absolute_error: 0.2585 - mean_squared_error: 0.1105 - val_loss: 0.0465 -
val_mean_absolute_error: 0.1636 - val_mean_squared_error: 0.0465
Epoch 269/350
mean_absolute_error: 0.2385 - mean_squared_error: 0.0972 - val_loss: 0.0545 -
val_mean_absolute_error: 0.1798 - val_mean_squared_error: 0.0545
Epoch 270/350
mean_absolute_error: 0.2542 - mean_squared_error: 0.1063 - val_loss: 0.1470 -
val_mean_absolute_error: 0.3173 - val_mean_squared_error: 0.1470
Epoch 271/350
mean_absolute_error: 0.2203 - mean_squared_error: 0.0872 - val_loss: 0.0498 -
val_mean_absolute_error: 0.1760 - val_mean_squared_error: 0.0498
Epoch 272/350
250/250 [============= ] - 1s 4ms/step - loss: 0.0959 -
mean_absolute_error: 0.2379 - mean_squared_error: 0.0959 - val_loss: 0.0563 -
val_mean_absolute_error: 0.1875 - val_mean_squared_error: 0.0563
Epoch 273/350
mean_absolute_error: 0.2562 - mean_squared_error: 0.1098 - val_loss: 0.0569 -
val_mean_absolute_error: 0.1775 - val_mean_squared_error: 0.0569
Epoch 274/350
mean_absolute_error: 0.2203 - mean_squared_error: 0.0865 - val_loss: 0.0425 -
val_mean_absolute_error: 0.1590 - val_mean_squared_error: 0.0425
Epoch 275/350
```

```
mean_absolute_error: 0.2313 - mean_squared_error: 0.0912 - val_loss: 0.2814 -
val mean absolute error: 0.4481 - val mean squared error: 0.2814
Epoch 276/350
mean_absolute_error: 0.2491 - mean_squared_error: 0.1050 - val_loss: 0.0433 -
val_mean_absolute_error: 0.1594 - val_mean_squared_error: 0.0433
Epoch 277/350
mean_absolute_error: 0.2550 - mean_squared_error: 0.1096 - val_loss: 0.0581 -
val_mean_absolute_error: 0.2016 - val_mean_squared_error: 0.0581
Epoch 278/350
mean_absolute_error: 0.2459 - mean_squared_error: 0.0974 - val_loss: 0.3449 -
val_mean_absolute_error: 0.4787 - val_mean_squared_error: 0.3449
Epoch 279/350
250/250 [=========== ] - 1s 3ms/step - loss: 0.0896 -
mean_absolute_error: 0.2304 - mean_squared_error: 0.0896 - val_loss: 0.2124 -
val_mean_absolute_error: 0.3703 - val_mean_squared_error: 0.2124
Epoch 280/350
mean_absolute_error: 0.2378 - mean_squared_error: 0.0963 - val_loss: 0.0453 -
val_mean_absolute_error: 0.1685 - val_mean_squared_error: 0.0453
Epoch 281/350
mean_absolute_error: 0.2320 - mean_squared_error: 0.0904 - val_loss: 0.1115 -
val_mean_absolute_error: 0.2647 - val_mean_squared_error: 0.1115
Epoch 282/350
mean_absolute_error: 0.2239 - mean_squared_error: 0.0845 - val_loss: 0.0472 -
val_mean_absolute_error: 0.1593 - val_mean_squared_error: 0.0472
Epoch 283/350
mean_absolute_error: 0.2401 - mean_squared_error: 0.0960 - val_loss: 0.1075 -
val_mean_absolute_error: 0.2655 - val_mean_squared_error: 0.1075
Epoch 284/350
250/250 [============= ] - 1s 4ms/step - loss: 0.0760 -
mean_absolute_error: 0.2077 - mean_squared_error: 0.0760 - val_loss: 0.0431 -
val_mean_absolute_error: 0.1548 - val_mean_squared_error: 0.0431
Epoch 285/350
mean_absolute_error: 0.2458 - mean_squared_error: 0.1033 - val_loss: 0.0869 -
val_mean_absolute_error: 0.2470 - val_mean_squared_error: 0.0869
Epoch 286/350
mean_absolute_error: 0.2298 - mean_squared_error: 0.0888 - val_loss: 0.0880 -
val_mean_absolute_error: 0.2300 - val_mean_squared_error: 0.0880
Epoch 287/350
```

```
mean_absolute_error: 0.2449 - mean_squared_error: 0.0962 - val_loss: 0.1346 -
val mean absolute error: 0.2971 - val mean squared error: 0.1346
Epoch 288/350
mean_absolute_error: 0.2237 - mean_squared_error: 0.0863 - val_loss: 0.0297 -
val_mean_absolute_error: 0.1250 - val_mean_squared_error: 0.0297
Epoch 289/350
mean_absolute_error: 0.2406 - mean_squared_error: 0.0952 - val_loss: 0.0640 -
val_mean_absolute_error: 0.2028 - val_mean_squared_error: 0.0640
Epoch 290/350
mean_absolute_error: 0.2082 - mean_squared_error: 0.0748 - val_loss: 0.2313 -
val_mean_absolute_error: 0.4067 - val_mean_squared_error: 0.2313
Epoch 291/350
250/250 [=========== ] - 1s 3ms/step - loss: 0.0815 -
mean_absolute_error: 0.2175 - mean_squared_error: 0.0815 - val_loss: 0.0373 -
val_mean_absolute_error: 0.1503 - val_mean_squared_error: 0.0373
Epoch 292/350
mean_absolute_error: 0.2222 - mean_squared_error: 0.0865 - val_loss: 0.0794 -
val_mean_absolute_error: 0.2420 - val_mean_squared_error: 0.0794
Epoch 293/350
mean_absolute_error: 0.2010 - mean_squared_error: 0.0697 - val_loss: 0.0507 -
val_mean_absolute_error: 0.1675 - val_mean_squared_error: 0.0507
Epoch 294/350
mean_absolute_error: 0.2252 - mean_squared_error: 0.0850 - val_loss: 0.1937 -
val_mean_absolute_error: 0.3524 - val_mean_squared_error: 0.1937
Epoch 295/350
250/250 [============ ] - 1s 4ms/step - loss: 0.0818 -
mean_absolute_error: 0.2143 - mean_squared_error: 0.0818 - val_loss: 0.0340 -
val_mean_absolute_error: 0.1339 - val_mean_squared_error: 0.0340
Epoch 296/350
250/250 [============ ] - 1s 3ms/step - loss: 0.0686 -
mean_absolute_error: 0.1984 - mean_squared_error: 0.0686 - val_loss: 0.0841 -
val_mean_absolute_error: 0.2282 - val_mean_squared_error: 0.0841
Epoch 297/350
mean_absolute_error: 0.2269 - mean_squared_error: 0.0869 - val_loss: 0.0326 -
val_mean_absolute_error: 0.1351 - val_mean_squared_error: 0.0326
Epoch 298/350
mean_absolute_error: 0.1942 - mean_squared_error: 0.0664 - val_loss: 0.1131 -
val_mean_absolute_error: 0.2804 - val_mean_squared_error: 0.1131
Epoch 299/350
```

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mean_absolute_error: 0.2060 - mean_squared_error: 0.0744 - val_loss: 0.0913 -
val mean absolute error: 0.2479 - val mean squared error: 0.0913
Epoch 300/350
mean_absolute_error: 0.2267 - mean_squared_error: 0.0840 - val_loss: 0.1185 -
val_mean_absolute_error: 0.2674 - val_mean_squared_error: 0.1185
Epoch 301/350
mean_absolute_error: 0.2147 - mean_squared_error: 0.0795 - val_loss: 0.1355 -
val_mean_absolute_error: 0.3047 - val_mean_squared_error: 0.1355
Epoch 302/350
mean_absolute_error: 0.2041 - mean_squared_error: 0.0708 - val_loss: 0.0266 -
val_mean_absolute_error: 0.1179 - val_mean_squared_error: 0.0266
Epoch 303/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.0821 -
mean_absolute_error: 0.2208 - mean_squared_error: 0.0821 - val_loss: 0.1060 -
val_mean_absolute_error: 0.2637 - val_mean_squared_error: 0.1060
Epoch 304/350
mean_absolute_error: 0.2265 - mean_squared_error: 0.0844 - val_loss: 0.1155 -
val_mean_absolute_error: 0.2624 - val_mean_squared_error: 0.1155
Epoch 305/350
mean_absolute_error: 0.2117 - mean_squared_error: 0.0783 - val_loss: 0.0381 -
val_mean_absolute_error: 0.1559 - val_mean_squared_error: 0.0381
Epoch 306/350
mean_absolute_error: 0.1956 - mean_squared_error: 0.0686 - val_loss: 0.0485 -
val_mean_absolute_error: 0.1643 - val_mean_squared_error: 0.0485
Epoch 307/350
mean_absolute_error: 0.1906 - mean_squared_error: 0.0638 - val_loss: 0.1348 -
val_mean_absolute_error: 0.3063 - val_mean_squared_error: 0.1348
Epoch 308/350
250/250 [============= ] - 1s 3ms/step - loss: 0.0734 -
mean_absolute_error: 0.2036 - mean_squared_error: 0.0734 - val_loss: 0.0339 -
val_mean_absolute_error: 0.1335 - val_mean_squared_error: 0.0339
Epoch 309/350
mean_absolute_error: 0.2188 - mean_squared_error: 0.0814 - val_loss: 0.0358 -
val_mean_absolute_error: 0.1490 - val_mean_squared_error: 0.0358
Epoch 310/350
mean_absolute_error: 0.1935 - mean_squared_error: 0.0672 - val_loss: 0.2091 -
val_mean_absolute_error: 0.3668 - val_mean_squared_error: 0.2091
Epoch 311/350
```

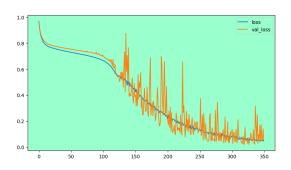
```
mean_absolute_error: 0.1783 - mean_squared_error: 0.0578 - val_loss: 0.0829 -
val_mean_absolute_error: 0.2214 - val_mean_squared_error: 0.0829
Epoch 312/350
mean_absolute_error: 0.1807 - mean_squared_error: 0.0596 - val_loss: 0.0738 -
val_mean_absolute_error: 0.2237 - val_mean_squared_error: 0.0738
Epoch 313/350
mean_absolute_error: 0.1997 - mean_squared_error: 0.0699 - val_loss: 0.0767 -
val_mean_absolute_error: 0.2154 - val_mean_squared_error: 0.0767
Epoch 314/350
mean_absolute_error: 0.1943 - mean_squared_error: 0.0645 - val_loss: 0.0706 -
val_mean_absolute_error: 0.1975 - val_mean_squared_error: 0.0706
Epoch 315/350
250/250 [============ ] - 1s 3ms/step - loss: 0.0595 -
mean_absolute_error: 0.1797 - mean_squared_error: 0.0595 - val_loss: 0.0390 -
val_mean_absolute_error: 0.1493 - val_mean_squared_error: 0.0390
Epoch 316/350
mean_absolute_error: 0.1970 - mean_squared_error: 0.0722 - val_loss: 0.0278 -
val_mean_absolute_error: 0.1182 - val_mean_squared_error: 0.0278
Epoch 317/350
mean_absolute_error: 0.1921 - mean_squared_error: 0.0653 - val_loss: 0.0279 -
val_mean_absolute_error: 0.1203 - val_mean_squared_error: 0.0279
Epoch 318/350
mean_absolute_error: 0.1774 - mean_squared_error: 0.0548 - val_loss: 0.0829 -
val_mean_absolute_error: 0.2287 - val_mean_squared_error: 0.0829
Epoch 319/350
mean_absolute_error: 0.1922 - mean_squared_error: 0.0655 - val_loss: 0.0218 -
val_mean_absolute_error: 0.1134 - val_mean_squared_error: 0.0218
Epoch 320/350
250/250 [============ ] - 1s 3ms/step - loss: 0.0531 -
mean_absolute_error: 0.1721 - mean_squared_error: 0.0531 - val_loss: 0.0194 -
val_mean_absolute_error: 0.1003 - val_mean_squared_error: 0.0194
Epoch 321/350
mean_absolute_error: 0.1963 - mean_squared_error: 0.0642 - val_loss: 0.1481 -
val_mean_absolute_error: 0.3179 - val_mean_squared_error: 0.1481
Epoch 322/350
mean_absolute_error: 0.1711 - mean_squared_error: 0.0530 - val_loss: 0.0561 -
val_mean_absolute_error: 0.1927 - val_mean_squared_error: 0.0561
Epoch 323/350
```

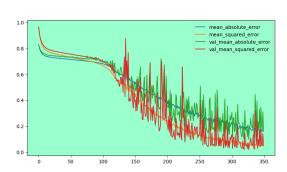
```
mean_absolute_error: 0.1921 - mean_squared_error: 0.0622 - val_loss: 0.0473 -
val_mean_absolute_error: 0.1634 - val_mean_squared_error: 0.0473
Epoch 324/350
mean_absolute_error: 0.1805 - mean_squared_error: 0.0579 - val_loss: 0.0557 -
val_mean_absolute_error: 0.1722 - val_mean_squared_error: 0.0557
Epoch 325/350
mean_absolute_error: 0.1991 - mean_squared_error: 0.0667 - val_loss: 0.0266 -
val mean absolute error: 0.1171 - val mean squared error: 0.0266
Epoch 326/350
mean_absolute_error: 0.1768 - mean_squared_error: 0.0572 - val_loss: 0.0700 -
val_mean_absolute_error: 0.2087 - val_mean_squared_error: 0.0700
Epoch 327/350
250/250 [=========== ] - 1s 4ms/step - loss: 0.0506 -
mean_absolute_error: 0.1600 - mean_squared_error: 0.0506 - val_loss: 0.0184 -
val_mean_absolute_error: 0.0985 - val_mean_squared_error: 0.0184
Epoch 328/350
mean_absolute_error: 0.1957 - mean_squared_error: 0.0661 - val_loss: 0.0515 -
val_mean_absolute_error: 0.1831 - val_mean_squared_error: 0.0515
Epoch 329/350
mean_absolute_error: 0.1835 - mean_squared_error: 0.0612 - val_loss: 0.0687 -
val_mean_absolute_error: 0.1984 - val_mean_squared_error: 0.0687
Epoch 330/350
mean_absolute_error: 0.1777 - mean_squared_error: 0.0545 - val_loss: 0.1063 -
val_mean_absolute_error: 0.2599 - val_mean_squared_error: 0.1063
Epoch 331/350
250/250 [============= ] - 1s 4ms/step - loss: 0.0527 -
mean_absolute_error: 0.1670 - mean_squared_error: 0.0527 - val_loss: 0.0688 -
val_mean_absolute_error: 0.2035 - val_mean_squared_error: 0.0688
Epoch 332/350
250/250 [============= ] - 1s 4ms/step - loss: 0.0472 -
mean_absolute_error: 0.1639 - mean_squared_error: 0.0472 - val_loss: 0.0584 -
val_mean_absolute_error: 0.1909 - val_mean_squared_error: 0.0584
Epoch 333/350
mean_absolute_error: 0.1761 - mean_squared_error: 0.0559 - val_loss: 0.0485 -
val_mean_absolute_error: 0.1820 - val_mean_squared_error: 0.0485
Epoch 334/350
mean_absolute_error: 0.1730 - mean_squared_error: 0.0554 - val_loss: 0.1315 -
val_mean_absolute_error: 0.2916 - val_mean_squared_error: 0.1315
Epoch 335/350
```

```
mean_absolute_error: 0.1743 - mean_squared_error: 0.0586 - val_loss: 0.0292 -
val_mean_absolute_error: 0.1334 - val_mean_squared_error: 0.0292
Epoch 336/350
mean_absolute_error: 0.1794 - mean_squared_error: 0.0574 - val_loss: 0.0207 -
val_mean_absolute_error: 0.1043 - val_mean_squared_error: 0.0207
Epoch 337/350
mean_absolute_error: 0.1723 - mean_squared_error: 0.0513 - val_loss: 0.3146 -
val_mean_absolute_error: 0.4416 - val_mean_squared_error: 0.3146
Epoch 338/350
mean_absolute_error: 0.1798 - mean_squared_error: 0.0557 - val_loss: 0.1674 -
val_mean_absolute_error: 0.3248 - val_mean_squared_error: 0.1674
Epoch 339/350
250/250 [============ ] - 1s 4ms/step - loss: 0.0472 -
mean_absolute_error: 0.1553 - mean_squared_error: 0.0472 - val_loss: 0.0578 -
val_mean_absolute_error: 0.1810 - val_mean_squared_error: 0.0578
Epoch 340/350
mean_absolute_error: 0.1601 - mean_squared_error: 0.0453 - val_loss: 0.2474 -
val_mean_absolute_error: 0.3908 - val_mean_squared_error: 0.2474
Epoch 341/350
mean_absolute_error: 0.1850 - mean_squared_error: 0.0584 - val_loss: 0.0276 -
val_mean_absolute_error: 0.1244 - val_mean_squared_error: 0.0276
Epoch 342/350
mean_absolute_error: 0.1780 - mean_squared_error: 0.0575 - val_loss: 0.0956 -
val_mean_absolute_error: 0.2293 - val_mean_squared_error: 0.0956
Epoch 343/350
250/250 [============= ] - 1s 4ms/step - loss: 0.0514 -
mean_absolute_error: 0.1698 - mean_squared_error: 0.0514 - val_loss: 0.0503 -
val_mean_absolute_error: 0.1794 - val_mean_squared_error: 0.0503
Epoch 344/350
250/250 [============ ] - 1s 3ms/step - loss: 0.0437 -
mean_absolute_error: 0.1538 - mean_squared_error: 0.0437 - val_loss: 0.1672 -
val_mean_absolute_error: 0.3292 - val_mean_squared_error: 0.1672
Epoch 345/350
mean_absolute_error: 0.1700 - mean_squared_error: 0.0513 - val_loss: 0.0406 -
val_mean_absolute_error: 0.1509 - val_mean_squared_error: 0.0406
Epoch 346/350
mean_absolute_error: 0.1652 - mean_squared_error: 0.0506 - val_loss: 0.1745 -
val_mean_absolute_error: 0.3466 - val_mean_squared_error: 0.1745
Epoch 347/350
```

```
mean_absolute_error: 0.1665 - mean_squared_error: 0.0487 - val_loss: 0.0540 -
     val_mean_absolute_error: 0.1936 - val_mean_squared_error: 0.0540
     Epoch 348/350
     mean_absolute_error: 0.1798 - mean_squared_error: 0.0599 - val_loss: 0.0470 -
     val_mean_absolute_error: 0.1720 - val_mean_squared_error: 0.0470
     Epoch 349/350
     mean_absolute_error: 0.1546 - mean_squared_error: 0.0444 - val_loss: 0.1405 -
     val_mean_absolute_error: 0.3086 - val_mean_squared_error: 0.1405
     Epoch 350/350
     250/250 [============= ] - 1s 3ms/step - loss: 0.0496 -
     mean_absolute_error: 0.1680 - mean_squared_error: 0.0496 - val_loss: 0.0528 -
     val_mean_absolute_error: 0.1761 - val_mean_squared_error: 0.0528
[104]: fig,axs=plt.subplots(1,2,figsize=(20,5))
     axs[0].plot(his['loss'])
     axs[0].plot(his['val_loss'])
     axs[0].legend(['loss','val_loss'])
     axs[1].plot(his['mean_absolute_error'])
     axs[1].plot(his['mean_squared_error'])
     axs[1].plot(his['val mean absolute error'])
     axs[1].plot(his['val_mean_squared_error'])
      →legend(['mean_absolute_error','mean_squared_error','val_mean_absolute_error','val_mean_squa
```

[104]: <matplotlib.legend.Legend at 0x7c5e8923f190>





```
[105]: ypred=regNN.predict(xtest)
  print(f"MSE: {mean_squared_error(ytest,ypred)}")
  print(f"MAE: {mean_absolute_error(ytest,ypred)}")
  print(f"R^2: {round(r2_score(ytest,ypred)*100,2)}%")
```

19 GAMMA REGRESSOR

```
[187]: from sklearn.linear_model import ARDRegression
[190]: a=ARDRegression()
       gg={'alpha_1':[1e-3,1e-5,1e-7,1e-9],
          'alpha_2':[1e-3,1e-5,1e-7],
          'lambda_1':[1e-1,1e-3,1e-5,1e-7],
          'n_iter': [100,200,300],
          'lambda_2':[1e-3,1e-5,1e-7,1e-9],
          'tol': [1e-3,1e-5,1e-7,1e-9]}
       p(gg,a,'ardreg')
      ARDRegression()
      Best Parameters {'tol': 1e-07, 'n_iter': 300, 'lambda_2': 1e-07, 'lambda_1':
      1e-07, 'alpha_2': 0.001, 'alpha_1': 0.001}
      Best Score 22.66 %
[190]:
                        Metric_Name
                                     Values
                                      23.160
       0
                            R.sq(\%)
       1
                 Mean-Squared-Error
                                       0.756
       2
         Rooted-Mean-Squared-Error
                                       0.850
       3
               Mean-Absoulute-Error
                                       0.723
       4
              Median-Absoulte-Error
                                       0.666
```

20 CONCULSION:

ML MODEL FOR RECOMENDATION BASED ON ITS PERFORMANCES

- Adaptive Boosting Regressor
- Hist Gradient Boost Regressor
- Bagging Regressor
- Radius Neighbors Regressor
- Artificial Neural Network Multiple Linear Regression Model
- We need more data and also diversified features which can able to capture the hidden patterns and trends among the customer behaviours purchasing in the store.