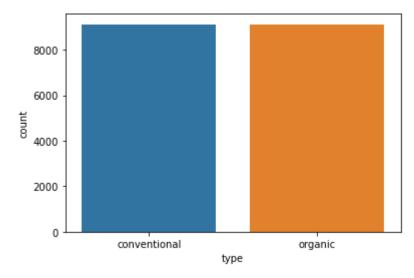
```
import pandas as pd
In [3]:
          import numpy as np
          import seaborn as sns
          import matplotlib.pyplot as plt
          %matplotlib inline
          df = pd.read csv('avocado.csv')
In [4]:
In [5]:
          df.head()
Out[5]:
            Unnamed:
                                               Total
                                                                                 Total
                                                                                         Small
                                                                                                Large XLa
                                                       4046
                                                                 4225
                                                                         4770
                       Date AveragePrice
                    0
                                            Volume
                                                                                 Bags
                                                                                         Bags
                                                                                                Bags
                                                                                                        В
                       2015-
         0
                    0
                                     1.33
                                            64236.62
                                                     1036.74
                                                              54454.85
                                                                        48.16
                                                                               8696.87
                                                                                       8603.62
                                                                                                93.25
                       12-27
                       2015-
                                            54876.98
                                                                               9505.56
                                                                                       9408.07
                                                                                                97.49
         1
                    1
                                     1.35
                                                      674.28
                                                              44638.81
                                                                         58.33
                       12-20
                       2015-
         2
                    2
                                     0.93
                                           118220.22
                                                      794.70
                                                             109149.67
                                                                       130.50 8145.35
                                                                                       8042.21
                                                                                               103.14
                       12-13
                       2015-
         3
                    3
                                     1.08
                                            78992.15
                                                    1132.00
                                                              71976.41
                                                                        72.58 5811.16 5677.40 133.76
                       12-06
                       2015-
                                     1.28
                                            51039.60
                                                      941.48
                                                              43838.39
                                                                        75.78 6183.95 5986.26 197.69
                       11-29
In [6]:
          df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 18249 entries, 0 to 18248
         Data columns (total 14 columns):
          #
              Column
                              Non-Null Count
                                               Dtvpe
          0
              Unnamed: 0
                              18249 non-null int64
                                               object
          1
              Date
                              18249 non-null
          2
              AveragePrice
                             18249 non-null
                                               float64
          3
              Total Volume
                             18249 non-null
                                               float64
                              18249 non-null
          4
              4046
                                               float64
          5
              4225
                              18249 non-null
                                               float64
          6
              4770
                              18249 non-null
                                               float64
          7
              Total Bags
                              18249 non-null
                                              float64
          8
              Small Bags
                              18249 non-null float64
          9
              Large Bags
                              18249 non-null float64
          10
              XLarge Bags
                              18249 non-null
                                              float64
          11
              type
                              18249 non-null
                                               object
          12
              year
                              18249 non-null
                                               int64
              region
                              18249 non-null
                                               object
         dtypes: float64(9), int64(2), object(3)
         memory usage: 1.9+ MB
In [7]:
          df.columns
Out[7]: Index(['Unnamed: 0', 'Date', 'AveragePrice', 'Total Volume', '4046', '4225',
                 '4770', 'Total Bags', 'Small Bags', 'Large Bags', 'XLarge Bags', 'type', 'year', 'region'],
               dtype='object')
```

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```
In [8]: sns.countplot(x='type',data=df)
```

```
Out[8]: <AxesSubplot:xlabel='type', ylabel='count'>
```



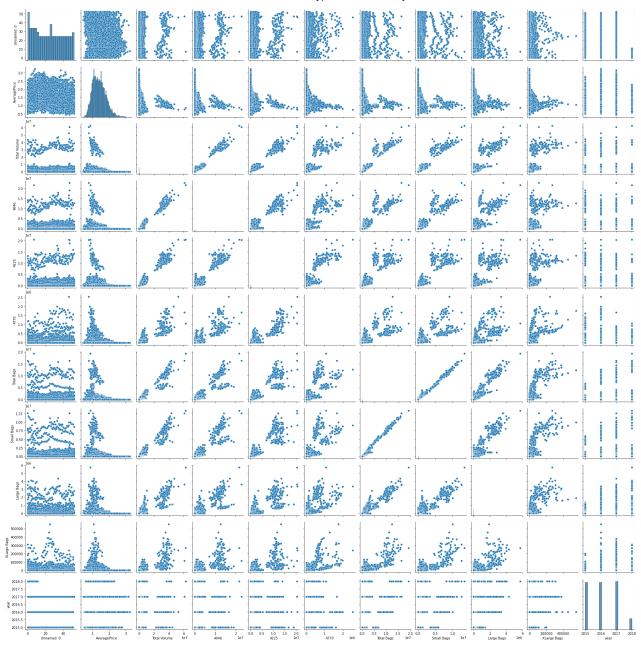
```
In [9]: df['type'].value_counts()
```

Out[9]: conventional 9126 organic 9123 Name: type, dtype: int64

```
In [10]: sns.pairplot(df)
```

Out[10]: <seaborn.axisgrid.PairGrid at 0x287594d3d90>

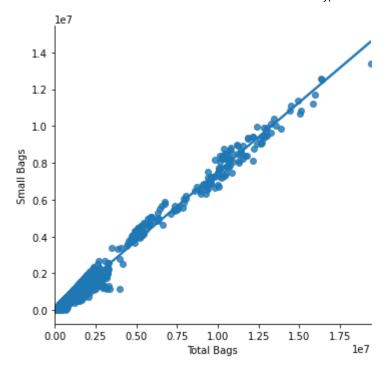
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In [11]: sns.lmplot(x='Total Bags',y='Small Bags',data=df)

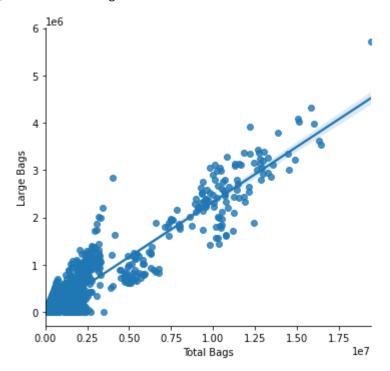
Out[11]: <seaborn.axisgrid.FacetGrid at 0x2871cea0a90>

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In [12]: sns.lmplot(x='Total Bags',y='Large Bags',data=df)

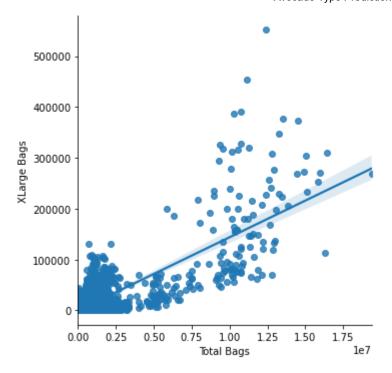
Out[12]: <seaborn.axisgrid.FacetGrid at 0x2875eacc610>



In [13]: sns.lmplot(x='Total Bags',y='XLarge Bags',data=df)

Out[13]: <seaborn.axisgrid.FacetGrid at 0x2875f0d6280>

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In [14]: #There is a better correlation between small bags and total bags, hinting that small ba

In [15]: #Convert types into binary numbers (conventional and organic)

In [16]: df.head()

Out[16]:		Unnamed:	Date	AveragePrice	Total Volume	4046	4225	4770	Total Bags	Small Bags	Large Bags	XLa B
	0	0	2015- 12-27	1.33	64236.62	1036.74	54454.85	48.16	8696.87	8603.62	93.25	
	1	1	2015- 12-20	1.35	54876.98	674.28	44638.81	58.33	9505.56	9408.07	97.49	
	2	2	2015- 12-13	0.93	118220.22	794.70	109149.67	130.50	8145.35	8042.21	103.14	
	3	3	2015- 12-06	1.08	78992.15	1132.00	71976.41	72.58	5811.16	5677.40	133.76	
	4	4	2015- 11-29	1.28	51039.60	941.48	43838.39	75.78	6183.95	5986.26	197.69	

In [22]: df1 = pd.get_dummies(df, columns=['type', 'region'],drop_first=True)

In [23]: df1.head()

Out[23]: **Unnamed:** Total Total Small Large Date AveragePrice 4046 4225 4770 Volume Bags Bags **Bags** 2015-0 0 1.33 64236.62 1036.74 54454.85 48.16 8696.87 8603.62 93.25 12-27

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Out[94]

	Unnamed: 0	Date	AveragePrice	Total Volume	4046	4225	4770	Total Bags	Small Bags	Large Bags	•••
1	1	2015- 12-20	1.35	54876.98	674.28	44638.81	58.33	9505.56	9408.07	97.49	
2	2	2015- 12-13	0.93	118220.22	794.70	109149.67	130.50	8145.35	8042.21	103.14	
3	3	2015- 12-06	1.08	78992.15	1132.00	71976.41	72.58	5811.16	5677.40	133.76	
4	4	2015- 11-29	1.28	51039.60	941.48	43838.39	75.78	6183.95	5986.26	197.69	

5 rows × 66 columns

```
In [26]: df1.columns
```

```
In [94]: corrtype = df.corr()
    corrtype
```

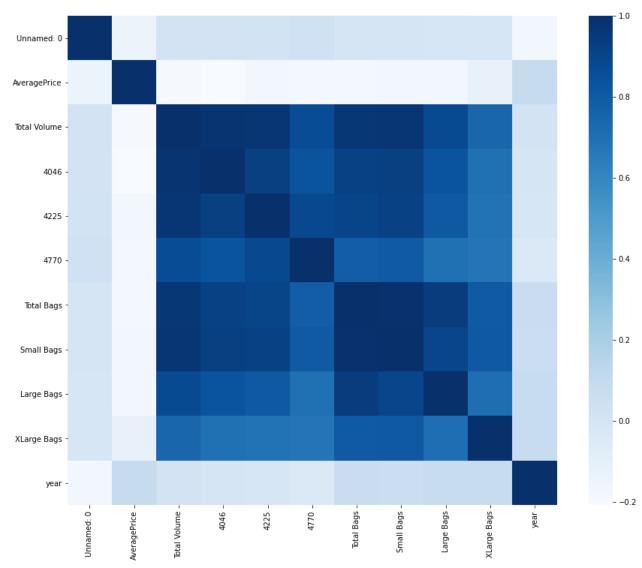
:		Unnamed:	AveragePrice	Total Volume	4046	4225	4770	Total Bags	Small Bags
	Unnamed: 0	1.000000	-0.133008	0.014035	0.017628	0.019829	0.041752	-0.002219	0.000347
	AveragePrice	-0.133008	1.000000	-0.192752	-0.208317	-0.172928	-0.179446	-0.177088	-0.174730
	Total Volume	0.014035	-0.192752	1.000000	0.977863	0.974181	0.872202	0.963047	0.967238
	4046	0.017628	-0.208317	0.977863	1.000000	0.926110	0.833389	0.920057	0.925280
	4225	0.019829	-0.172928	0.974181	0.926110	1.000000	0.887855	0.905787	0.916031
	4770	0.041752	-0.179446	0.872202	0.833389	0.887855	1.000000	0.792314	0.802733
	Total Bags	-0.002219	-0.177088	0.963047	0.920057	0.905787	0.792314	1.000000	0.994335
	Small Bags	0.000347	-0.174730	0.967238	0.925280	0.916031	0.802733	0.994335	1.000000

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	Unnamed: 0	AveragePrice	Total Volume	4046	4225	4770	Total Bags	Small Bags
Large Bags	-0.009196	-0.172940	0.880640	0.838645	0.810015	0.698471	0.943009	0.902589
XLarge Bags	-0.011546	-0.117592	0.747157	0.699377	0.688809	0.679861	0.804233	0.806845
year	-0.171667	0.093197	0.017193	0.003353	-0.009559	-0.036531	0.071552	0.063915

In [95]: plt.subplots(figsize=(15,12))
 sns.heatmap(corrtype, cmap='Blues',square=True)

Out[95]: <AxesSubplot:>



```
In [56]: X = df1.drop('type_organic',axis=1).values
y = df1['type_organic'].values
```

In [57]: from sklearn.model_selection import train_test_split

```
In [58]: X_train, X_test, y_train, y_test = train_test_split(X, y,
    test_size=0.3, random_state=101)
```

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from sklearn.preprocessing import MinMaxScaler

In [59]:

```
In [60]:
           scaler = MinMaxScaler()
In [49]:
           df1.drop('Date',axis=1,inplace=True)
           df1.drop('Unnamed: 0',axis=1,inplace=True)
In [53]:
           df1
In [54]:
                                    Total
Out[54]:
                                                                        Total
                                                                                 Small
                                                                                        Large
                                                                                              XLarge
                                            4046
                                                              4770
                  AveragePrice
                                                       4225
                                                                                                       year
                                 Volume
                                                                        Bags
                                                                                 Bags
                                                                                         Bags
                                                                                                 Bags
               0
                                                                                                  0.0 2015
                          1.33
                                 64236.62
                                          1036.74
                                                    54454.85
                                                              48.16
                                                                      8696.87
                                                                               8603.62
                                                                                        93.25
                                 54876.98
                                                              58.33
                                                                               9408.07
                                                                                                  0.0 2015
               1
                          1.35
                                           674.28
                                                    44638.81
                                                                     9505.56
                                                                                        97.49
               2
                          0.93
                                118220.22
                                           794.70
                                                  109149.67
                                                             130.50
                                                                                                  0.0 2015
                                                                     8145.35
                                                                               8042.21
                                                                                       103.14
                                                    71976.41
               3
                          1.08
                                 78992.15 1132.00
                                                                                                      2015
                                                              72.58
                                                                      5811.16
                                                                               5677.40 133.76
                                                                                                   0.0
               4
                          1.28
                                 51039.60
                                           941.48
                                                    43838.39
                                                              75.78
                                                                      6183.95
                                                                               5986.26
                                                                                       197.69
                                                                                                   0.0
                                                                                                       2015
           18244
                          1.63
                                 17074.83
                                          2046.96
                                                     1529.20
                                                               0.00
                                                                    13498.67
                                                                              13066.82
                                                                                       431.85
                                                                                                   0.0
                                                                                                      2018
           18245
                          1.71
                                 13888.04
                                          1191.70
                                                     3431.50
                                                               0.00
                                                                     9264.84
                                                                               8940.04
                                                                                       324.80
                                                                                                   0.0
                                                                                                      2018
           18246
                          1.87
                                 13766.76
                                         1191.92
                                                     2452.79 727.94
                                                                     9394.11
                                                                               9351.80
                                                                                        42.31
                                                                                                   0.0
                                                                                                      2018
           18247
                          1.93
                                 16205.22
                                         1527.63
                                                     2981.04 727.01
                                                                    10969.54
                                                                              10919.54
                                                                                         50.00
                                                                                                   0.0
                                                                                                      2018
                                 17489.58 2894.77
           18248
                          1.62
                                                     2356.13 224.53 12014.15 11988.14
                                                                                        26.01
                                                                                                  0.0 2018
          18249 rows × 64 columns
In [61]:
           X train = scaler.fit transform(X train)
           X_test = scaler.transform(X_test)
In [62]:
In [63]:
           from tensorflow.keras.models import Sequential
           from tensorflow.keras.layers import Dense
In [64]:
           X_train.shape
          (12774, 63)
Out[64]:
In [67]:
           model = Sequential()
           model.add(Dense(63,activation='relu'))
           model.add(Dense(47,activation='relu'))
           model.add(Dense(31,activation='relu'))
           model.add(Dense(15,activation='relu'))
           model.add(Dense(1,activation='sigmoid'))
```

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```
model.compile(optimizer='adam',loss='mse')
```

```
In [68]:
```

Epoch 1/400

```
model.fit(x=X train,y=y train,
          validation_data=(X_test,y_test),
         batch size=128,epochs=400)
```

```
Epoch 2/400
Epoch 3/400
Epoch 4/400
Epoch 5/400
Epoch 6/400
Epoch 7/400
Epoch 8/400
Epoch 9/400
Epoch 10/400
Epoch 11/400
Epoch 12/400
Epoch 13/400
Epoch 14/400
Epoch 15/400
Epoch 16/400
Epoch 17/400
Epoch 18/400
Epoch 19/400
Epoch 20/400
Epoch 21/400
Epoch 22/400
Epoch 23/400
Epoch 24/400
Epoch 25/400
Epoch 26/400
e-04
Epoch 27/400
100/100 [============= ] - 0s 3ms/step - loss: 7.1566e-04 - val loss: 0.
0010
Epoch 28/400
```

9/33 localhost:8954/lab

```
100/100 [============= ] - 0s 3ms/step - loss: 4.6730e-04 - val loss: 0.
0020
Epoch 29/400
0010
Epoch 30/400
100/100 [============== ] - 0s 3ms/step - loss: 3.4695e-04 - val loss: 9.
7480e-04
Epoch 31/400
100/100 [============== ] - 0s 3ms/step - loss: 8.2447e-04 - val loss: 0.
0011
Epoch 32/400
Epoch 33/400
Epoch 34/400
0.0014 - val loss: 8.4818e-04
Epoch 35/400
100/100 [============== ] - 0s 3ms/step - loss: 6.3110e-04 - val loss: 0.
0013
Epoch 36/400
100/100 [============== ] - 0s 3ms/step - loss: 4.5011e-04 - val loss: 5.
6430e-04
Epoch 37/400
100/100 [============= ] - 0s 3ms/step - loss: 4.5272e-04 - val loss: 0.
0013
Epoch 38/400
Epoch 39/400
100/100 [============== ] - 0s 3ms/step - loss: 6.1405e-04 - val loss: 9.
1763e-04
Epoch 40/400
100/100 [============= ] - 0s 3ms/step - loss: 4.6570e-04 - val loss: 0.
0049
Epoch 41/400
Epoch 42/400
100/100 [============= ] - 0s 3ms/step - loss: 3.1465e-04 - val loss: 7.
0099e-04
Epoch 43/400
100/100 [============== ] - 0s 3ms/step - loss: 4.4832e-05 - val loss: 6.
9683e-04
Epoch 44/400
0435e-04
Epoch 45/400
0023
Epoch 46/400
Epoch 47/400
Epoch 48/400
Epoch 49/400
3909e-04
Epoch 50/400
100/100 [=============== ] - 0s 2ms/step - loss: 7.4217e-04 - val loss: 0.
0026
Epoch 51/400
100/100 [=============== ] - ETA: 0s - loss: 0.001 - 0s 2ms/step - loss:
7.7015e-04 - val loss: 6.8822e-04
```

localhost:8954/lab 10/33

```
Epoch 52/400
100/100 [============== ] - 0s 2ms/step - loss: 1.6237e-04 - val loss: 6.
2931e-04
Epoch 53/400
0569e-04
Epoch 54/400
100/100 [============== ] - 0s 2ms/step - loss: 1.2144e-05 - val loss: 5.
8346e-04
Epoch 55/400
100/100 [============= ] - 0s 3ms/step - loss: 1.1864e-05 - val loss: 6.
0142e-04
Epoch 56/400
100/100 [============== ] - 0s 2ms/step - loss: 1.2065e-05 - val loss: 5.
8717e-04
Epoch 57/400
100/100 [============== ] - 0s 2ms/step - loss: 8.5260e-06 - val loss: 5.
9009e-04
Epoch 58/400
100/100 [============== ] - 0s 2ms/step - loss: 7.3052e-06 - val loss: 5.
9604e-04
Epoch 59/400
8895e-04
Epoch 60/400
100/100 [============= ] - 0s 2ms/step - loss: 7.5729e-06 - val loss: 5.
7584e-04
Epoch 61/400
100/100 [============== ] - 0s 2ms/step - loss: 6.7196e-06 - val loss: 5.
7729e-04
Epoch 62/400
100/100 [============== ] - 0s 2ms/step - loss: 6.7800e-06 - val loss: 5.
8144e-04
Epoch 63/400
100/100 [============= ] - 0s 2ms/step - loss: 8.5173e-06 - val loss: 5.
8299e-04
Epoch 64/400
100/100 [============= ] - 0s 2ms/step - loss: 6.7622e-06 - val loss: 5.
8591e-04
Epoch 65/400
100/100 [============== ] - 0s 2ms/step - loss: 5.4508e-06 - val loss: 5.
7464e-04
Epoch 66/400
100/100 [============= ] - 0s 3ms/step - loss: 1.1923e-05 - val loss: 6.
8328e-04
Epoch 67/400
100/100 [============== ] - 0s 2ms/step - loss: 7.0532e-06 - val loss: 5.
7411e-04
Epoch 68/400
Epoch 69/400
Epoch 70/400
Epoch 71/400
Epoch 72/400
Epoch 73/400
Epoch 74/400
Epoch 75/400
Epoch 76/400
```

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```
Epoch 77/400
Epoch 78/400
Epoch 79/400
Epoch 80/400
Epoch 81/400
Epoch 82/400
Epoch 83/400
Epoch 84/400
Epoch 85/400
Epoch 86/400
Epoch 87/400
Epoch 88/400
Epoch 89/400
Epoch 90/400
Epoch 91/400
Epoch 92/400
Epoch 93/400
Epoch 94/400
Epoch 95/400
Epoch 96/400
Epoch 97/400
Epoch 98/400
Epoch 99/400
Epoch 100/400
Epoch 101/400
Epoch 102/400
Epoch 103/400
Epoch 104/400
Epoch 105/400
Epoch 106/400
Epoch 107/400
Epoch 108/400
```

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```
Epoch 109/400
Epoch 110/400
Epoch 111/400
100/100 [============== ] - 0s 3ms/step - loss: 4.5956e-04 - val loss: 8.
9261e-04
Epoch 112/400
100/100 [============== ] - 0s 2ms/step - loss: 6.6280e-05 - val loss: 4.
2187e-04
Epoch 113/400
100/100 [============= ] - 0s 2ms/step - loss: 2.3725e-05 - val loss: 3.
9352e-04
Epoch 114/400
100/100 [============== ] - 0s 3ms/step - loss: 2.0784e-05 - val loss: 3.
6688e-04
Epoch 115/400
100/100 [============== ] - 0s 3ms/step - loss: 1.6208e-05 - val loss: 3.
5635e-04
Epoch 116/400
100/100 [============= ] - 0s 3ms/step - loss: 1.1163e-05 - val loss: 3.
5229e-04
Epoch 117/400
100/100 [============== ] - 0s 3ms/step - loss: 9.2332e-06 - val loss: 3.
4782e-04
Epoch 118/400
100/100 [============= ] - 0s 3ms/step - loss: 7.3485e-06 - val loss: 3.
4847e-04
Epoch 119/400
100/100 [============== ] - Os 2ms/step - loss: 6.3582e-06 - val loss: 3.
4767e-04
Epoch 120/400
100/100 [============== ] - Os 2ms/step - loss: 5.9314e-06 - val loss: 3.
4903e-04
Epoch 121/400
100/100 [=============== ] - 0s 2ms/step - loss: 4.9509e-06 - val_loss: 3.
4590e-04
Epoch 122/400
100/100 [============== ] - Os 2ms/step - loss: 4.1949e-06 - val loss: 3.
4302e-04
Epoch 123/400
100/100 [============== ] - 0s 2ms/step - loss: 3.9766e-06 - val loss: 3.
4064e-04
Epoch 124/400
100/100 [============== ] - 0s 2ms/step - loss: 3.5646e-06 - val loss: 3.
3864e-04
Epoch 125/400
100/100 [============== ] - 0s 2ms/step - loss: 2.9872e-06 - val loss: 3.
3866e-04
Epoch 126/400
4362e-04
Epoch 127/400
100/100 [============== ] - Os 2ms/step - loss: 2.6583e-06 - val loss: 3.
3797e-04
Epoch 128/400
100/100 [============= ] - Os 2ms/step - loss: 2.3066e-06 - val loss: 3.
3689e-04
Epoch 129/400
100/100 [============== ] - 0s 3ms/step - loss: 2.1418e-06 - val loss: 3.
3583e-04
Epoch 130/400
100/100 [============= ] - 0s 3ms/step - loss: 1.9219e-06 - val loss: 3.
3485e-04
Epoch 131/400
```

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```
100/100 [============= ] - 0s 3ms/step - loss: 1.8439e-06 - val loss: 3.
3604e-04
Epoch 132/400
3367e-04
Epoch 133/400
100/100 [============== ] - Os 3ms/step - loss: 1.6033e-06 - val loss: 3.
3302e-04
Epoch 134/400
100/100 [============== ] - Os 3ms/step - loss: 1.5001e-06 - val loss: 3.
3341e-04
Epoch 135/400
100/100 [============== ] - 0s 3ms/step - loss: 1.4039e-06 - val loss: 3.
3099e-04
Epoch 136/400
100/100 [============== ] - 0s 3ms/step - loss: 1.4173e-06 - val loss: 3.
4093e-04
Epoch 137/400
100/100 [============= ] - 0s 3ms/step - loss: 1.2734e-06 - val loss: 3.
3782e-04
Epoch 138/400
3445e-04
Epoch 139/400
100/100 [============== ] - 0s 3ms/step - loss: 1.0860e-06 - val loss: 3.
3425e-04
Epoch 140/400
3130e-04
Epoch 141/400
3277e-04
Epoch 142/400
100/100 [============== ] - 0s 3ms/step - loss: 8.9856e-07 - val loss: 3.
3369e-04
Epoch 143/400
100/100 [============= ] - 0s 3ms/step - loss: 8.6999e-07 - val loss: 3.
3481e-04
Epoch 144/400
100/100 [============== ] - 0s 3ms/step - loss: 7.7210e-07 - val loss: 3.
3137e-04
Epoch 145/400
100/100 [============= ] - 0s 3ms/step - loss: 7.6558e-07 - val loss: 3.
3498e-04
Epoch 146/400
3476e-04
Epoch 147/400
100/100 [============== ] - 0s 3ms/step - loss: 6.6157e-07 - val loss: 3.
3633e-04
Epoch 148/400
100/100 [============== ] - 0s 3ms/step - loss: 6.4485e-07 - val loss: 3.
2998e-04
Epoch 149/400
100/100 [============= ] - 0s 3ms/step - loss: 6.1087e-07 - val loss: 3.
3058e-04
Epoch 150/400
100/100 [============= ] - Os 3ms/step - loss: 5.7209e-07 - val loss: 3.
3705e-04
Epoch 151/400
100/100 [============== ] - 0s 3ms/step - loss: 5.7901e-07 - val loss: 3.
3523e-04
Epoch 152/400
100/100 [============== ] - Os 3ms/step - loss: 4.9363e-07 - val loss: 3.
4011e-04
```

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```
Epoch 153/400
100/100 [============== ] - 0s 2ms/step - loss: 4.8078e-07 - val loss: 3.
3878e-04
Epoch 154/400
3724e-04
Epoch 155/400
100/100 [============== ] - Os 2ms/step - loss: 4.2703e-07 - val loss: 3.
3106e-04
Epoch 156/400
100/100 [============= ] - 0s 3ms/step - loss: 4.6386e-07 - val loss: 3.
3957e-04
Epoch 157/400
100/100 [============= ] - 0s 3ms/step - loss: 4.0737e-07 - val loss: 3.
3422e-04
Epoch 158/400
100/100 [============== ] - 0s 3ms/step - loss: 4.2214e-07 - val loss: 3.
3752e-04
Epoch 159/400
100/100 [============== ] - 0s 3ms/step - loss: 3.4917e-07 - val loss: 3.
3313e-04
Epoch 160/400
3503e-04
Epoch 161/400
100/100 [============== ] - 0s 3ms/step - loss: 3.1654e-07 - val loss: 3.
3931e-04
Epoch 162/400
100/100 [============= ] - 0s 2ms/step - loss: 3.1223e-07 - val loss: 3.
3751e-04
Epoch 163/400
100/100 [============== ] - Os 2ms/step - loss: 3.6792e-07 - val loss: 3.
3751e-04
Epoch 164/400
100/100 [============= ] - 0s 2ms/step - loss: 2.8378e-07 - val loss: 3.
3876e-04
Epoch 165/400
100/100 [============= ] - 0s 3ms/step - loss: 3.5227e-07 - val loss: 3.
3990e-04
Epoch 166/400
100/100 [============= ] - 0s 3ms/step - loss: 2.3372e-07 - val loss: 3.
3004e-04
Epoch 167/400
100/100 [============== ] - 0s 3ms/step - loss: 2.5453e-07 - val loss: 3.
3819e-04
Epoch 168/400
100/100 [============= ] - 0s 3ms/step - loss: 2.5799e-07 - val loss: 3.
4426e-04
Epoch 169/400
3479e-04
Epoch 170/400
100/100 [============== ] - 0s 3ms/step - loss: 2.3680e-07 - val loss: 3.
4280e-04
Epoch 171/400
100/100 [============= ] - Os 4ms/step - loss: 2.8232e-07 - val loss: 3.
4202e-04
Epoch 172/400
100/100 [============== ] - 0s 3ms/step - loss: 1.8398e-07 - val loss: 3.
2672e-04
Epoch 173/400
3701e-04
Epoch 174/400
100/100 [============== ] - 0s 3ms/step - loss: 1.8980e-07 - val loss: 3.
```

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```
3394e-04
Epoch 175/400
100/100 [============== ] - 0s 2ms/step - loss: 2.8447e-07 - val loss: 3.
4800e-04
Epoch 176/400
Epoch 177/400
Epoch 178/400
Epoch 179/400
e-04
Epoch 180/400
100/100 [============== ] - Os 2ms/step - loss: 3.5092e-05 - val loss: 5.
2305e-04
Epoch 181/400
100/100 [============== ] - 0s 2ms/step - loss: 7.9868e-06 - val loss: 5.
3255e-04
Epoch 182/400
100/100 [============== ] - 0s 2ms/step - loss: 6.0476e-06 - val loss: 5.
1910e-04
Epoch 183/400
1677e-04
Epoch 184/400
100/100 [============== ] - 0s 2ms/step - loss: 4.8879e-06 - val loss: 4.
9908e-04
Epoch 185/400
100/100 [============== ] - 0s 2ms/step - loss: 4.3665e-06 - val loss: 4.
9999e-04
Epoch 186/400
100/100 [============== ] - 0s 2ms/step - loss: 3.9880e-06 - val loss: 4.
8275e-04
Epoch 187/400
100/100 [================ ] - 0s 2ms/step - loss: 3.8404e-06 - val_loss: 4.
8373e-04
Epoch 188/400
100/100 [============== ] - 0s 2ms/step - loss: 3.3414e-06 - val loss: 4.
8393e-04
Epoch 189/400
100/100 [============== ] - Os 2ms/step - loss: 3.0185e-06 - val loss: 4.
7865e-04
Epoch 190/400
100/100 [============== ] - Os 2ms/step - loss: 2.9102e-06 - val loss: 4.
7487e-04
Epoch 191/400
100/100 [============== ] - 0s 2ms/step - loss: 2.6254e-06 - val loss: 4.
6327e-04
Epoch 192/400
100/100 [============= ] - 0s 3ms/step - loss: 2.3612e-06 - val loss: 4.
7567e-04
Epoch 193/400
100/100 [============== ] - 0s 3ms/step - loss: 2.2339e-06 - val loss: 4.
8339e-04
Epoch 194/400
100/100 [============= ] - 0s 2ms/step - loss: 2.0714e-06 - val loss: 4.
6786e-04
Epoch 195/400
100/100 [============== ] - 0s 2ms/step - loss: 1.8945e-06 - val loss: 4.
5815e-04
Epoch 196/400
100/100 [============= ] - Os 2ms/step - loss: 1.7660e-06 - val loss: 4.
5698e-04
Epoch 197/400
```

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```
100/100 [============= ] - 0s 2ms/step - loss: 1.6337e-06 - val loss: 4.
5506e-04
Epoch 198/400
5248e-04
Epoch 199/400
100/100 [============== ] - 0s 3ms/step - loss: 1.4721e-06 - val loss: 4.
5655e-04
Epoch 200/400
100/100 [============== ] - 0s 2ms/step - loss: 1.3134e-06 - val loss: 4.
5335e-04
Epoch 201/400
100/100 [============== ] - 0s 2ms/step - loss: 1.2110e-06 - val loss: 4.
4412e-04
Epoch 202/400
100/100 [============== ] - 0s 2ms/step - loss: 1.1878e-06 - val loss: 4.
4612e-04
Epoch 203/400
100/100 [============= ] - 0s 3ms/step - loss: 1.0741e-06 - val loss: 4.
4726e-04
Epoch 204/400
5065e-04
Epoch 205/400
100/100 [============== ] - 0s 3ms/step - loss: 9.2463e-07 - val loss: 4.
4500e-04
Epoch 206/400
4248e-04
Epoch 207/400
5295e-04
Epoch 208/400
100/100 [============= ] - 0s 3ms/step - loss: 7.7843e-07 - val loss: 4.
4538e-04
Epoch 209/400
100/100 [============= ] - 0s 3ms/step - loss: 7.0635e-07 - val loss: 4.
3119e-04
Epoch 210/400
100/100 [============== ] - 0s 3ms/step - loss: 6.6850e-07 - val loss: 4.
4348e-04
Epoch 211/400
100/100 [============= ] - Os 3ms/step - loss: 6.3289e-07 - val loss: 4.
2955e-04
Epoch 212/400
3215e-04
Epoch 213/400
100/100 [============== ] - 0s 3ms/step - loss: 5.4141e-07 - val loss: 4.
2992e-04
Epoch 214/400
100/100 [============== ] - 0s 3ms/step - loss: 5.1645e-07 - val loss: 4.
2066e-04
Epoch 215/400
100/100 [============== ] - 0s 4ms/step - loss: 4.8808e-07 - val loss: 4.
2783e-04
Epoch 216/400
2726e-04
Epoch 217/400
100/100 [============= ] - 0s 3ms/step - loss: 4.4257e-07 - val loss: 4.
2415e-04
Epoch 218/400
100/100 [============== ] - Os 3ms/step - loss: 4.0941e-07 - val loss: 4.
1727e-04
```

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```
Epoch 219/400
100/100 [============= ] - 0s 3ms/step - loss: 3.9260e-07 - val loss: 4.
1921e-04
Epoch 220/400
2399e-04
Epoch 221/400
1721e-04
Epoch 222/400
100/100 [============== ] - 0s 3ms/step - loss: 3.2661e-07 - val loss: 4.
1976e-04
Epoch 223/400
100/100 [============== ] - 0s 3ms/step - loss: 3.0484e-07 - val loss: 4.
1776e-04
Epoch 224/400
100/100 [============== ] - 0s 3ms/step - loss: 2.8612e-07 - val loss: 4.
1437e-04
Epoch 225/400
100/100 [============== ] - 0s 3ms/step - loss: 2.7044e-07 - val loss: 4.
1773e-04
Epoch 226/400
1114e-04
Epoch 227/400
100/100 [============== ] - 0s 3ms/step - loss: 2.5110e-07 - val loss: 4.
1472e-04
Epoch 228/400
100/100 [============== ] - 0s 3ms/step - loss: 2.2824e-07 - val loss: 4.
0951e-04
Epoch 229/400
100/100 [============= ] - 0s 3ms/step - loss: 2.2033e-07 - val loss: 4.
0294e-04
Epoch 230/400
100/100 [============= ] - 0s 3ms/step - loss: 2.0583e-07 - val loss: 4.
0203e-04
Epoch 231/400
100/100 [============== ] - Os 3ms/step - loss: 2.0181e-07 - val loss: 4.
0545e-04
Epoch 232/400
100/100 [============= ] - 0s 3ms/step - loss: 1.8687e-07 - val loss: 4.
0685e-04
Epoch 233/400
100/100 [============== ] - Os 2ms/step - loss: 1.7391e-07 - val loss: 4.
0200e-04
Epoch 234/400
100/100 [============== ] - 0s 2ms/step - loss: 1.6958e-07 - val loss: 4.
0303e-04
Epoch 235/400
100/100 [============= ] - 0s 2ms/step - loss: 1.6102e-07 - val loss: 4.
0507e-04
Epoch 236/400
100/100 [============== ] - Os 2ms/step - loss: 1.4897e-07 - val loss: 3.
9688e-04
Epoch 237/400
100/100 [============= ] - Os 1ms/step - loss: 1.4215e-07 - val loss: 4.
1268e-04
Epoch 238/400
100/100 [============== ] - 0s 2ms/step - loss: 1.3765e-07 - val loss: 3.
9745e-04
Epoch 239/400
100/100 [=============== ] - 0s 1ms/step - loss: 1.3472e-07 - val_loss: 4.
0517e-04
Epoch 240/400
100/100 [============== ] - Os 1ms/step - loss: 1.3124e-07 - val loss: 4.
```

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```
0202e-04
Epoch 241/400
100/100 [============== ] - 0s 2ms/step - loss: 1.1946e-07 - val loss: 4.
0709e-04
Epoch 242/400
100/100 [============== ] - Os 1ms/step - loss: 1.1047e-07 - val loss: 3.
9268e-04
Epoch 243/400
100/100 [============== ] - Os 1ms/step - loss: 1.0730e-07 - val loss: 3.
9981e-04
Epoch 244/400
100/100 [============= ] - 0s 2ms/step - loss: 1.0225e-07 - val loss: 3.
9408e-04
Epoch 245/400
100/100 [============== ] - Os 1ms/step - loss: 9.8306e-08 - val loss: 3.
9382e-04
Epoch 246/400
100/100 [============== ] - Os 2ms/step - loss: 9.1467e-08 - val loss: 4.
0594e-04
Epoch 247/400
100/100 [============= ] - Os 1ms/step - loss: 9.0265e-08 - val loss: 3.
9842e-04
Epoch 248/400
9497e-04
Epoch 249/400
100/100 [============== ] - Os 1ms/step - loss: 8.1510e-08 - val loss: 3.
9193e-04
Epoch 250/400
100/100 [============== ] - Os 2ms/step - loss: 7.7669e-08 - val loss: 3.
9235e-04
Epoch 251/400
100/100 [============== ] - 0s 2ms/step - loss: 7.7025e-08 - val loss: 3.
9140e-04
Epoch 252/400
8711e-04
Epoch 253/400
100/100 [============== ] - 0s 2ms/step - loss: 6.6762e-08 - val loss: 4.
0132e-04
Epoch 254/400
100/100 [============== ] - Os 2ms/step - loss: 6.8533e-08 - val loss: 3.
9928e-04
Epoch 255/400
100/100 [============== ] - Os 2ms/step - loss: 6.2734e-08 - val loss: 3.
9070e-04
Epoch 256/400
100/100 [============== ] - Os 2ms/step - loss: 6.0041e-08 - val loss: 3.
9859e-04
Epoch 257/400
100/100 [============= ] - 0s 2ms/step - loss: 5.5033e-08 - val loss: 3.
9645e-04
Epoch 258/400
100/100 [============== ] - Os 2ms/step - loss: 5.3088e-08 - val loss: 3.
9896e-04
Epoch 259/400
100/100 [============== ] - 0s 3ms/step - loss: 5.1411e-08 - val loss: 3.
9615e-04
Epoch 260/400
100/100 [============== ] - Os 2ms/step - loss: 4.8585e-08 - val loss: 3.
9641e-04
Epoch 261/400
100/100 [============= ] - Os 2ms/step - loss: 4.7567e-08 - val loss: 3.
9350e-04
Epoch 262/400
```

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```
8336e-04
Epoch 263/400
8959e-04
Epoch 264/400
100/100 [=============== ] - Os 2ms/step - loss: 3.8861e-08 - val loss: 3.
8694e-04
Epoch 265/400
100/100 [============== ] - Os 2ms/step - loss: 3.7202e-08 - val loss: 3.
8689e-04
Epoch 266/400
100/100 [============== ] - 0s 3ms/step - loss: 3.5381e-08 - val loss: 3.
8256e-04
Epoch 267/400
100/100 [============== ] - 0s 3ms/step - loss: 3.8543e-08 - val loss: 3.
7916e-04
Epoch 268/400
100/100 [============= ] - 0s 3ms/step - loss: 3.2347e-08 - val loss: 3.
9393e-04
Epoch 269/400
8768e-04
Epoch 270/400
100/100 [============== ] - 0s 3ms/step - loss: 2.9730e-08 - val loss: 3.
9014e-04
Epoch 271/400
100/100 [============== ] - 0s 3ms/step - loss: 3.1430e-08 - val loss: 3.
8723e-04
Epoch 272/400
9248e-04
Epoch 273/400
100/100 [============== ] - 0s 3ms/step - loss: 3.2788e-08 - val loss: 3.
9513e-04
Epoch 274/400
100/100 [============== ] - 0s 3ms/step - loss: 2.7194e-08 - val loss: 3.
9184e-04
Epoch 275/400
100/100 [=============== ] - 0s 3ms/step - loss: 2.4864e-08 - val loss: 3.
7845e-04
Epoch 276/400
100/100 [============== ] - 0s 3ms/step - loss: 2.3135e-08 - val loss: 3.
8241e-04
Epoch 277/400
8517e-04
Epoch 278/400
100/100 [============== ] - 0s 3ms/step - loss: 2.2689e-08 - val loss: 3.
8384e-04
Epoch 279/400
100/100 [============== ] - Os 2ms/step - loss: 2.0432e-08 - val loss: 3.
8852e-04
Epoch 280/400
100/100 [============== ] - Os 2ms/step - loss: 1.9629e-08 - val loss: 3.
8879e-04
Epoch 281/400
100/100 [============= ] - 0s 2ms/step - loss: 1.8577e-08 - val loss: 3.
9266e-04
Epoch 282/400
100/100 [============== ] - 0s 3ms/step - loss: 1.7233e-08 - val loss: 3.
7833e-04
Epoch 283/400
100/100 [============== ] - Os 3ms/step - loss: 1.9430e-08 - val loss: 3.
8309e-04
```

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```
Epoch 284/400
100/100 [============== ] - 0s 3ms/step - loss: 1.5532e-08 - val loss: 3.
7629e-04
Epoch 285/400
100/100 [============== ] - 0s 3ms/step - loss: 2.2215e-08 - val loss: 3.
8785e-04
Epoch 286/400
100/100 [============== ] - 0s 3ms/step - loss: 1.5361e-08 - val loss: 3.
7930e-04
Epoch 287/400
100/100 [============= ] - 0s 3ms/step - loss: 1.6851e-08 - val loss: 3.
8424e-04
Epoch 288/400
100/100 [============== ] - 0s 3ms/step - loss: 2.2881e-08 - val loss: 3.
7418e-04
Epoch 289/400
100/100 [=============== ] - 0s 3ms/step - loss: 1.4585e-08 - val_loss: 3.
8720e-04
Epoch 290/400
9677e-04
Epoch 291/400
8486e-04
Epoch 292/400
100/100 [============= ] - 0s 3ms/step - loss: 1.2844e-08 - val loss: 3.
8365e-04
Epoch 293/400
Epoch 294/400
Epoch 295/400
Epoch 296/400
Epoch 297/400
Epoch 298/400
Epoch 299/400
Epoch 300/400
Epoch 301/400
Epoch 302/400
Epoch 303/400
Epoch 304/400
Epoch 305/400
Epoch 306/400
Epoch 307/400
Epoch 308/400
Epoch 309/400
Epoch 310/400
Epoch 311/400
```

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```
Epoch 312/400
Epoch 313/400
Epoch 314/400
Epoch 315/400
Epoch 316/400
Epoch 317/400
Epoch 318/400
Epoch 319/400
Epoch 320/400
Epoch 321/400
Epoch 322/400
Epoch 323/400
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Epoch 332/400
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Epoch 335/400
Epoch 336/400
Epoch 337/400
Epoch 338/400
Epoch 339/400
Epoch 340/400
Epoch 341/400
Epoch 342/400
Epoch 343/400
Epoch 344/400
```

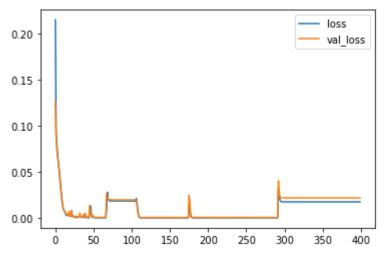
localhost:8954/lab 22/33

```
Epoch 345/400
Epoch 346/400
Epoch 347/400
Epoch 348/400
Epoch 349/400
Epoch 350/400
Epoch 351/400
Epoch 352/400
Epoch 353/400
Epoch 354/400
Epoch 355/400
Epoch 356/400
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Epoch 362/400
Epoch 363/400
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Epoch 365/400
Epoch 366/400
Epoch 367/400
Epoch 368/400
Epoch 369/400
Epoch 370/400
Epoch 371/400
Epoch 372/400
Epoch 373/400
Epoch 374/400
Epoch 375/400
Epoch 376/400
```

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```
Epoch 377/400
 Epoch 378/400
 Epoch 379/400
 Epoch 380/400
 Epoch 381/400
 Epoch 382/400
 Epoch 383/400
 Epoch 384/400
 Epoch 385/400
 Epoch 386/400
 Epoch 387/400
 Epoch 388/400
 Epoch 389/400
 Epoch 390/400
 Epoch 391/400
 Epoch 392/400
 Epoch 393/400
 Epoch 394/400
 Epoch 395/400
 Epoch 396/400
 Epoch 397/400
 Epoch 398/400
 Epoch 399/400
 Epoch 400/400
 Out[68]: <keras.callbacks.History at 0x2872d93d220>
 model loss = pd.DataFrame(model.history.history)
In [69]:
 model loss.plot()
Out[69]: <AxesSubplot:>
```

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```
Epoch 1/400
Epoch 2/400
Epoch 3/400
Epoch 4/400
Epoch 5/400
Epoch 6/400
Epoch 7/400
Epoch 8/400
Epoch 9/400
Epoch 10/400
Epoch 11/400
Epoch 12/400
Epoch 13/400
Epoch 14/400
Epoch 15/400
Epoch 16/400
Epoch 17/400
Epoch 18/400
Epoch 19/400
```

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```
Epoch 20/400
Epoch 21/400
Epoch 22/400
Epoch 23/400
Epoch 24/400
Epoch 25/400
Epoch 26/400
Epoch 27/400
Epoch 28/400
Epoch 29/400
Epoch 30/400
Epoch 31/400
Epoch 32/400
Epoch 33/400
Epoch 34/400
Epoch 35/400
Epoch 36/400
Epoch 37/400
Epoch 38/400
Epoch 39/400
Epoch 40/400
Epoch 41/400
Epoch 42/400
e-04
Epoch 43/400
400/400 [============= ] - 1s 3ms/step - loss: 6.9747e-04 - val loss: 3.
5312e-04
Epoch 44/400
400/400 [============= ] - 1s 2ms/step - loss: 3.2136e-05 - val loss: 3.
3227e-04
Epoch 45/400
400/400 [============= ] - 1s 2ms/step - loss: 4.2066e-06 - val loss: 3.
0644e-04
Epoch 46/400
1392e-04
Epoch 47/400
400/400 [============= ] - 1s 3ms/step - loss: 4.6051e-07 - val loss: 3.
0867e-04
Epoch 48/400
400/400 [============= ] - 1s 3ms/step - loss: 2.7386e-07 - val loss: 3.
0686e-04
```

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```
Epoch 49/400
400/400 [============= ] - 1s 3ms/step - loss: 1.8898e-07 - val loss: 3.
0299e-04
Epoch 50/400
0680e-04
Epoch 51/400
400/400 [============= ] - 1s 3ms/step - loss: 1.1499e-07 - val loss: 3.
1375e-04
Epoch 52/400
400/400 [============= ] - 1s 3ms/step - loss: 9.3453e-08 - val loss: 3.
0438e-04
Epoch 53/400
400/400 [============== ] - 1s 3ms/step - loss: 8.3383e-08 - val loss: 3.
1184e-04
Epoch 54/400
400/400 [============= ] - 1s 2ms/step - loss: 6.0949e-08 - val loss: 3.
0064e-04
Epoch 55/400
400/400 [============= ] - 1s 2ms/step - loss: 4.9849e-08 - val loss: 3.
0140e-04
Epoch 56/400
0692e-04
Epoch 57/400
400/400 [============== ] - 1s 2ms/step - loss: 3.1444e-08 - val loss: 3.
0140e-04
Epoch 58/400
400/400 [============== ] - 1s 2ms/step - loss: 3.0704e-08 - val loss: 3.
0890e-04
Epoch 59/400
400/400 [============== ] - 1s 2ms/step - loss: 2.4423e-08 - val loss: 3.
0521e-04
Epoch 60/400
400/400 [============== ] - 1s 2ms/step - loss: 2.3847e-08 - val loss: 3.
1268e-04
Epoch 61/400
400/400 [============== ] - 1s 2ms/step - loss: 1.9114e-08 - val loss: 3.
0368e-04
Epoch 62/400
0607e-04
Epoch 63/400
400/400 [============= ] - 1s 3ms/step - loss: 1.3365e-08 - val loss: 3.
1449e-04
Epoch 64/400
400/400 [============= ] - 1s 3ms/step - loss: 1.1139e-08 - val loss: 3.
0676e-04
Epoch 65/400
400/400 [============= ] - 1s 2ms/step - loss: 1.0556e-08 - val loss: 3.
0924e-04
Epoch 66/400
400/400 [============= ] - 1s 2ms/step - loss: 9.2955e-09 - val loss: 3.
0247e-04
Epoch 67/400
400/400 [============== ] - 1s 2ms/step - loss: 8.3164e-09 - val loss: 3.
0251e-04
Epoch 68/400
400/400 [============= ] - 1s 2ms/step - loss: 6.6939e-09 - val loss: 2.
9127e-04
Epoch 69/400
9999e-04
Epoch 70/400
400/400 [============= ] - 1s 2ms/step - loss: 5.1736e-09 - val loss: 2.
```

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```
9130e-04
Epoch 71/400
400/400 [============= ] - 1s 3ms/step - loss: 4.8427e-09 - val loss: 3.
0437e-04
Epoch 72/400
400/400 [=============] - 1s 2ms/step - loss: 4.6056e-09 - val loss: 2.
9183e-04
Epoch 73/400
400/400 [============= ] - 1s 2ms/step - loss: 3.6746e-09 - val loss: 2.
8668e-04
Epoch 74/400
400/400 [============= ] - 1s 3ms/step - loss: 4.6099e-09 - val loss: 2.
8745e-04
Epoch 75/400
400/400 [============= ] - 1s 2ms/step - loss: 8.3305e-09 - val loss: 3.
5985e-04
Epoch 76/400
Epoch 77/400
Epoch 78/400
Epoch 79/400
Epoch 80/400
Epoch 81/400
Epoch 82/400
Epoch 83/400
Epoch 84/400
Epoch 85/400
Epoch 86/400
Epoch 87/400
Epoch 88/400
Epoch 89/400
Epoch 90/400
Epoch 91/400
Epoch 92/400
Epoch 93/400
Epoch 94/400
Epoch 95/400
Epoch 96/400
Epoch 97/400
Epoch 98/400
Epoch 00098: early stopping
```

Out[72]: <keras.callbacks.History at 0x2872dd7d6a0>

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```
model loss = pd.DataFrame(model.history.history)
In [73]:
      model loss.plot()
     <AxesSubplot:>
Out[73]:
      0.05
                                  055
                                  val loss
      0.04
      0.03
      0.02
      0.01
      0.00
                          60
         0
               20
                    40
                               80
                                     100
      #Early stopping did not do well in this scenario.
In [74]:
      #Try dropouts
      from tensorflow.keras.layers import Dropout
In [82]:
In [83]:
      model = Sequential()
      model.add(Dense(63,activation='relu'))
      model.add(Dropout(0.5))
      model.add(Dense(47,activation='relu'))
      model.add(Dropout(0.5))
      model.add(Dense(31,activation='relu'))
      model.add(Dropout(0.5))
      model.add(Dense(15,activation='relu'))
      model.add(Dropout(0.5))
      model.add(Dense(1,activation='sigmoid'))
      model.compile(loss='binary_crossentropy',optimizer='adam')
In [84]:
      model.fit(x=X_train,y=y_train,epochs=600,validation_data=(X_test,y_test),
            callbacks=[early stop])
      Epoch 1/600
      Epoch 2/600
      Epoch 3/600
      Epoch 4/600
      Epoch 5/600
      Epoch 6/600
```

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Epoch 7/600

```
Epoch 8/600
Epoch 9/600
Epoch 10/600
Epoch 11/600
Epoch 12/600
Epoch 13/600
Epoch 14/600
Epoch 15/600
Epoch 16/600
Epoch 17/600
Epoch 18/600
Epoch 19/600
Epoch 20/600
Epoch 21/600
Epoch 22/600
Epoch 23/600
Epoch 24/600
Epoch 25/600
Epoch 26/600
Epoch 27/600
Epoch 28/600
Epoch 29/600
Epoch 30/600
Epoch 31/600
Epoch 32/600
Epoch 33/600
Epoch 34/600
Epoch 35/600
Epoch 36/600
Epoch 37/600
Epoch 38/600
Epoch 39/600
```

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```
Epoch 40/600
Epoch 41/600
Epoch 42/600
Epoch 43/600
Epoch 44/600
Epoch 45/600
Epoch 46/600
Epoch 47/600
Epoch 48/600
Epoch 49/600
Epoch 50/600
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Epoch 66/600
Epoch 67/600
Epoch 68/600
Epoch 69/600
Epoch 70/600
Epoch 71/600
Epoch 72/600
```

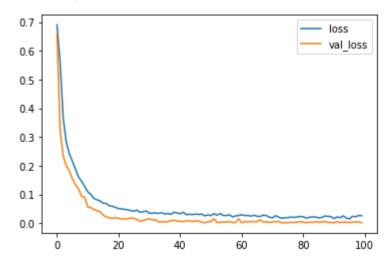
localhost:8954/lab 31/33

```
Epoch 73/600
 Epoch 74/600
 Epoch 75/600
 e-04
 Epoch 76/600
 Epoch 77/600
 Epoch 78/600
 Epoch 79/600
 Epoch 80/600
 Epoch 81/600
 Epoch 82/600
 Epoch 83/600
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 Epoch 91/600
 Epoch 92/600
 Epoch 93/600
 Epoch 94/600
 Epoch 95/600
 Epoch 96/600
 Epoch 97/600
 Epoch 98/600
 Epoch 99/600
 Epoch 100/600
 Epoch 00100: early stopping
Out[84]: <keras.callbacks.History at 0x2873043ca30>
```

In [85]: model_loss = pd.DataFrame(model.history.history)
 model_loss.plot()

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Out[85]: <AxesSubplot:>



In [79]: #Model performed much better after dropouts were added!
In []:

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