

In []:

executed in 11.6s, finished 16:21:01 2020-10-07

In [1]:

```
from GlucoCheck.glucoCheck import glucoCheckOps
import pandas as pd
import random
import numpy as np
from tqdm.auto import tqdm

from scipy import stats

import random
import re
from dateutil.parser import parse

import warnings
warnings.filterwarnings('ignore')

import os
```

executed in 12.3s, finished 10:05:40 2020-10-14

Using TensorFlow backend.

In [2]:

```
def createGap(df,start,end):
    """
    Creating a Gap
    input:
        start: seed
        end: seed + gap
    output:
        df: dataframe with index => DisplayTime value => GlucoseValues and a gap from start to end (inputs)
    """

    #df = readData()
    l = len(df.index)
    if end>l:
        end = l

    for i in range(start,end):
        df['GlucoseValue'][i]=0

    return df
```

executed in 24ms, finished 10:05:40 2020-10-14

In [3]:

```
#Extract Data
data = pd.read_csv("~/Desktop/NCSA_genomics/Python - notebooks/GlucoCheck/Data/Hall/data_hall_raw.csv")
```

executed in 242ms, finished 10:05:41 2020-10-14

In [4]:

```
data = data[data['subjectId']=='1636-69-032']
data = data.reset_index(drop=True)
```

executed in 62ms, finished 10:05:41 2020-10-14

In [5]:

```
# data.drop(['subjectId'], axis=1, inplace=True)
# data['Display Time'] = data['Display Time'].apply(lambda x: pd.datetime.strptime(x, '%Y-%m-%d %H:%M:%S'))
# data = data.set_index(['Display Time'], drop=True)
# data.to_csv("~/Desktop/original.csv")
data
```

executed in 249ms, finished 10:05:41 2020-10-14

Out[5]:

	subjectId	Display Time	GlucoseValue
0	1636-69-032	1/13/16 12:58	122
1	1636-69-032	1/13/16 13:03	123
2	1636-69-032	1/13/16 13:08	124
3	1636-69-032	1/13/16 13:13	128
4	1636-69-032	1/13/16 13:18	133
...
1778	1636-69-032	1/19/16 17:12	101
1779	1636-69-032	1/19/16 17:17	98
1780	1636-69-032	1/19/16 17:22	101
1781	1636-69-032	1/19/16 17:27	106
1782	1636-69-032	1/19/16 17:32	107

1783 rows x 3 columns

In [29]:

```
obj = glucoCheckOps()
```

executed in 7ms, finished 12:07:53 2020-10-14

Object Created!

In [77]:

```
1 fullData = obj.hall_data#.append(data)
2 fullData = fullData.reset_index(drop=True)
```

executed in 41ms, finished 13:26:44 2020-10-14

In [78]:

```
1 fullData
```

executed in 47ms, finished 13:26:45 2020-10-14

Out[78]:

	subjectId	Display Time	GlucoseValue
0	1636-69-001	2/3/14 03:42	93
1	1636-69-001	2/3/14 03:47	93
2	1636-69-001	2/3/14 03:52	93
3	1636-69-001	2/3/14 03:57	95
4	1636-69-001	2/3/14 04:02	96
...
105421	2133-041	7/11/17 20:21	70
105422	2133-041	7/11/17 20:26	64
105423	2133-041	7/11/17 20:31	61
105424	2133-041	7/11/17 20:36	62
105425	2133-041	7/11/17 20:41	66

105426 rows x 3 columns

In [79]:

```
obj.train(fullData)
```

executed in 16m 43s, finished 13:43:37 2020-10-14

Model trained successfully!

In [107]:

```
seed_points = [288,490,921,1036,1160,1604]
```

executed in 32ms, finished 14:06:19 2020-10-14

In []:

executed in 19ms, finished 13:02:49 2020-10-07

In [108]:

```
#for gap size 100
ioa_gap100 = list()
fb_gap100 = list()
mad_gap100 = list()
rmse_aan100 = list()
```

```

mape_gap100 = list()

for seed in tqdm(seed_points):
    start = seed
    end = seed+99
    data_with_missing = data.copy()
    data_with_missing = createGap(data_with_missing, start, end)

    imputed_data = obj.impute(data_with_missing, 1)

    mad = obj.mad(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    ioa = obj.index_agreement(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    fb = obj.fracBias(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    rmse = obj.rmse(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    mape = obj.mape(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())

    mad_gap100.append(mad)
    ioa_gap100.append(iao)
    fb_gap100.append(fb)
    rmse_gap100.append(rmse)
    mape_gap100.append(mape)

```

executed in 2.15s, finished 14:06:22 2020-10-14

100%  6/6 [00:04:00:00, 1.36t/s]

In [109]:

```

# for gap size 50
ioa_gap50 = list()
fb_gap50 = list()
mad_gap50 = list()
rmse_gap50 = list()
mape_gap50 = list()

for seed in tqdm(seed_points):
    start = seed
    end = seed+49
    data_with_missing = data.copy()
    data_with_missing = createGap(data_with_missing, start, end)


    imputed_data = obj.impute(data_with_missing, 1)

    mad = obj.mad(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    ioa = obj.index_agreement(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    fb = obj.fracBias(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    rmse = obj.rmse(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    mape = obj.mape(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())

    mad_gap50.append(mad)
    ioa_gap50.append(iao)
    fb_gap50.append(fb)
    rmse_gap50.append(rmse)
    mape_gap50.append(mape)

```

executed in 1.25s, finished 14:06:23 2020-10-14

100%  6/6 [00:02:00:00, 2.66t/s]

In [110]:

```

# for gap size 30
ioa_gap30 = list()
fb_gap30 = list()
mad_gap30 = list()
rmse_gap30 = list()
mape_gap30 = list()

for seed in tqdm(seed_points):
    start = seed
    end = start+29
    data_with_missing = data.copy()
    data_with_missing = createGap(data_with_missing, start, end)

    imputed_data = obj.impute(data_with_missing, 1)

    mad = obj.mad(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    ioa = obj.index_agreement(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    fb = obj.fracBias(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    rmse = obj.rmse(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    mape = obj.mape(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())

    mad_gap30.append(mad)
    ioa_gap30.append(iao)
    fb_gap30.append(fb)
    rmse_gap30.append(rmse)
    mape_gap30.append(mape)

```

executed in 909ms, finished 14:06:24 2020-10-14

100%  6/6 [00:01:00:00, 5.89t/s]

In [111]:

```

# for gap size 15
ioa_gap15 = list()
fb_gap15 = list()
mad_gap15 = list()
rmse_gap15 = list()
mape_gap15 = list()

for seed in tqdm(seed_points):
    start = seed
    end = start+14
    data_with_missing = data.copy()
    data_with_missing = createGap(data_with_missing, start, end)

    imputed_data = obj.impute(data_with_missing, 1)

    mad = obj.mad(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    ioa = obj.index_agreement(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    fb = obj.fracBias(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    rmse = obj.rmse(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())
    mape = obj.mape(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist(), np.asarray(data['GlucoseValue'])[start:end-1].tolist())

    mad_gap15.append(mad)
    ioa_gap15.append(iao)
    fb_gap15.append(fb)
    rmse_gap15.append(rmse)
    mape_gap15.append(mape)

```

executed in 597ms, finished 14:06:25 2020-10-14

100%  6/6 [05:39:00:00, 56.52s/t]

In [112]:

```

# for gap size 5
ioa_gap5 = list()
fb_gap5 = list()
mad_gap5 = list()
rmse_gap5 = list()
mape_gap5 = list()

for seed in tqdm(seed_points):
    start = seed
    end = start+4
    data_with_missing = data.copy()
    data_with_missing = createGap(data_with_missing, start, end)

    imputed_data = obj.impute(data_with_missing, 1)

```

```

mad = obj.mad(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist()),np.asarray(data['GlucoseValue'])[start:end-1].tolist())
ioa = obj.index_agreement(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist()),np.asarray(data['GlucoseValue'])[start:end-1].tolist())
fb = obj.fracBias(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist()),np.asarray(data['GlucoseValue'])[start:end-1].tolist())
rmse = obj.rmse(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist()),np.asarray(data['GlucoseValue'])[start:end-1].tolist())
mape = obj.mape(np.asarray(imputed_data['GlucoseValue'])[start:end-1].tolist()),np.asarray(data['GlucoseValue'])[start:end-1].tolist())

mad_gap5.append(mad)
ioa_gap5.append(ioa)
fb_gap5.append(fb)
rmse_gap5.append(rmse)
mape_gap5.append(mape)

```

executed in 402ms, finished 14:06:25 2020-10-14

100% 6/6 [00:04<00:00, 1.28it/s]

In [113]:

```

IOA = pd.DataFrame({'Gap:5':ioa_gap5, 'Gap:15':ioa_gap15, 'Gap:30':ioa_gap30, 'Gap:50':ioa_gap50, 'Gap:100':ioa_gap100})
IOA

```

executed in 78ms, finished 14:06:25 2020-10-14

Out[113]:

	Gap:5	Gap:15	Gap:30	Gap:50	Gap:100
0	0.524590	0.732829	0.794466	0.688385	0.633212
1	0.282136	0.599247	0.496989	0.514146	0.471362
2	0.418301	0.846204	0.910656	0.851831	0.619313
3	0.444444	0.730517	0.683936	0.419430	0.440915
4	0.830189	0.929628	0.783506	0.754882	0.461313
5	0.512315	0.639980	0.680301	0.605894	0.318035

In [114]:

```

1 import matplotlib.pyplot as plt
2 import matplotlib.ticker as ticker
3 import seaborn as sns

```

executed in 8ms, finished 14:06:25 2020-10-14

In []:

```

1

```

executed in 11ms, finished 14:05:48 2020-10-14

In []:

```

1

```

executed in 8ms, finished 14:05:45 2020-10-14

In [119]:

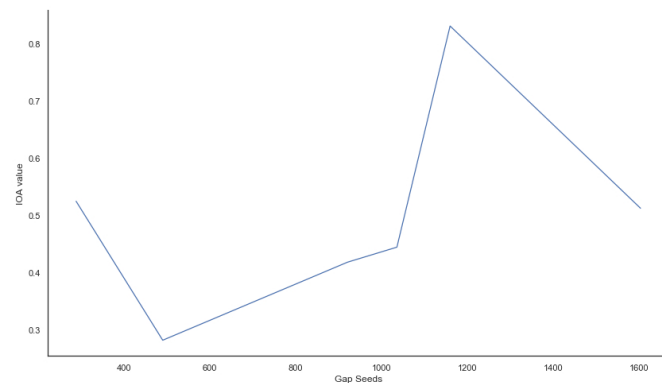
```

1 plt.figure(figsize=(14,8))
2 sns.set(style="white")
3 fig = sns.lineplot(x = seed_points, y = IOA['Gap:5'], data = IOA, palette="tab10", linewidth=1.25)
4 sns.despine()
5
6 fig.set_xlabel('Gap Seeds')
7 fig.set_ylabel('IOA value')
8

```

executed in 848ms, finished 14:10:52 2020-10-14

Out[119]: Text(0, 0.5, 'IOA value')



In [120]:

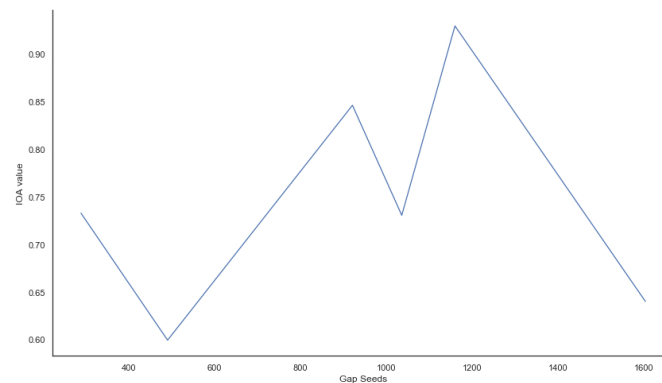
```

1 plt.figure(figsize=(14,8))
2 sns.set(style="white")
3 fig = sns.lineplot(x = seed_points, y = IOA['Gap:15'], data = IOA, palette="tab10", linewidth=1.25)
4 sns.despine()
5
6 fig.set_xlabel('Gap Seeds')
7 fig.set_ylabel('IOA value')
8

```

executed in 1.02s, finished 14:11:45 2020-10-14

Out[120]: Text(0, 0.5, 'IOA value')



In [121]:

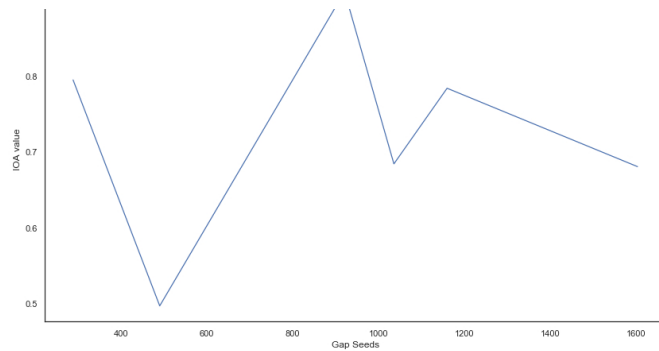
```

1 plt.figure(figsize=(14,8))
2 sns.set(style="white")
3 fig = sns.lineplot(x = seed_points, y = IOA['Gap:30'], data = IOA, palette="tab10", linewidth=1.25)
4 sns.despine()
5
6 fig.set_xlabel('Gap Seeds')
7 fig.set_ylabel('IOA value')
8

```

executed in 815ms, finished 14:11:51 2020-10-14

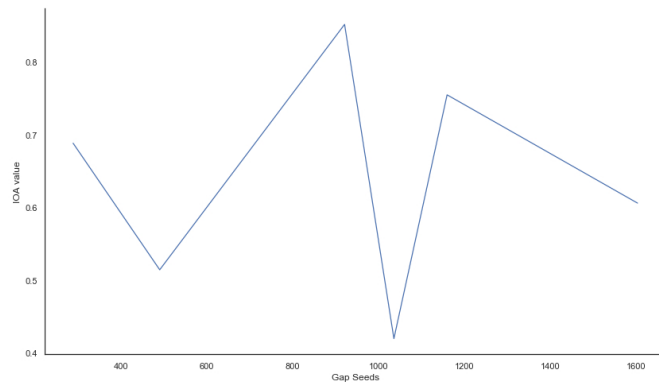
Out[121]: Text(0, 0.5, 'IOA value')



```
In [122]: 1 plt.figure(figsize=(14,8))
2 sns.set(style="white")
3 fig = sns.lineplot(x = seed_points, y = IOA['Gap:50'], data = IOA, palette="tab10", linewidth=1.25)
4 sns.despine()
5
6 fig.set_xlabel('Gap Seeds')
7 fig.set_ylabel('IOA value')
```

executed in 1.15s, finished 14:11:56 2020-10-14

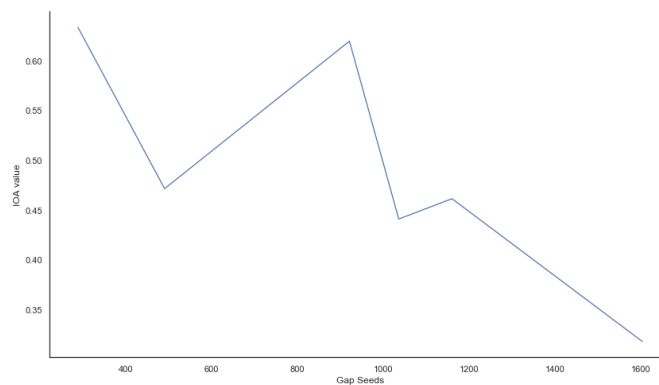
Out[122]: Text(0, 0.5, 'IOA value')



```
In [123]: 1 plt.figure(figsize=(14,8))
2 sns.set(style="white")
3 fig = sns.lineplot(x = seed_points, y = IOA['Gap:100'], data = IOA, palette="tab10", linewidth=1.25)
4 sns.despine()
5
6 fig.set_xlabel('Gap Seeds')
7 fig.set_ylabel('IOA value')
```

executed in 843ms, finished 14:12:04 2020-10-14

Out[123]: Text(0, 0.5, 'IOA value')



In []:

1

In []:

1

```
In [73]: MAD = pd.DataFrame({'Gap:5':mad_gap5, 'Gap:15':mad_gap15, 'Gap:30':mad_gap30, 'Gap:50':mad_gap50, 'Gap:100':mad_gap100})
MAD
```

executed in 47ms, finished 13:25:47 2020-10-14

Out[73]:

	Gap:5	Gap:15	Gap:30	Gap:50	Gap:100
0	4.333333	4.692308	6.428571	10.520833	17.091837
1	6.333333	2.846154	5.178571	10.041667	14.704082
2	5.333333	13.307692	21.250000	25.750000	24.551020
3	2.000000	11.538462	12.142857	8.833333	9.163265
4	1.333333	8.076923	11.535714	14.645833	12.540816
5	5.333333	17.230769	22.428571	21.875000	16.387755

```
In [74]: FB = pd.DataFrame({'Gap:5':fb_gap5, 'Gap:15':fb_gap15, 'Gap:30':fb_gap30, 'Gap:50':fb_gap50, 'Gap:100':fb_gap100})
FB
```

executed in 42ms, finished 13:25:47 2020-10-14

Out[74]:

	Gap:5	Gap:15	Gap:30	Gap:50	Gap:100
0	0.031083	0.035674	0.051890	0.088006	0.149169
1	0.069425	0.030546	0.052337	0.091316	0.124905
2	0.038118	0.103091	0.170869	0.209447	0.198190
3	0.017316	0.108942	0.114510	0.082013	0.085342
4	0.010479	0.066834	0.097473	0.125138	0.106183
5	0.047094	0.162981	0.216415	0.210963	0.155640

```
In [75]: FB = pd.DataFrame({'Gap:5':fb_gap5, 'Gap:15':fb_gap15, 'Gap:30':fb_gap30, 'Gap:50':fb_gap50, 'Gap:100':fb_gap100})
```

```

In [75]: glu.to_dataframe('glu10', time_gap10, 'glu10', time_gap10, 'glu10', time_gap10, 'glu10', time_gap10, 'glu10', time_gap10)
executed in 44ms, finished 13:25:47 2020-10-14

Out[75]:
   Gap:5  Gap:15  Gap:30  Gap:50  Gap:100
0  4.725816  5.650051  8.053393  12.470799  19.849176
1  6.350853  3.562627  6.657434  13.288466  17.827690
2  5.597619  16.462078  24.834452  28.605798  27.065680
3  2.449490  13.716918  13.907860  11.011358  11.471349
4  1.825742  9.227884  13.238202  16.149690  14.067258
5  5.416026  18.757563  23.503799  23.116553  18.599429

In [76]: .DataFrame({'Gap:5':mape_gap5, 'Gap:15':mape_gap15, 'Gap:30':mape_gap30, 'Gap:50':mape_gap50, 'Gap:100':mape_gap100})
executed in 44ms, finished 13:25:47 2020-10-14

Out[76]:
   Gap:5  Gap:15  Gap:30  Gap:50  Gap:100
0  3.152333  3.657676  5.416582  9.402892  16.687821
1  6.708753  2.997509  5.413667  9.864004  13.746207
2  3.733897  11.190585  19.467499  24.221142  22.700530
3  1.709402  11.807833  12.392205  8.738206  9.132209
4  1.056410  6.992134  10.447777  13.577102  11.397967
5  4.826640  18.081855  24.602487  23.999033  17.393321

In [19]: # IOA.to_csv("~/Desktop/NCSA_genomics/Python - notebooks/GlucoCheck/Metrics/IOA.csv")
# FB.to_csv("~/Desktop/NCSA_genomics/Python - notebooks/GlucoCheck/Metrics/FB.csv")
# RMSE.to_csv("~/Desktop/NCSA_genomics/Python - notebooks/GlucoCheck/Metrics/RMSE.csv")
# MAPE.to_csv("~/Desktop/NCSA_genomics/Python - notebooks/GlucoCheck/Metrics/MAPE.csv")
# MAD.to_csv("~/Desktop/NCSA_genomics/Python - notebooks/GlucoCheck/Metrics/MAD.csv")
executed in 17ms, finished 22:08:20 2020-10-13

In [20]: # getting model metrics
executed in 9ms, finished 22:08:20 2020-10-13

In [ ]: 1

In [21]: print("Model loss on training set:")
print(np.mean(obj.model_history.history['loss']))
print("Model Accuracy on training set:")
print(np.mean(obj.model_history.history['accuracy']))
print("Model loss on validation set:")
print(np.mean(obj.model_history.history['val_loss']))
print("Model accuracy on validation set:")
print(np.mean(obj.model_history.history['val_accuracy']))
executed in 1.66s, finished 22:08:21 2020-10-13

Model loss on training set:
66.09112285241825
Model Accuracy on training set:
0.09277845
Model loss on validation set:

-----
KeyError                                Traceback (most recent call last)
<ipython-input-21-eac7f6462949> in <module>
      4 print(np.mean(obj.model_history.history['accuracy']))
      5 print("Model loss on validation set:")
----> 6 print(np.mean(obj.model_history.history['val_loss']))
      7 print("Model accuracy on validation set:")
      8 print(np.mean(obj.model_history.history['val_accuracy']))

KeyError: 'val_loss'

In [ ]: from matplotlib import pyplot
# plot train and validation loss
pyplot.plot(obj.model_history.history['loss'])
pyplot.plot(obj.model_history.history['val_loss'])
pyplot.title('model train vs validation loss')
pyplot.ylabel('loss')
pyplot.xlabel('epoch')
pyplot.legend(['train', 'validation'], loc='upper right')
pyplot.show()
executed in 47m 19s, finished 22:08:21 2020-10-13

In [ ]: # plot train and validation accuracy
pyplot.plot(obj.model_history.history['accuracy'])
pyplot.plot(obj.model_history.history['val_accuracy'])
pyplot.title('model train vs validation accuracy')
pyplot.ylabel('loss')
pyplot.xlabel('epoch')
pyplot.legend(['train', 'validation'], loc='upper right')
pyplot.show()
executed in 47m 19s, finished 22:08:21 2020-10-13

In [ ]:

```