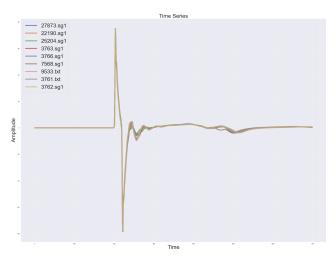
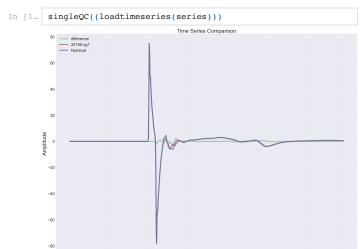
```
In [1... import pandas as pd import csv
             import matplotlib.pyplot as plt
             import numpy as np
import seaborn as sns; sns.set()
             from sklearn.linear_model import LinearRegression
             from scipy.fft import rfft, rfftfreq
In [1... def loadtimeseries(series):
                    dict_ts={}
for i in series:
    dict_ts[i]=[]
                           linecount=0
                          out=[0]*200
                           with open(i) as filein:
                                 spamreader=csv.reader(filein)
for x in spamreader:
                                        if linecount>24:
                                             out.append(float(x[0]))
                                        linecount=linecount+1
                                 for t in out:
                                        dict_ts[i].append(t)
                    return dict ts
In [1... def plott(zz):
                    fig, ax=plt.subplots(1,1)
                    plt.rcParams["figure.figsize"] = (40,30)
                    for r in zz:
                           x=[]
                           for i in range(len(zz[r])):
                   x.append(i)
plt.plot(x[:-3500],zz[r][:-3500], label=r,linewidth=f
plt.legend(loc='upper left',fontsize=30)
ax.set_xlabel('Imine', fontsize=40)
ax.set_ylabel('Amplitude',fontsize=40)
plt.title('Time Series', fontdict=None, loc='center', pace left very labelsize=20)
                    plt.rc('xtick',labelsize=30)
plt.rc('ytick',labelsize=30)
plt.legend(loc="upper left",fontsize=40)
                    plt.show()
                    return
In [1... def singleQC(zz):
                    for s in zz:
                           if n==0:
                                 anom=s
                           else:
                                 x=[]
                                 for i in range(len(zz[s])):
                                       x.append(i)
                                 fig, ax=plt.subplots(1,1)
                                 plt.rcParams["figure.figsize"] = (40,30)
                                 diff=[]
                                 for i in range(len(zz[s])):
                                       diff.append(zz[anom][i]-zz[s][i])
                                diff.append(zz[anom][i]-zz[s][i])
plt.plot(x[:-3500],diff[:-3500],"-g", label="diff;
plt.plot(x[:-3500],zz[s][:-3500],"-r", label=s,!;
plt.plot(x[:-3500],zz[anom][:-3500],"-b", label=
ax.set_xlabel('Time', fontsize=40)
ax.set_ylabel('Amplitude',fontsize=40)
plt.rc('xtick',labelsize=30)
plt.rc('ytick',labelsize=30)
plt.legend(loc="upper left",fontsize=30)
plt.title('Time Series Comparison', fontdict=None
plt.show()
                                 plt.show()
                                 print ("the correlation is " ,np.corrcoef(zz[anor
print ('peak-2-peak Nominal --> ' + str(max(zz[at
print ('peak-2-peak '+ s +' --> ' + str(max(zz[s]
                                 print ('peak-2-peak difference % --> negative
                                      + str((((max(zz[anom])+abs(min(zz[anom])))-(max
                                 print ("")
print ("")
                           n=n+1
                    return
In [1... series=['27873.sg1','22190.sg1','25204.sg1','3763.sg1','3766
In [1... plott(loadtimeseries(series))
```

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Timeseries_qc 10/09/21, 05:24



Let's analyze one by one

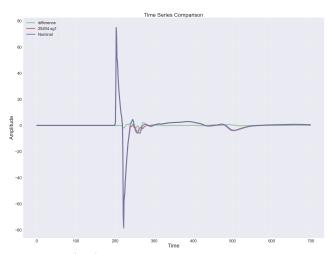


the correlation is 0.9982482294743984

peak-2-peak Nominal --> 153.27775

peak-2-peak 22190.sg1 --> 151.71439

peak-2-peak difference % --> negative 1.0199523414194094



the correlation is 0.9974651816696463

peak-2-peak Nominal --> 153.27775

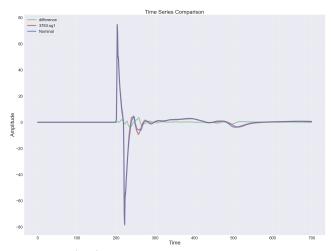
peak-2-peak 25204.sg1 --> 151.69946

peak-2-peak difference % --> negative 1.0296928288678624

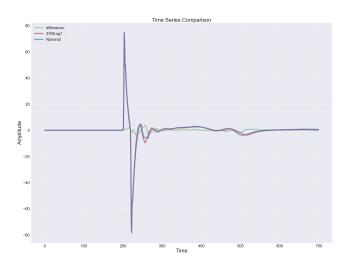
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Timeseries or

10/09/21, 05:24



the correlation is 0.99750014521467
peak-2-peak Nominal --> 153.27775
peak-2-peak 3763.sg1 --> 151.55373
peak-2-peak difference % --> negative 1.1247685981820557

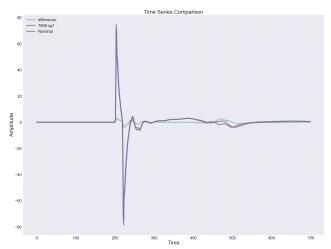


the correlation is 0.9964580242088446
peak-2-peak Nominal --> 153.27775
peak-2-peak 3766.sg1 --> 150.68613
peak-2-peak difference % --> negative 1.6907998714751529

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Timeseries or

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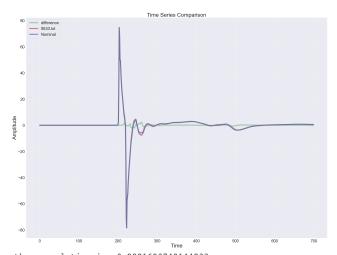


the correlation is 0.9978688761246371

peak-2-peak Nominal --> 153.27775

peak-2-peak 7568.sg1 --> 146.3600799999998

peak-2-peak difference % --> negative 4.513159933519389



the correlation is 0.9991693749144833

peak-2-peak Nominal --> 153.27775

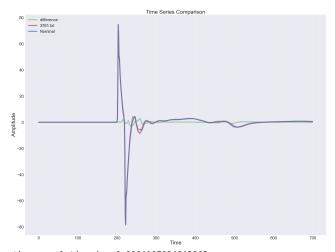
peak-2-peak 9533.txt --> 152.65571

peak-2-peak difference % --> negative 0.4058253725671197

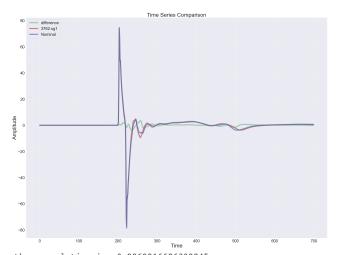
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Timespries or

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the correlation is 0.9984107324945865 peak-2-peak Nominal --> 153.27775 peak-2-peak 3761.txt --> 152.25617 peak-2-peak difference % --> negative 0.6664894285047896



the correlation is 0.9969916686309245

peak-2-peak Nominal --> 153.27775

peak-2-peak 3762.sg1 --> 151.0988899999998

peak-2-peak difference % --> negative 1.4215109498932588

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