Code:

 $sed 's/,/g; s/-,-,//; s/,+0000]//; s/\[[//; s/Apr/04/;s/\04\//-04-/; s/:/,/; s/"//g; s/,HTTP\[[0-9,X].[0-9,X]]//;s/GET,//' wc_day91_1.log > wc_day91_1.csv \\ log_df = pd.read_csv("/home/datascience/wc_day91_1.csv", names=['ClientID', 'Date', 'Time', 'URL', 'ResponseCode', 'Size'], na_values=['-'])$

Code for running commands from Q3

is200df = log_df[log_df['ResponseCode'] == '200']
finaldf = is200df[is200df['URL'].str.endswith('jpg') | is200df['URL'].str.endswith('jpeg') |
is200df['URL'].str.endswith('gif')]
finaldf['Size'].mean()
finaldf['Size'].std()

Result from running commands from Q3

3219.9428340117138 6302.9825349485855

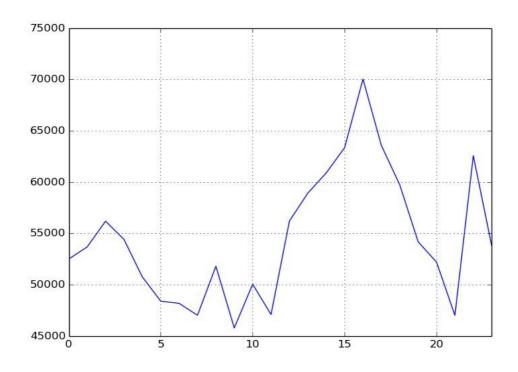
Conclusion: The results are different from the previous sample; i.e. the one in Q3.

Code for running commands from Q4

log_df['DateTime'] = pd.to_datetime(log_df.apply(lambda row: row['Date'] + ' ' + row['Time'],
axis=1))

hour_grouped = log_df.groupby(lambda row: log_df['DateTime'][row].hour)
hour_grouped.size().plot()
show()

Result from running commands from Q4



Conclusion : The structure of the graph is similar to that of previous sample; i.e. the one in Q4.