Data Analytics Course: 18-899

Spring 2019
Carnegie Mellon University

- Read the file
- Generate dates/timestamps
- Plot the wind generation against the dates
- Is there any there evidence of intra annual seasonality

- Calculate the change in wind generation over time as a percentage of the maximum generation
- Plot it against time
- Is there any evidence of seasonality?

- Ramps \rightarrow r(t,d) =100*[x(t+d)x(t)]/max(x)
- Calculate the ramps where d = 1
- Separate them into positive and negative ramps
- Normalize and sort them
- Plot them
- Similary plot the normal distribution

- Challenge: balancing supply and demand
- investigate the variability in wind generation over different timescales
- Timescales: 1h, 2h,, 24h
- Use the percentile analysis on the ramps
- Plot the results
- What did you learn

Question 4 - continued

- Some hints:
- → Calculate the ramps for a timescale eg: d=1, d=2 or d=4
- → Calculate the percentile of the ramps distribution at 1%, 5%, 95%,99% functions: prctile(matlab), numpy.percentile(python)
- → Repeat the process for all timescales
- → Plot all the four percentiles as a function of the timescale

- Calculate the autocorrelation
- Remember you still have Nan values
- Hint: xcorr, google the nanautocorr function somebody implemented it, or google any other new function
- Comment on the structure of the autocorrelation function

- Calculate Autocorrelation of change in wind generation for lags over 10 days
 - Calculate the change in wind generation
 - Calculate the autocorrelation with lags of 10days(240 hours)
- Plot it
- horizontal lines to detect statistically significance values (p<0.05)
 - Corresponding value can be calculated from the normal distribution
 - plot it for every value
 - → question on this post on piazza or skype us
- Is there any evidence of diurnal seasonality?
- Might it be more appropriate to model the change in wind generation than the wind generation?

- Remember: You might still be having NaN values, use function that handle them
- The variance ratio test will be used to investigate the structure of the wind generation timeseries
- Hint: vratiotest
- What are the returns of the function, doc it
- Can the null hypothesis of a random walk be rejected?
- Test mean reversion
- Is there evidence of either mean-reversion of mean aversion?

- For each n
- Calculate the simple moving average(function: tsmovavg)
- Calculate the mean absolute error between the simple moving average and the real wind power
- For which n, do you obtain the minimum error
- Is there a simple benchmark that improves on the persistence benchmark?

- For each n
- N ranges from 1 to 24 (one hour to one day)
- Calculate the persistence of n → X_predicted(t) = X(t-n)
- Calculate the mean absolute error between the simple moving average and the real wind power
- Plot MAE as a percentage of the maximum generation for the persistence benchmark.

- doc/google the ARIMA model
- Find parameters that it takes
- Add the model to your environment, if you don't have it already
- Loop through a range of parameters to find the optimal parameters
 - pass the parameter to the arima model
 - estimate
 - calculate the AIC and BIC from the estimation
- find if the current value improve(small AIC and BIC are better) on the previously selected
- What are the parameters that give the best model?