# Bunched Random Forest in Monte Carlo Risk Simulation

Eina Ooka May, 2017



**R/Finance 2017** In Chicago



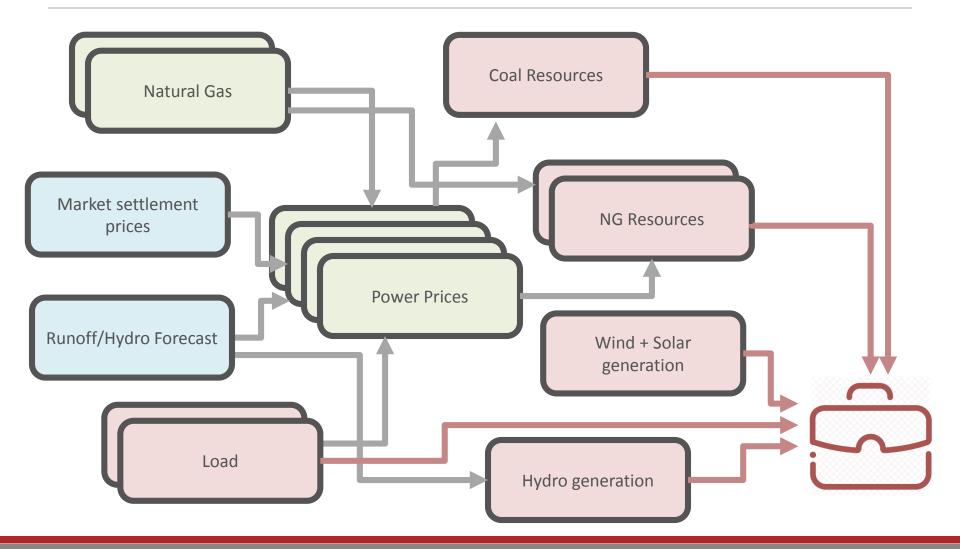
# Power Utility Industry

- The Energy Authority serves public utilities nationwide for trading and analytics.
- Mid-term (1 month 5 years) portfolio management.
- Stochastic simulation models for energy and gas market.



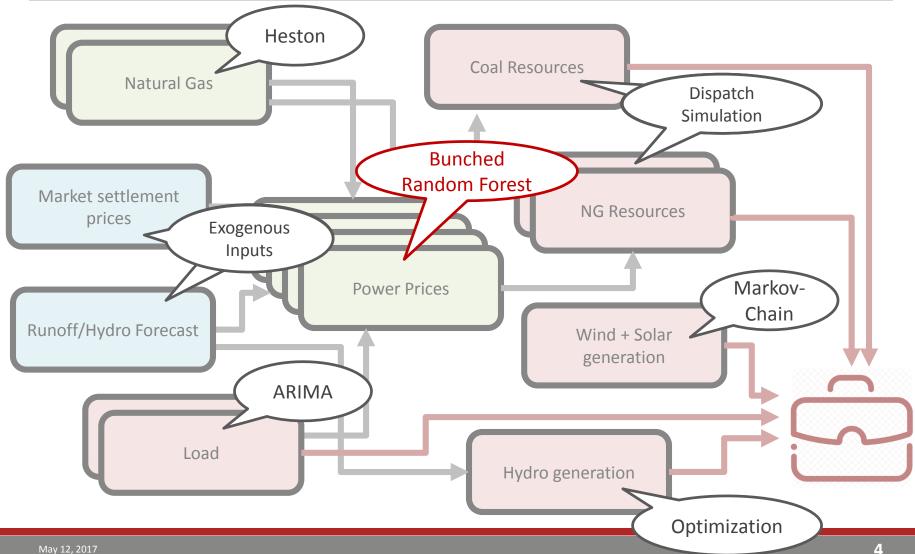


## MC Simulation Approach





### **MC** Simulation Approach

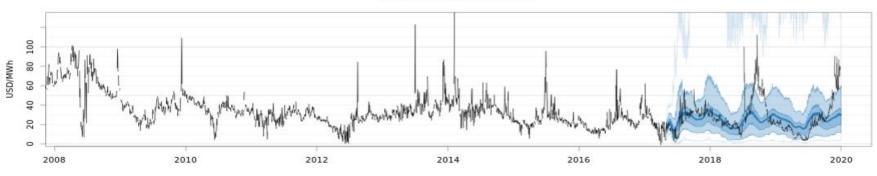




# Power Price TS Characteristics

- Autocorrelation
- Seasonal and weekly shapes
- Volatility & Heteroscedasticity
  - Seasonal and weekly variability
- Multivariate cross-correlation and non-linear dependency
  - NG, load, regional hydro and other variable generation

- Non-normal distributions
  - fat tails
  - Extreme peaks and drops
- Negative prices
- Consistency with market expectations
- Consistency between monthly, daily and hourly data.



#### MidC On-Peak Power Price

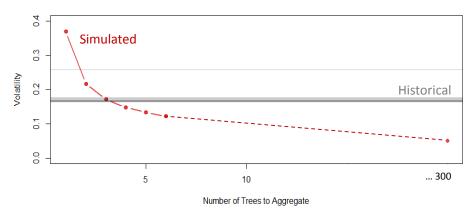


### **Bunched Random Forest**

- Traditional Approach for MC
  - Price = f(...) + Error(...)
- Random Forest
  - Aggregation of a few hundred trees moderate values too much → Low volatility
  - a single tree encompasses too little predictability → High volatility
- Bunched RF in MC
  - Aggregating a selected number of trees for each Monte Carlo iteration
  - Achieve plausible volatility in each MC series, while maintaining the

same predictability as RF in the whole MC simulation.

- Implementation
  - randomForest::predict.randomForest() includes individual tree outputs.
  - After figuring out the bunching number based on historical volatility, sample and aggregate suitable number of individual tree outputs.



### Results



historical

max 80 100 95th 75th USD/MWh mean 60 iteration 25th 40 5th 20 min 0 2010 2012 2016 2018 2008 2014 2020 10 Simulate \_\_\_\_\_ 1 Yr \_\_\_\_ 2 Yr \_\_\_\_ 3 Yr **Price Distribution** Historical 8 . Historica Simulated Density 90 2007 2008 5 2012 2013 2014 2015 2016 2017 02 Autocorrelation 2 0 20 40 60 80 100 120 Lags (Days) MidC On-Peak Power Price (USD/MWh) Seasonality 8 2 0 2009 4 8 Weekly Shape 2011 ş 2012 4O 2013 2014 2 Я Non-linear 2015 AidC On dependency on 2016 2 2017 0 regional hydro Simulated 10

12

2

Mid-C Power Price (Historical + Simulated)

10000

12000

14000

8000

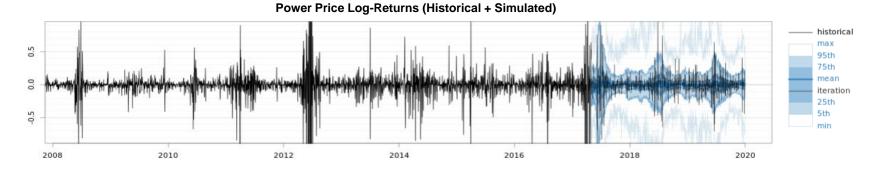
4000

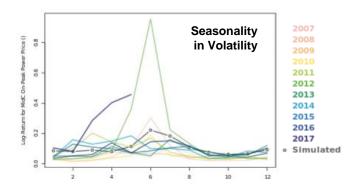
6000

Sal

## Results



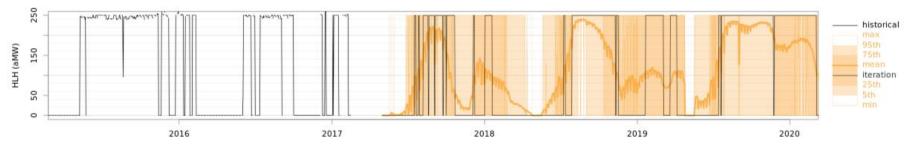




#### Scoring

- Pinball-loss scoring for stochastic forecasting
- Need to achieve all time series characteristics.
  - Volatility, seasonality, moments, ...
- A visual inspection with heuristic reasoning is sometimes necessary.
- Backcast and feed the generated power prices to Natural Gas dispatch model, and look at pinball-loss scoring on generation outputs.

#### Natural Gas Resource Dispatch (Historical + Simulated)







More on utility portfolio modeling on my poster from UseR 2016: https://github.com/ einaooka/useR2016

#### Thank you!

#### Contact: eooka@teainc.org

! The Energy Authority is looking for an intern.