TANGENT WORKS ADVANCED FORECASTING



Automatic Model Building for Timeseries in Energy Industry

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TIM Tangent Information Modeller

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- GefCom 2014, 2017 results & Summary



Birthday Problem

- $366 \rightarrow 100\%$
- 120
- 75
- 23
- 6



Time-series Problems in Energy Industry

- Electricity Load (aggregated and individual)
- Technical Losses
- System Imbalance
- Gas Consumption (District Heating)
- District Cooling

- Wind Production
- Solar Production
- Electricity Price







feature engineeringmodel selection

- tedious
- costs money



SAS, SPSS, R, ML libraries, Matlab



Which features and how many?

y(k) = f(Temp(k-3), Temp(k-22) * Wind(k-1), y(k-24))

→ NN, SVM, MARS, LASSO, RF, ... ?



$$y(k) = f(Temp(k-3), Temp(k-22) * Wind(k-1), y(k-24))$$

Temp(k-1), Temp(k-2), ..., Temp(k-24) Wind(k-1), Wind(k-2), ..., Wind(k-24) y(k-1), y(k-2), ..., y(k-24)

24 + 24 + 24 + 24*24 = 24*27 = 648

Temp(k-1), Temp(k-2), ..., Temp(k-96) Wind(k-1), Wind(k-2), ..., Wind(k-96) y(k-1), y(k-2), ..., y(k-96)



Information Geometry: Let Your Data Speak

Prof. Akaike – 1974 proposed AIC, one of the first information criteria

Prof. Amari – cca 1998 proposed to use differential geometry for non-parametric inference (e.g., training of neural-nets)

Prof. Konishi – 2004 proposed GBIC and general way of inferring new information criteria Book: Information Criteria and Statistical Modeling

IG is taking advantage of using manifolds and geometrical structures to understand data and generate meaningful and interpretable models.



Data intake

Synthesis of features based on a set of heuristic rules

Variable selection is carried out in Euclidean space using Bayesian inference

Model Complexity is estimated using a customized IC

Model is generated



Data Science Process – Industrial View on Data Science

$$y(k) = f(Temp(k-3), Temp(k-22) * Wind(k-1), y(k-24))$$

Business/Industrial users value models that are

- Transparent
- Easy to deploy and use
- Easy to maintain
- Have accuracy which is good enough



Data Science Process – Industrial View on Data Science

$$y(k) = f(Temp(k-3), Temp(k-22) * Wind(k-1), y(k-24))$$

Over-engineered models of e.g., Kaggle competitions are of limited use in industrial practice

- Such models are often (over) optimized for a single dataset
- Neural-net combined with genetic algorithm may have the best accuracy but ...





- TIM does it all automatic
 - ✓ Zero degrees of freedom
 - ✓ No feature engineering
 - ✓ High speed computing on standard hardware
- Accuracy often matches or outperforms manually build models
- TIM models are transparent, bring extra knowledge and confidence



Live Demonstration of TIM



Large-scale Forecasting Systems & Why Automation Matters

Example 1 (totaling in 8*2 + 5*2 + 3*1 = 29 models)

- 5 electricity load asset and 3 solar farms = 8 assets in total
- Day-ahead and week-ahead scenarios for each asset + nowcasting (intra-day) for the solar assest
- Dynamic data availability for a single asset e.g., historical data of yesterday may not come
- Models are re-built on-demand



Large-scale Forecasting Systems & Why Automation Matters

Example 2 (totaling in 1800*2 + 1800*2 = 7200 models)

- 1800 electrical assets
- Day-ahead and week-ahead model for each asset
- Dynamic data availability for any asset
- Automatic model re-building to account for structural changes



Julia Language

Walks like Python runs like C

- A single platform for prototyping and production \rightarrow significant gains in efficiency of product development
- Vectorized code runs equally fast as de-vectorized
- Vectorization on a single thread level
 - SIMD out of the box
 - Direct calls to BLAS
- Distributed parallel computation (multiple threads)









TIM Architecture legend: D - data, M - model, P - prediction —— Training flow —— Testing flow



Prototype Clients



Technology Partners



GefCom 2014 ex-post results

- Electricity Load track: 3rd place out of about 300 teams
- Electricity Price trak: 2nd place out of about 250 teams
- Wind Generation track: 7th place out of about 250 teams
- Solar Generation track: 9th place out of about 250 teams

Andritz Hackathon 2017

1st place out of 7 ML companies



GefCom 2017 current ex-ante results

• Electricity Load track: 2nd place out of about 73 teams



Summary

- Fully automated model building
- Accuracy often outperforms manually build models
- Quick insights into a time-series of interest
- Tedious model building is an option not necessity



Thank you.

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