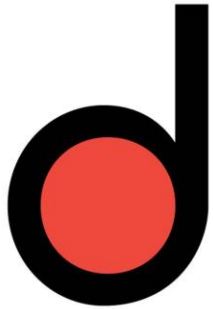


# Electricity Market Trading

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In consultation with Delaware  
Consulting Group



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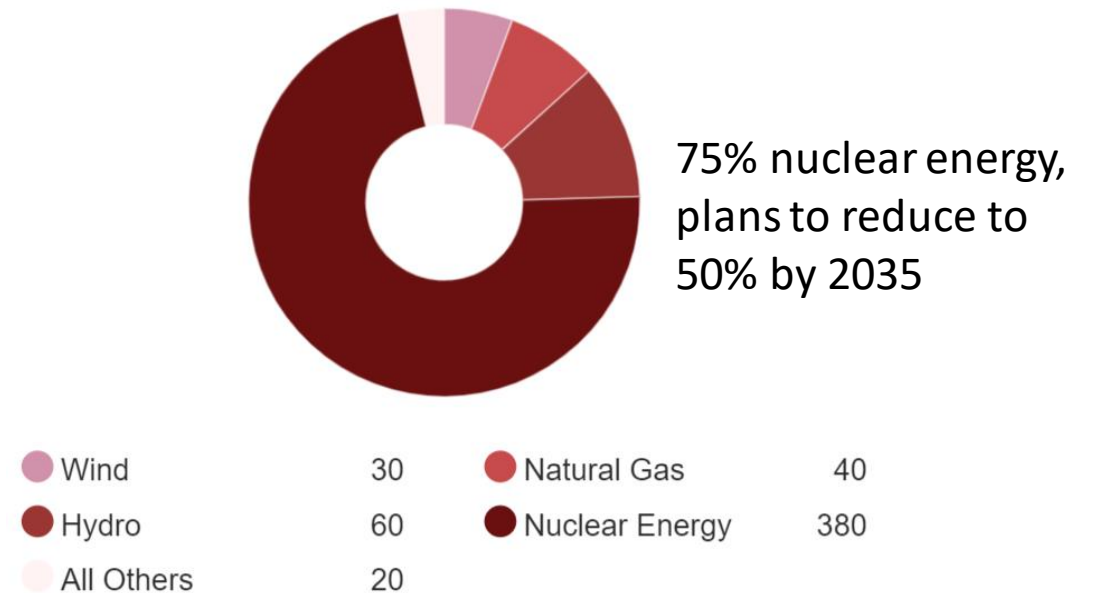


# Introduction

## French electricity market:

- Trading of Electricity happens on **epexspot**
- EPEX is a market for power spot trading in Europe operating in Germany, **France**, the U.K., BENELUX, Austria and Switzerland
- Market participants include energy generators, energy suppliers and traders (who help to ensure market liquidity)

French Electricity Sources, 2017



RTE France, 2017 generation data

# Market Design

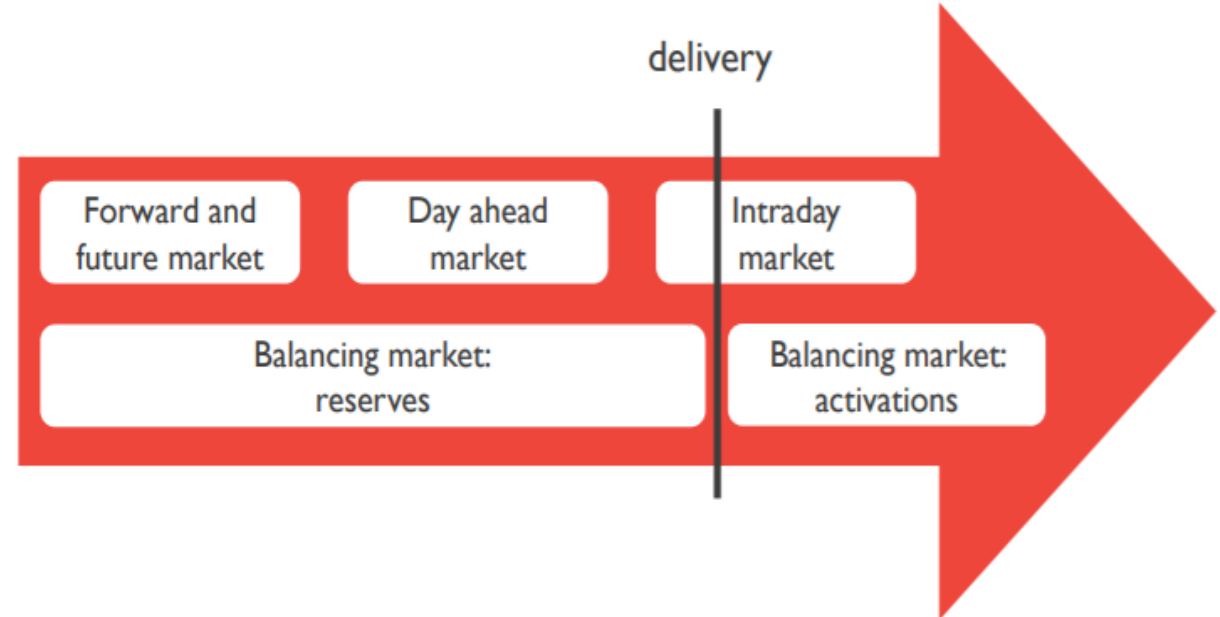
## DayAhead:

- One day before delivery
- Submit orders until 12 PM day ahead
- Needs to be in balance at end of day
- Coupled with other markets zones

## IntraDay:

- Traded on delivery day
- Continuous Market
- Corrects for shifts in day ahead nominations

### Market design



# Managerial Problems

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1. How can we trade between the day ahead and the intraday market?
2. Hard to predict price fluctuation (many factors have influence)
3. Decision between DayAhead and IntraDay can be very costly if not analyzed closely
4. Must develop a model and strategy to **maximize profit**



# Data Sources

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1 **epexspot**  
Electricity market data

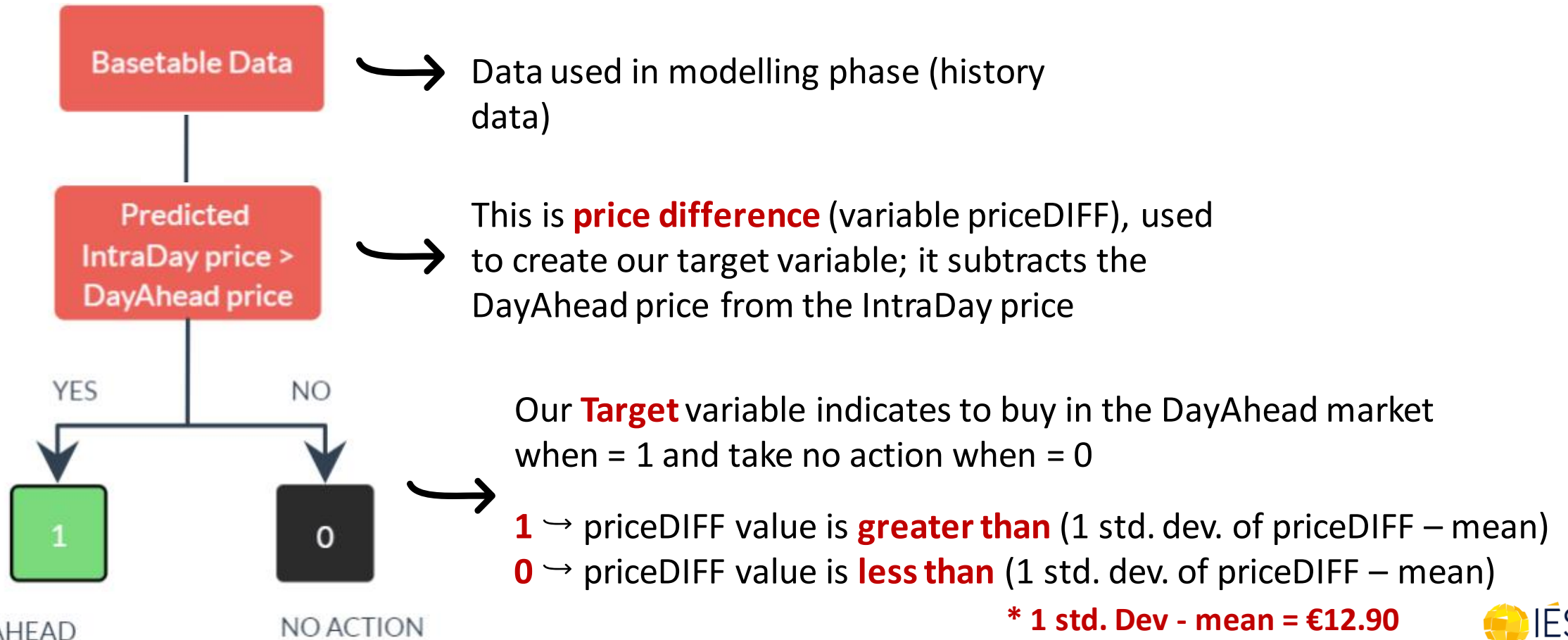
2 **WORLD WEATHER ONLINE**  
Weather data

3 **NOAA** NATIONAL CENTERS FOR ENVIRONMENTAL INFORMATION  
NATIONAL OCEANIC AND ATMOSPHERIC ADMINISTRATION  
Weather data

4 **PUBLIC HOLIDAYS GLOBAL**  
Calendar and holiday data

5 **entsoe** European Network of Transmission System Operators for Electricity  
Electricity market data

# Trading Strategy



# Trading Strategy

Our focus: buying in the **DayAhead** market

**Profit calculation** = *Profit with model* – *Profit without model*



Total net **priceDIFF** of positive predictions (April – Dec 2018) - (trading fee \* # of positive predictions April – Dec 2018)



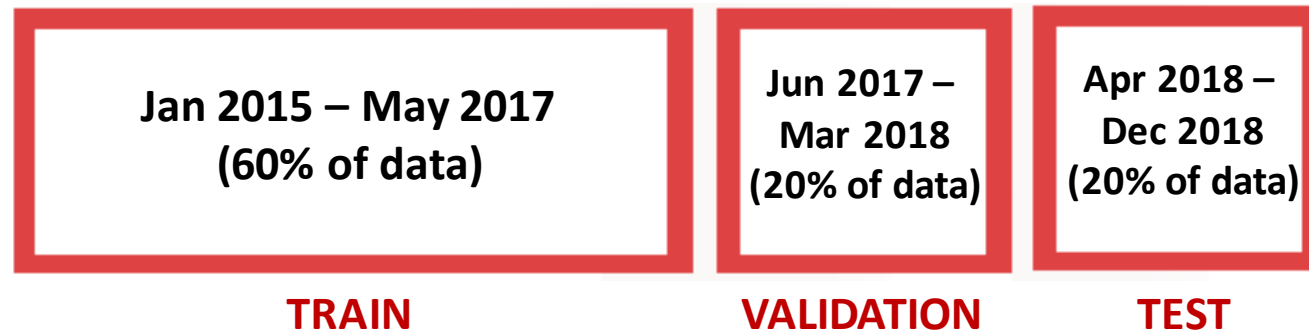
Total net **priceDIFF** of all days (April – Dec 2018) - (trading fee \* # of days April – Dec 2018)



# Methodological Approach

## 1. Data specifics

- Gathered external data (weather, weekends, holidays)
- Evaluated 4-year period (Jan 2015 – Dec 2018)



## 2. Modelling specifics

- Classification (predicting price difference increase/decrease)
- Logistic regression, XGBoost
- Evaluation metrics used : AUC and F1 Score

# Results Validation

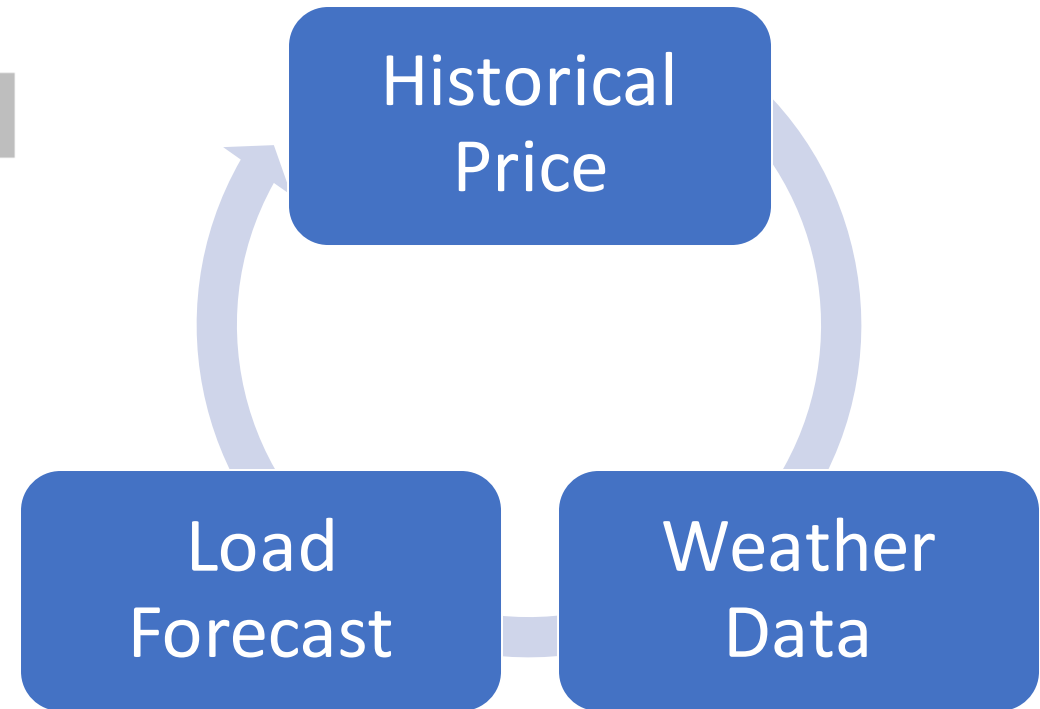
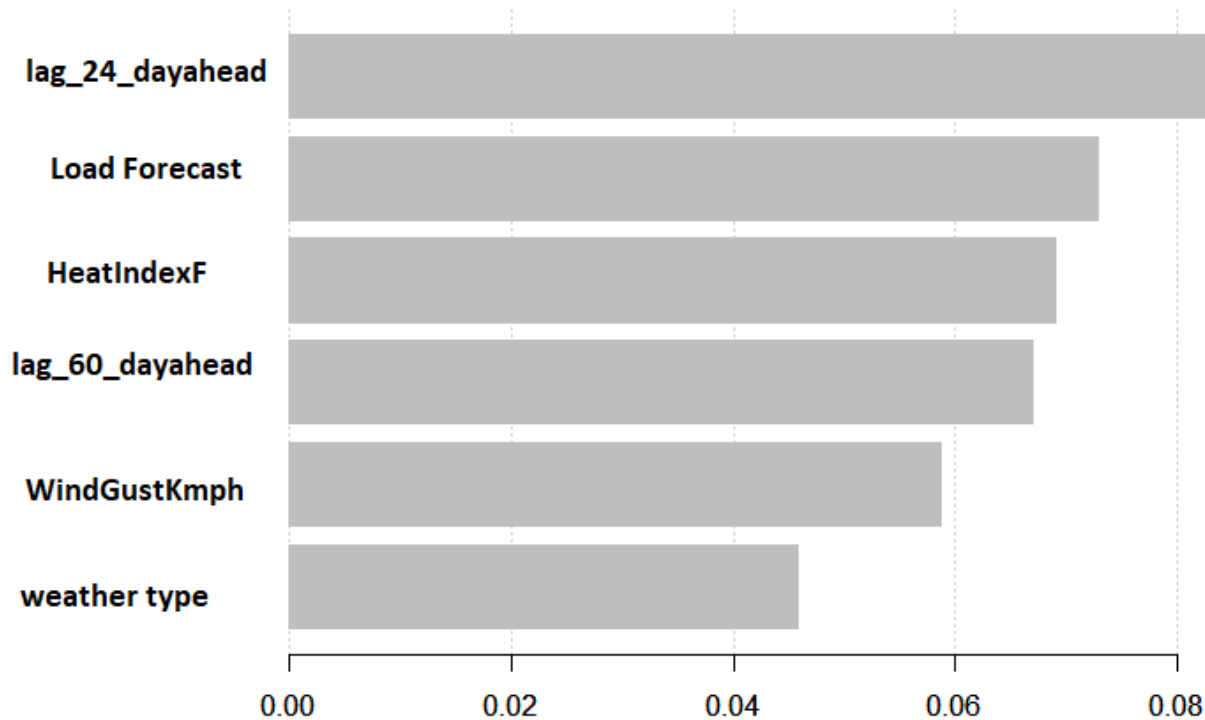
| Model Type          | Price diff Cut-Off | AUC |
|---------------------|--------------------|-----|
| Logistic Regression | 12.9               | .59 |
| XGBoost             | 12.9               | .52 |

\*Higher AUC = better accuracy in predicting price increase/decrease

Therefore, we selected **LOGISTIC REGRESSION** as our best model!

# Variable Importance

## Most important Variables:



# Business Impact

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Profit with model € **1,256**

Profit without model € **363**

**Monetary gain from model** € **893**

## Assumptions:

- Each time buying 1 unit (1MWh)
- Trading fee = €0.07 per transaction

Therefore, trader can expect to increase profits in the DayAhead market by **246%** from using the logistic classification model

# Business Impact

**Demo:** Further evaluation + visualization of data done through our **R Shiny APP**

It covers some of the following:

- If price simulator
- Wind energy forecast
- IntraDay/DayAhead price



[https://cyberchain.shinyapps.io/delaware\\_hackathon/](https://cyberchain.shinyapps.io/delaware_hackathon/)

Thank  
you !