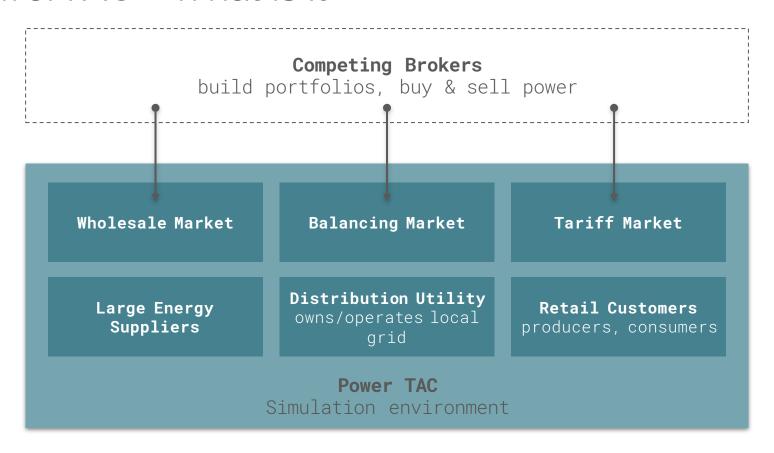
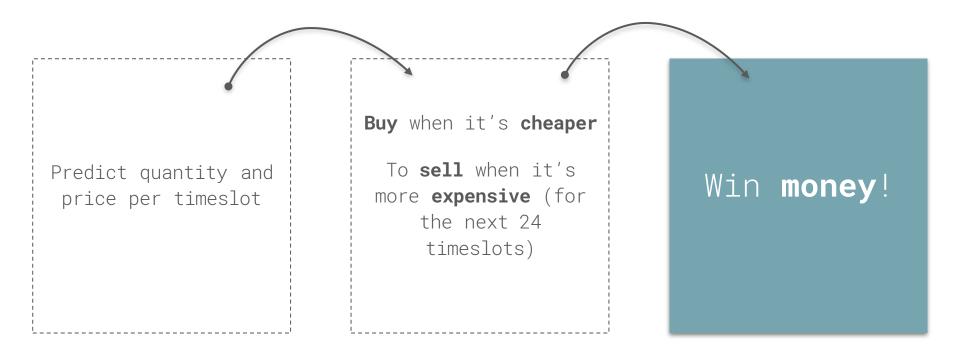
Agent TNE19

PowerTAC Broker

PowerTAC - What is it



Main Strategy (Only Wholesale)



Previous Data to Predict

Pick historic demand to predict the future demand

- Current Weather
 - a. Temperature
 - b. Wind speed
- 2. Time
 - a. Slot
 - b. Week of the day

To the current timeslot



For each of the previous 24 timeslots

- Cleared Trade: energy that was bought through 24 hours to be used in that timeslot
 - a. Cleared amount: quantity of energy that was cleared (bought) for that timeslot
 - b. Cleared price: average price/Mwh of the energy that was cleared for that timeslot
- 2. Weather Report
 - a. Temperature
 - b. Wind speed

For each of the next 24 timeslots

- Cleared Trades observed until now for the timeslot
 - a. Total quantity
 - b. Average price of all the cleared trades
- 2. Weather Forecast
 - a. Temperature
 - b. Weather

Prediction Model & API

When our agent requires a prediction, it sends a request to our REST Flask API in an asynchronous way, allowing the agent to keep trading in the market.

```
127.0.0.1 - - [27/May/2019 14:59:35] "POST /predict/price HTTP/1.1" 200 -
127.0.0.1 - - [27/May/2019 14:59:35] "POST /predict/energy HTTP/1.1" 200 -
```

At the moment our API uses a **linear regression model** to predict both amounts and prices. This is **highly modular** and one model can be swapped at any time by another one. In fact, we tested several models, such as **neural nets**, with **different parameters and train/test dataset sizes**. At any given moment this can be swapped without affecting the current agent architecture.

```
[7872 rows x 2 columns]
Mean Absolute Error: 0.18132006323021022
Mean Squared Error: 1.5458003629253028
Root Mean Squared Error: 1.2433022009653578
```

Current strategy Wholesale Money **Cumulative Money** Zoom 1m 3m 6m YTD 1y All default broker Sample broker Money Cumulative Zoom Im Im On YTD Iy All From Oct 25, 2009 To Oct 27, 2009 -150k 26. Oct 12:00 27. Oct 28. Oct 12:00 26. Oct 27. Oct 28. Oct 12:00 26. Oct 06:00 12:00 27, Oct 12:00 27. Oct 12:00

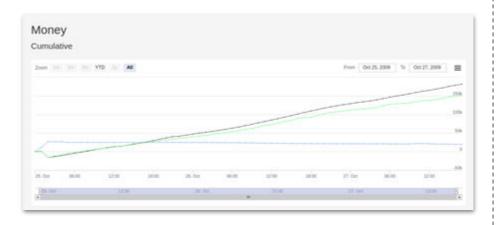
By only using an wholesale strategy as previously explained, we managed to win in the wholesale market.

However, we overall lost money. We conclude that the wholesale strategy by itself is not enough.

This is a **good indicator** that
PowerTAC is well designed and
prevents people (like us) from trying
to "cheat the system" and make money
without providing customers good
services.

Thus we found the need to complement with a retail strategy.

Our current **retail strategy** relies on **two stages**



- 1. Initially we set up tariffs at mean
 market price
- 2. Tit-For-Tat: After we identify several better tariffs from competitors, we choose the one to us that is the best and offer a similar one but with one or two parameters improved.

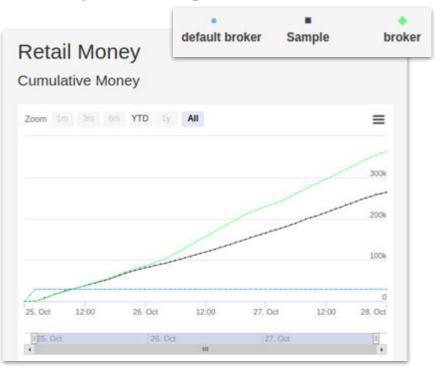
Usually this translates into a huge sign-up payment bonus to attract customers, along with a heavy withdrawal penalty and a more expensive periodic pay.

Initially, this cost us money but pays off in the long term.

It is clear that after a certain point the clients belong to us and we make money.

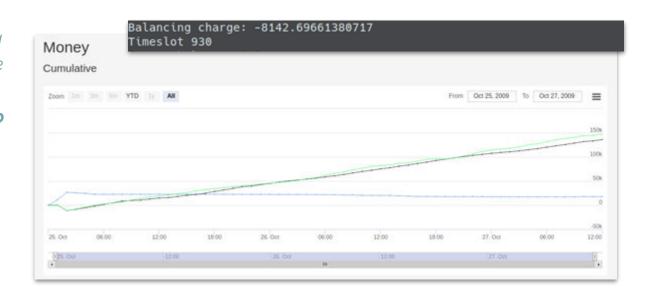
In fact, way more money than the sample-broker agent.





However, despite now making money and winning versus the sample-broker 7/10 test runs, we lose a lot of money due to DU (Distribution Utility) charges.

This translates into a **small win**, for such a big difference in the retail market.



Future improvements

1.

Clearly the **DU charges are the** condition holding us back.

As means to counter-attack this, as we've seen in our research, a future plan would be to implement Time-of-Use (TOU) tariffs in the retail market, that encourage customers to use energy when the DU will charge us less.

2.

Also, to avoid being so "unbalanced" relative to the market, in the future we would like to take into account our energy amount prediction plus the expected customer from our portfolio's consumption

n. Customers * customer average usage

thank you!