GOTO AARHUS 2023

#GOTOaar





Predicting
Optimal Supply
Temperature in
the Transmission
System

neuro

CTR







Readiness assessment



Knowledge

sharing







User-Story

As Centralkommunernes Transmissionsselskab
We want to predict the true supply temperature 12-36 hours in advance,
Such that we can ensure the best electrical efficiency as well as our
service obligation, at the lowest possible temperature



Why?

Opportunity to use the **best heating mixture**

Ensure a greener district heating, by reducing the return temperature



In the end, creating a cheaper district heating to end-users



What is
District Heating

What is District Heating?

1 unit fuel = 1/3 Electrical energy and 2/3 Thermal Energy



Overproduction - Wind Turbine



Waste Heat from Production Facilities



Waste Heat from Data Centres



Basically anything that produce waste heat





The Challenge District Heating

Heat can be **produced** from **different sources**



Heat can be distributed from different heat exchangers into the network



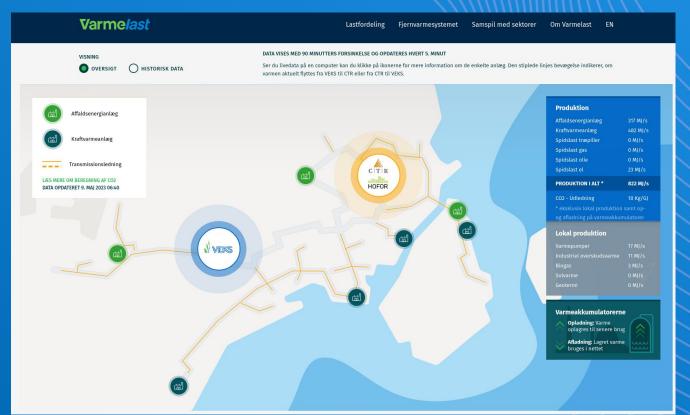
Centralkommunernes Transmissionsselskab



https://www.ctr.dk/vaerd-at-vide/om-ctr/



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Varmelast.dk



Natural hydraulic limitations

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It is **not** an option, to produce insufficient heat



Some customers require a **certain** supply temperature



Danish weather can provide you with **sunshine** and **+15 degrees** today

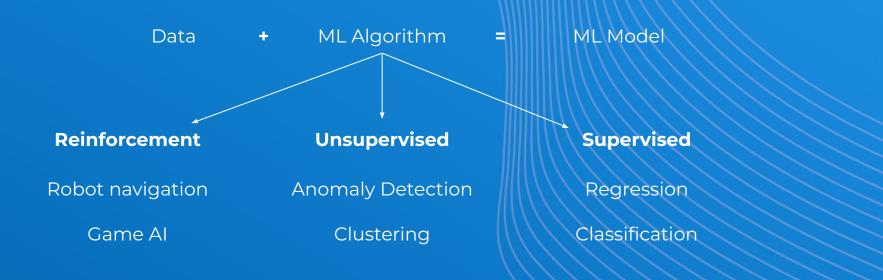


...and **snow** and **-2 degrees** tomorrow



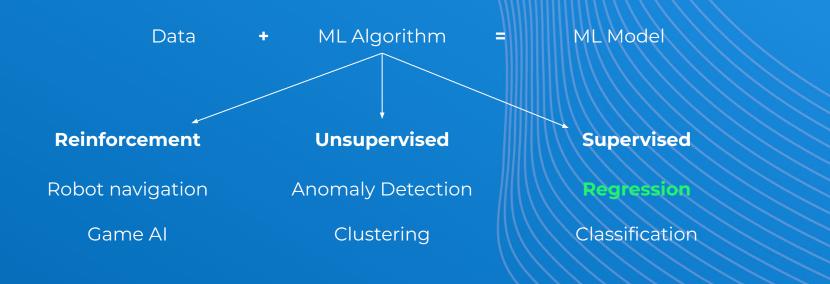
The Challenge Machine Learning

Predict Optimal Supply Temperature



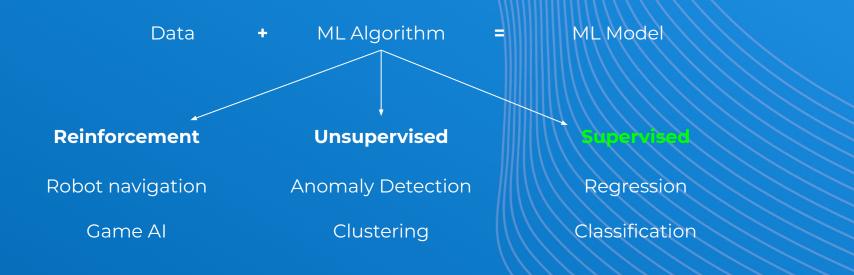


Predict Optimal Supply Temperature





Predict Optimal Supply Temperature





We train a Machine Learning algorithm on Historical Data



What we are looking for, must be represented in our dataset



District Heating will rather deliver too high temperature, than risk having a too low temperature



What is **Optimal?**

The Solution

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2 years of data

Pressure

Supply Temperature

Flow

Temperature (outside)

••

•



The data quality was high

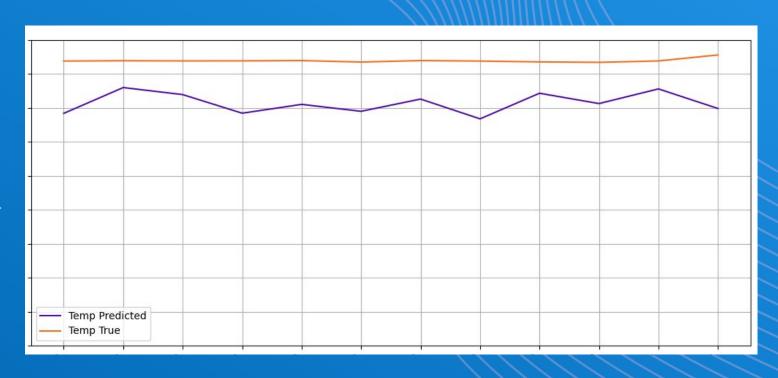
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Domain Experts in CTR, was able to provide crucial knowledge, for us to reduce the dataset significantly, and identify optimal supply temperature



Machine Learning Model #1 Predicting Optimal Supply Temperature

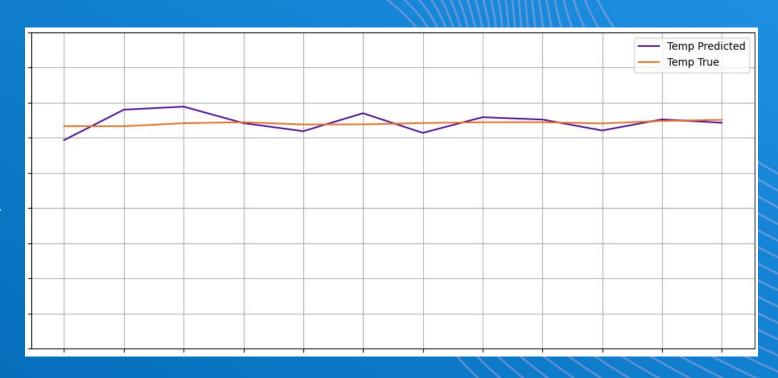




Timestamp







Timestamp



Experts from CTR were able to Validate weather the model's predictions were optimal or not

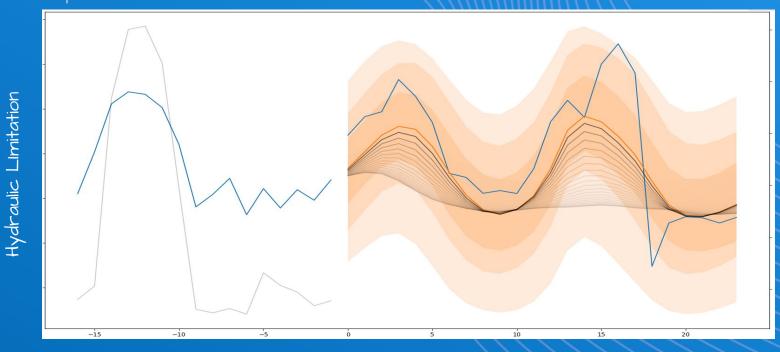


On average in a 14-day test-period, the model was able to optimize the supply temperature with 5.4 degrees



Machine Learning Model #2 Predicting hydraulic limitations





Timestamp



ML model 2

As Centralkommunernes Transmissionsselskab,

....

Such that we can ensure the best electrical efficiency as well as our service obligation, at the lowest possible temperature

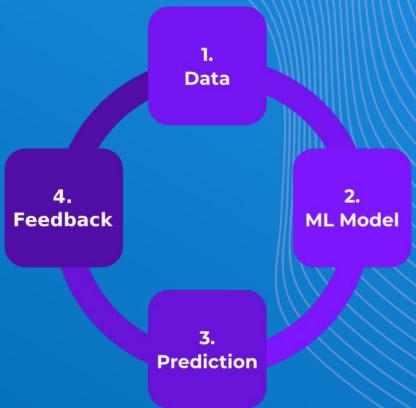


Key Takeaways

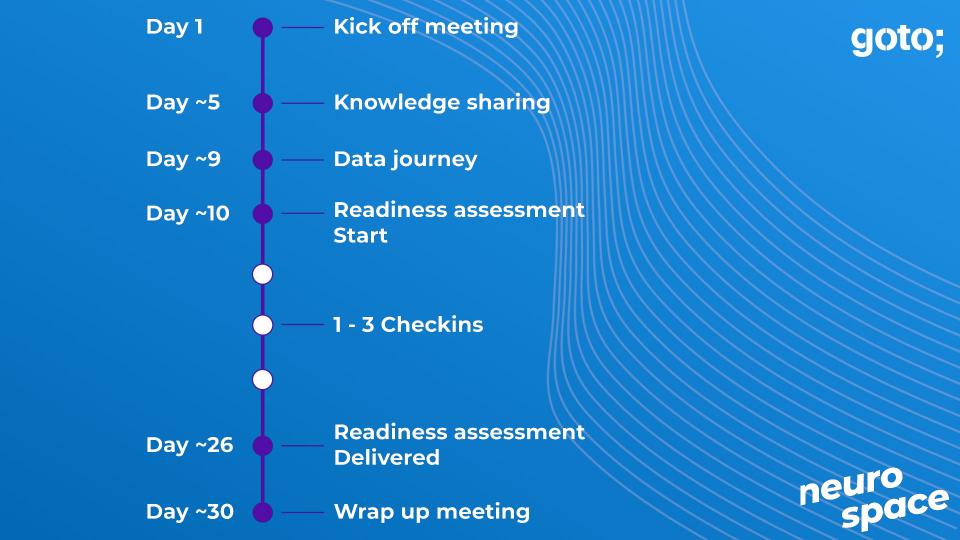
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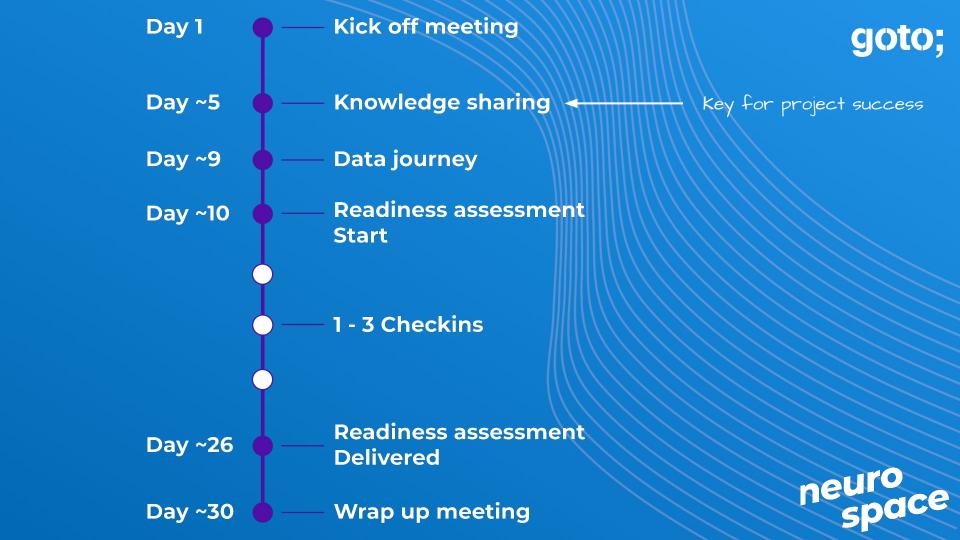
Right data != Big data Representative data

Machine Learning Flywheel





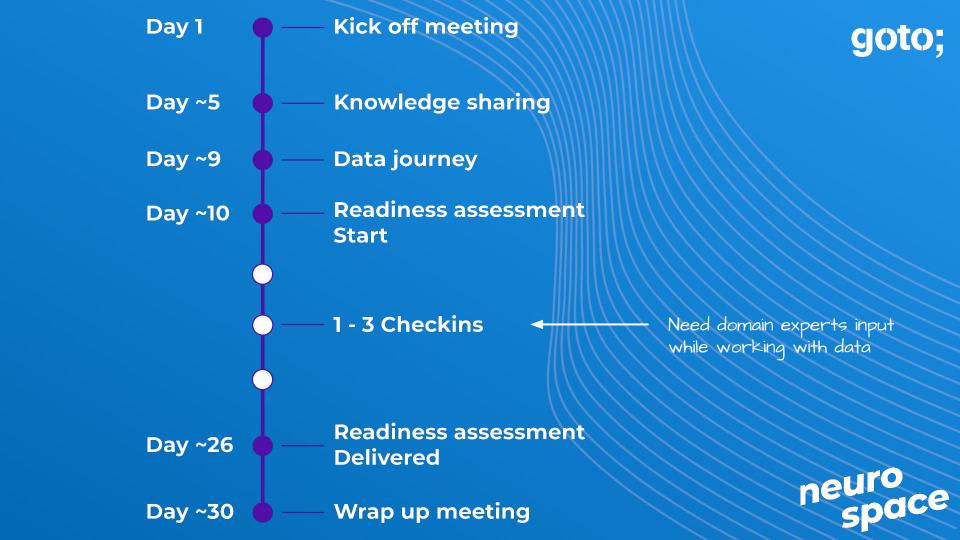


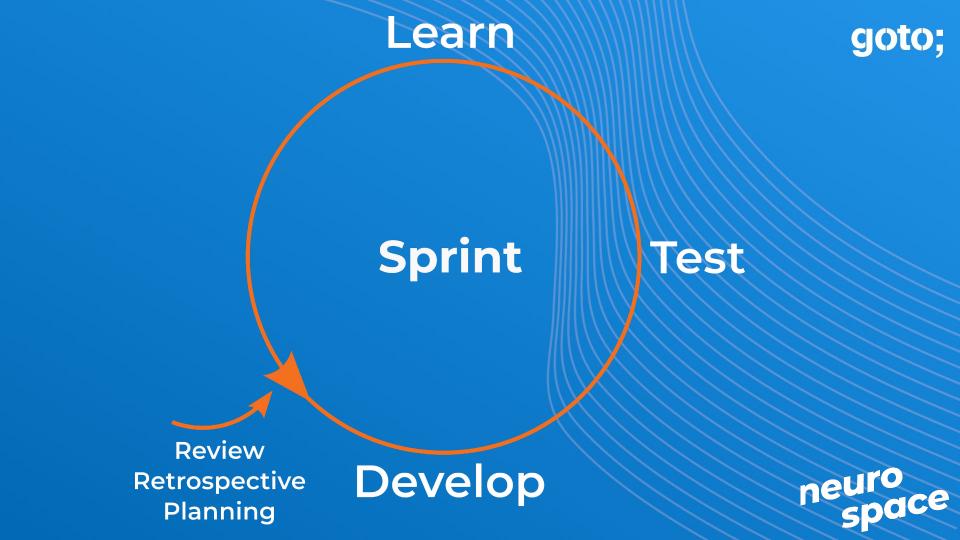




Domain experts

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