

# Microsoft Fabric Festival

## ML Engineering i

## Microsoft Fabric

# Anders Meged

## Data Engineer



# Agenda

- 1 **Today's Data Scientist**
- 2 **Solution Scenarios – Live Demos**
- 3 **Wrap up and benefits for Data Scientists**



### Personal Data

Age:

30

Job Title:

Data Scientist

Company:

EnergyCorp

# Kim

Data Scientist

Taking data to the next level

## Personal profile

Kim is a Data Scientist at EnergyCorp who works with:

- Building advanced analytic models
- Leveraging machine learning techniques to extract valuable insights from complex dataset.

While her background includes experience in traditional data analytics, Kim now focuses on pushing the boundaries of data science, exploring cutting-edge methodologies and algorithms to uncover patterns and trends that contribute to the overall success of EnergyCorp. She enjoys diving into the intricacies of data, applying statistical rigor, and deploying predictive models to solve challenging business problems.

## Skills

Python

SQL

Back-end Developer

# Needs

- Manipulate data ad-hoc
- Data exploration
- Train and Tracking ML models
- Dynamic and modifiable data foundation



# Pains



- Local environments – Lack of access across teams.
- Data cleaning and preprocessing
- Comparing Models
- Scaling and productionalizing
- Highlighting the benefits that models can provide.

# Solution Scenarios

*Starting point: Cleaning data for ML model*

## 1. Clean data for Model training

Using data wrangler to clean data.



## 2. Training models

Training ML models.



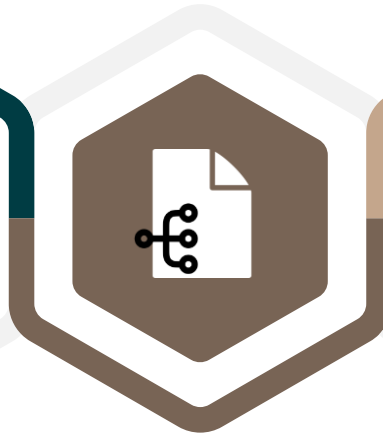
## 3. Comparing models

Comparing results of the different models.



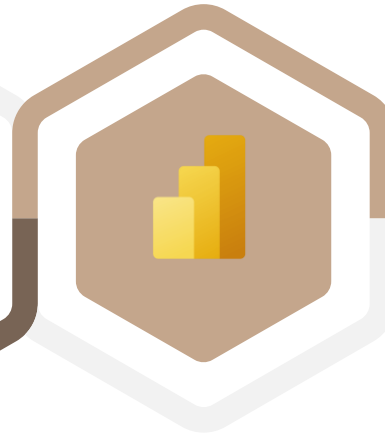
## 4. Productionalizing models

Make the model work on new data.



## 5. Visualizing

Visualizing the predictions of the model.



# Scenario 1: Prepare Data for Model Training



## Problem

- Kim is tasked with a cleaning data for training ML model on predicting energy production.
- The Data Scientist in the team uses different libraries and techniques for cleaning.



## Solution

- Kim builds a notebook report directly on the Fabric platform in the browser.
- Fabric notebook's "Data Wrangler" functionality combines coding and no code tools.
- This solution lowers the entry level for unifying a framework for cleaning preprocessing.



# Solution Scenarios

## Training ML models

### 1. Clean data for Model training

Using data wrangler to clean data.



### 3. Comparing models

Comparing results of the different models.



### 5. Visualizing

Visualizing the predictions of the model.



### 2. Training models

Training ML models.



### 4. Productionalizing models

Make the model work on new data.



# Scenario 2: Training ML Models



## Problem

- Kim needs to train ML models and keep track of the different Models.
- Kim's colleagues also benefits from seeing the models she trains, as they can take advantage of the knowledge.
- Currently it is difficult to share results as they don't have a platform for it.



## Solution

- Kim creates an experiment to keep track of the different model which is trained.
- Fabric enables centralisation of the model training, which solves Kim's and the team's problem.

# Solution Scenarios

## Comparing ML models

### 1. Clean data for Model training

Using data wrangler to clean data.



### 2. Training models

Training ML models.



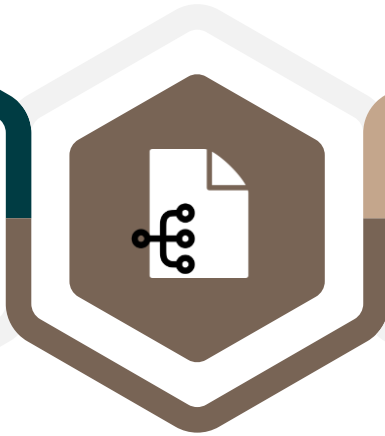
### 3. Comparing models

Comparing results of the different models.



### 4. Productionalizing models

Make the model work on new data.



### 5. Visualizing

Visualizing the predictions of the model.



# Scenario 3: Comparing Models



## Problem

- Kim would usually log the training of the ML models in a file to compare different configurations of the model. But it is time consuming as it is hard to keep track of which model performs the best.



## Solution

- Kim utilizes the list model to compare runs. Which is functionality of Fabric's experiment artifact. Kim is now able to keep track of the different runs and easily create or find models created from the runs.

# Solution Scenarios

## *Productionalizing ML models*

### 1. Clean data for Model training

Using data wrangler to clean data.



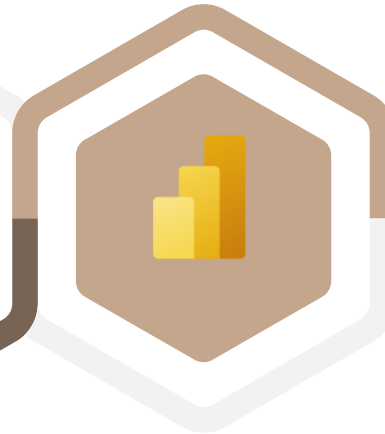
### 3. Comparing models

Comparing results of the different models.



### 5. Visualizing

Visualizing the predictions of the model.



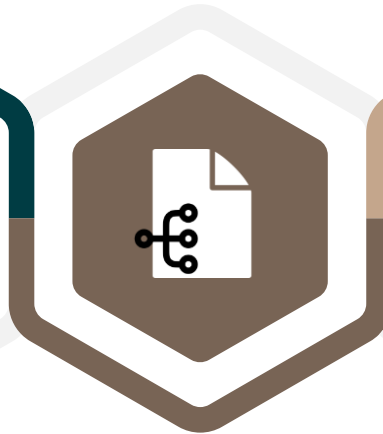
### 2. Training models

Training ML models.



### 4. Productionalizing models

Make the model work on new data.



# Scenario 4: Productionalizing Models



## Problem

- The Model is performing well enough for displaying it to management.
- Kim is tasked with applying the model to weather forecast data and making it accessible for visualization.



## Solution

- Kim creates a notebook with the model's API and predicts on the forecast data and loads the data to a table in a Lakehouse on the OneLake.

# Solution Scenarios

## *Vizualising ML models*

### 1. Clean data for Model training

Using data wrangler to clean data.



### 3. Comparing models

Comparing results of the different models.



### 5. Visualizing

Visualizing the predictions of the model.



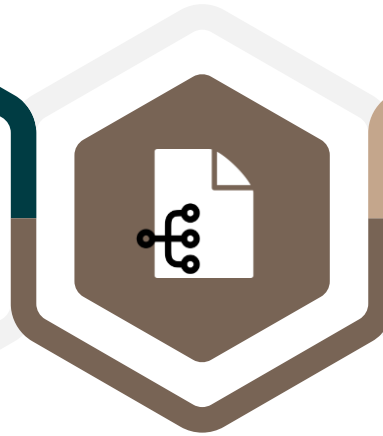
### 2. Training models

Training ML models.



### 4. Productionalizing models

Make the model work on new data.



# Scenario 5: Visualize Data



## Problem

- Kim needs to visualize the how the model can be used for management. The management is not technical profiles.



## Solution

- Kim creates a simple Power BI report where she visualizes the predictions on future energy production.



# Solution Wrap Up

*Starting point: ML Model delivered by Data Scientist*

## 1. Clean data for Model training

Using data wrangler to clean data.



## 3. Comparing models

Comparing results of the different models.



## 5. Visualizing

Visualizing the predictions of the model.



## 2. Training models

Training ML models.



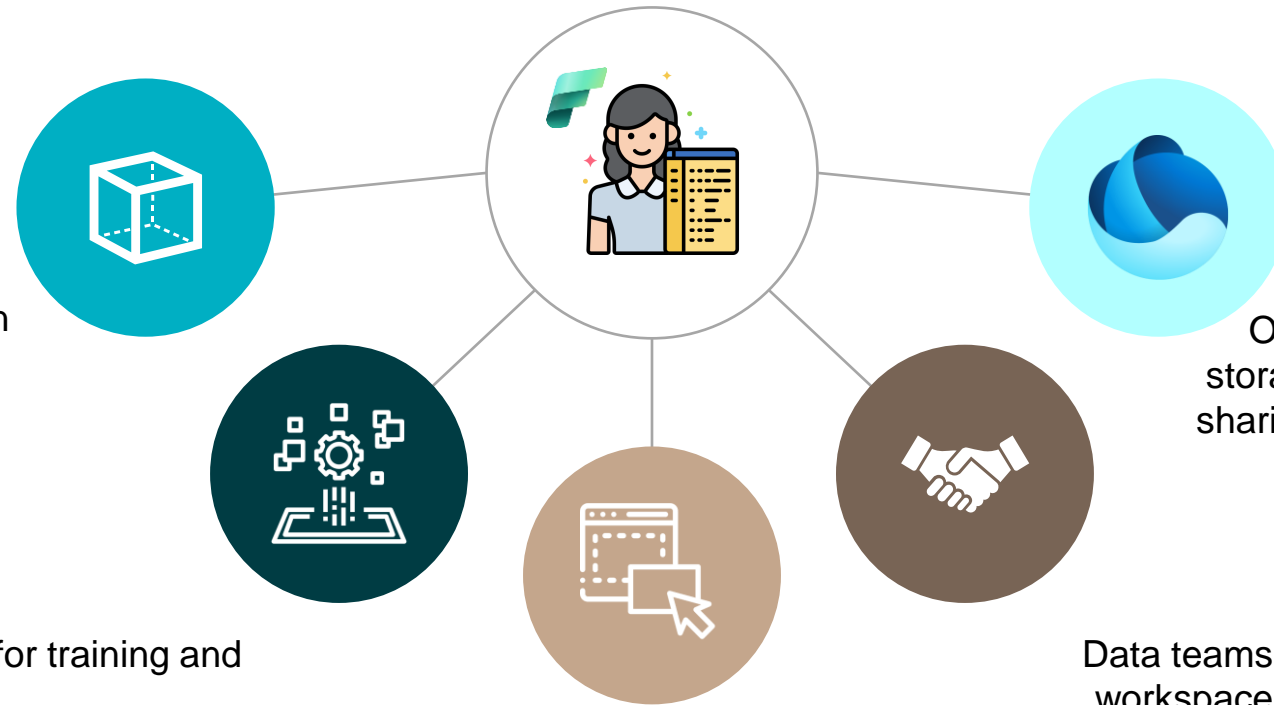
## 4. Productionalizing models

Make the model work on new data.



# Generate value from ML models

*Microsoft Fabric benefits for the Data Scientist*



## **Data Wrangler**

Easy to use data transformation and cleaning.

## **End-to-end solution**

A complete ML-platform for training and productionalizing models.

## **Easy to Use SaaS solution**

User Friendly interface with possibilities for code and no/low-code tools.

## **OneLake: OneDrive for data**

OneLake provides a single, unified storage system, where discovery and sharing of data of any format between users and applications is easy.

## **Collaboration**

Data teams can collaborate in a single workspace, on the same environment settings.