

Siemens Advanta Consulting

Data Science in Practice

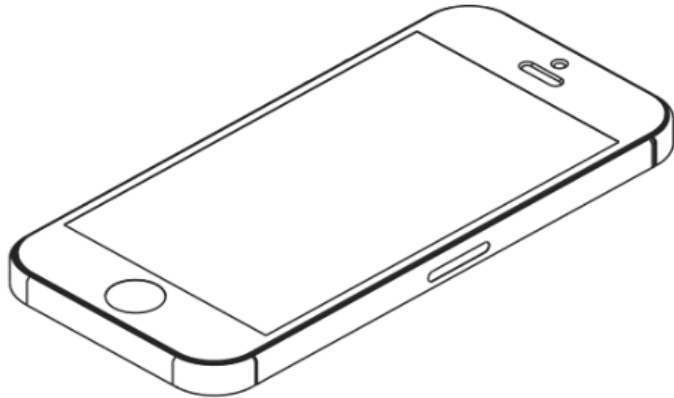
Jochen Gross | October 25th 2022

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[siemens-advanta.com](https://www.siemens-advanta.com)

Go to www.menti.com and use the code **4957 5929**

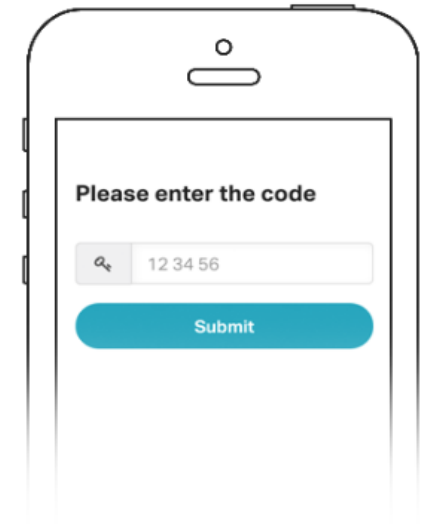
Let's get to know each other & discuss the topics you are interested in today



1

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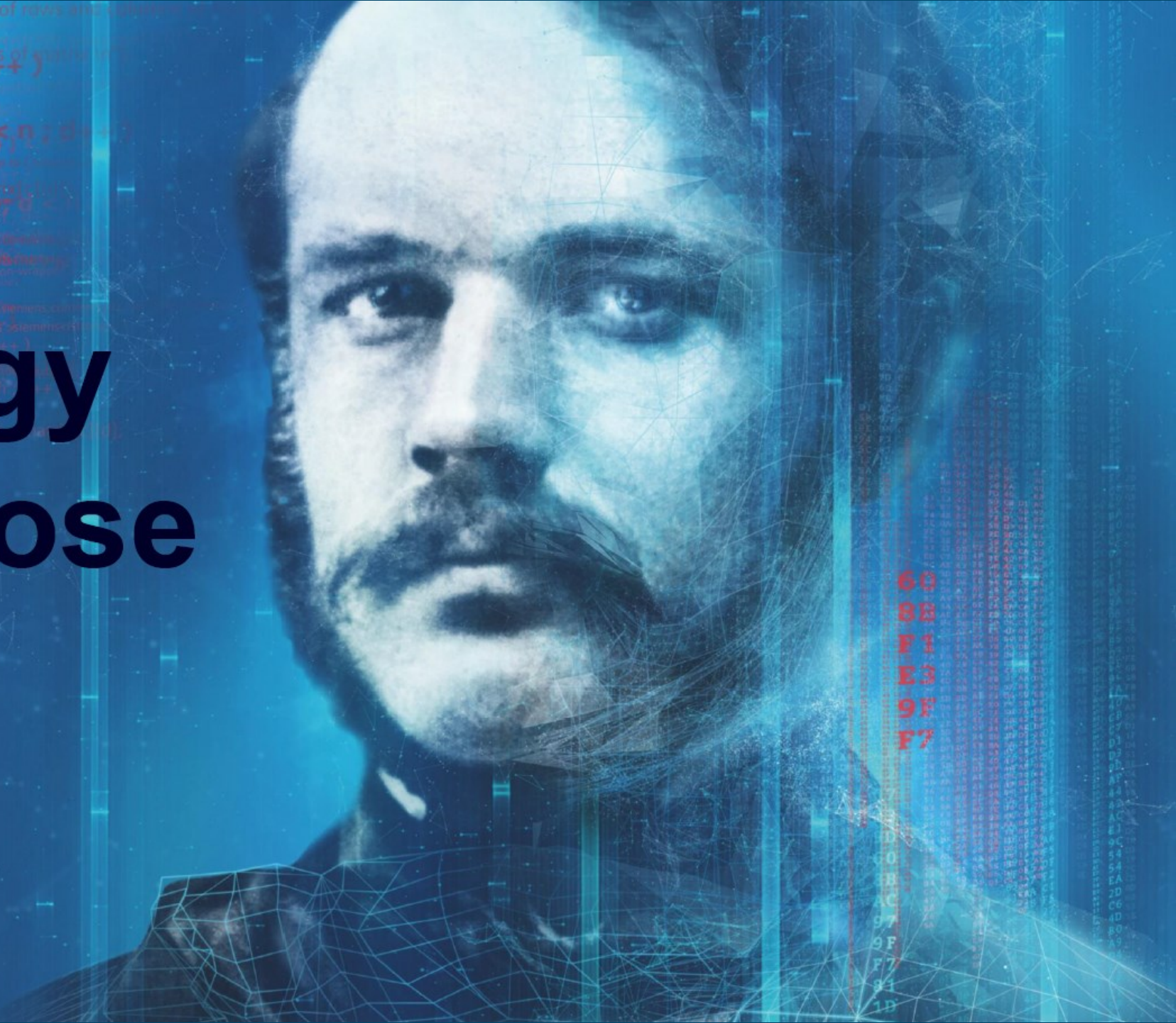
2



3

In addition, please use the chat function to raise your hand or ask questions directly during the talk!

Technology with purpose



Siemens innovations Milestones in a 173-year history

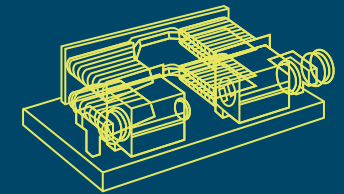
1816 – 1892

Company founder,
visionary and inventor



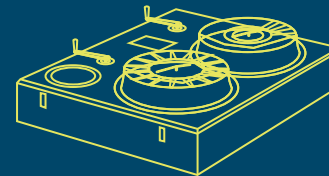
1866

The dynamo makes
electricity part of
everyday life



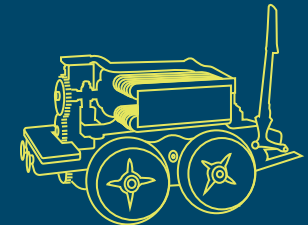
1847

Pointer telegraph lays the
foundation of Siemens as a
global company



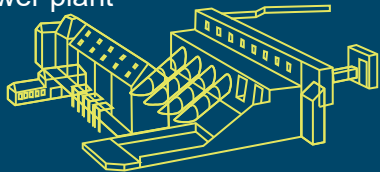
1879

World's first
electric railway



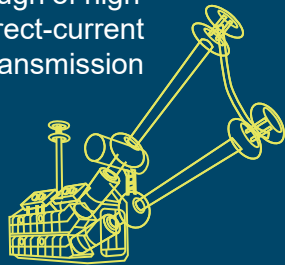
1925

Siemens electrifies the Irish Free State with a hydroelectric power plant



1975

Breakthrough of high-voltage direct-current (HVDC) transmission



2010

TIA Portal takes automation a stage further



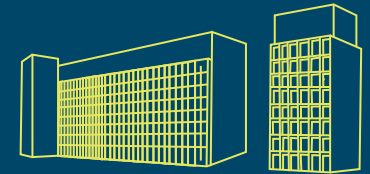
2016

MindSphere, the cloud-based IoT operating system



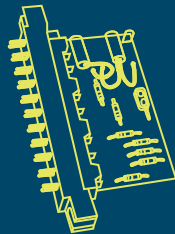
2019

Launch of first project for Siemensstadt 2.0



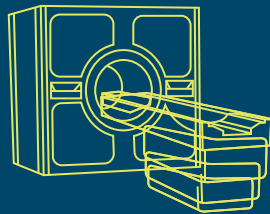
1959

SIMATIC revolutionizes automation



1983

First magnetic resonance imaging scanner



2012

Test operation of the world's largest rotor for offshore wind turbines



2018

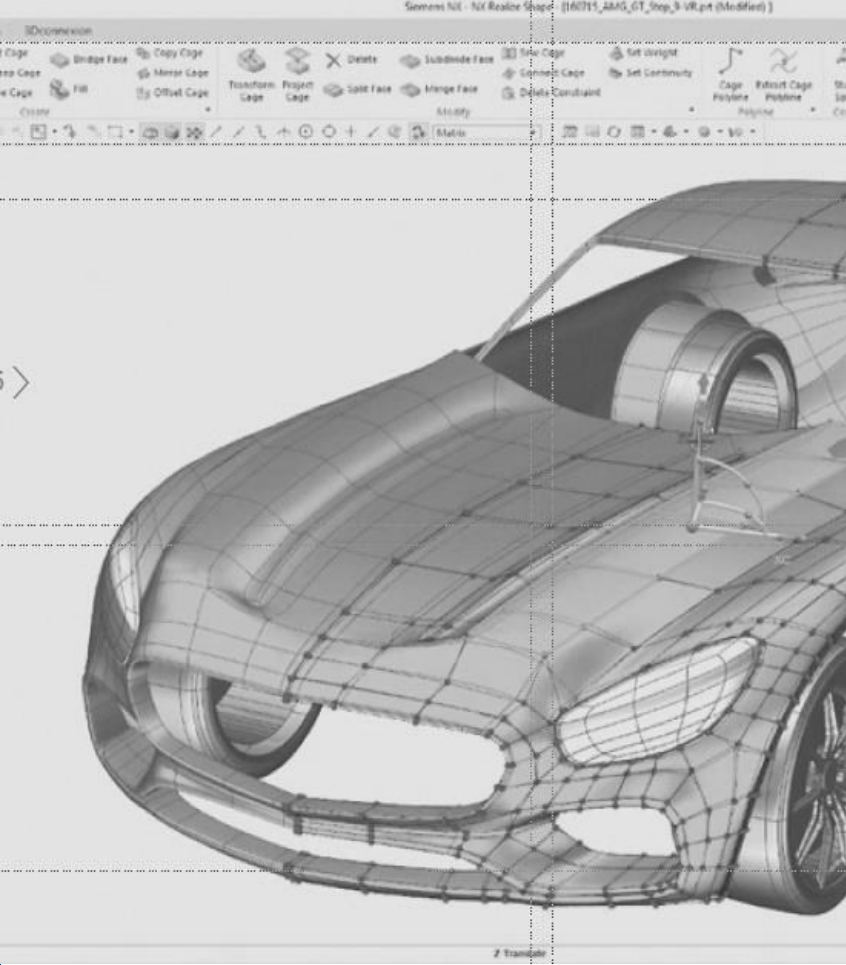
Charter of Trust: a joint initiative for a secure digital world



2020

Comfy workplace app makes it safe to return to the office during the coronavirus pandemic





We commute in cars,
designed with **Siemens
Software** ...

... built in **factories
automated by Siemens**
...

... and charged with
renewable and
decentral **Siemens
Smart Grid.**

Setup of Siemens AG

Businesses

Digital
Industries



Smart
Infrastructure



Mobility



Siemens
Advanta



Portfolio
Companies



Siemens
Healthineers¹



Countries

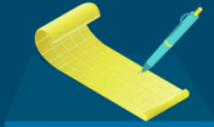
Service & Governance

¹ Publicly listed subsidiary of Siemens; Siemens' share in Siemens Healthineers: 75%

Siemens Advanta is a global consulting and professional service group with a strong focus on digitalization



Consulting

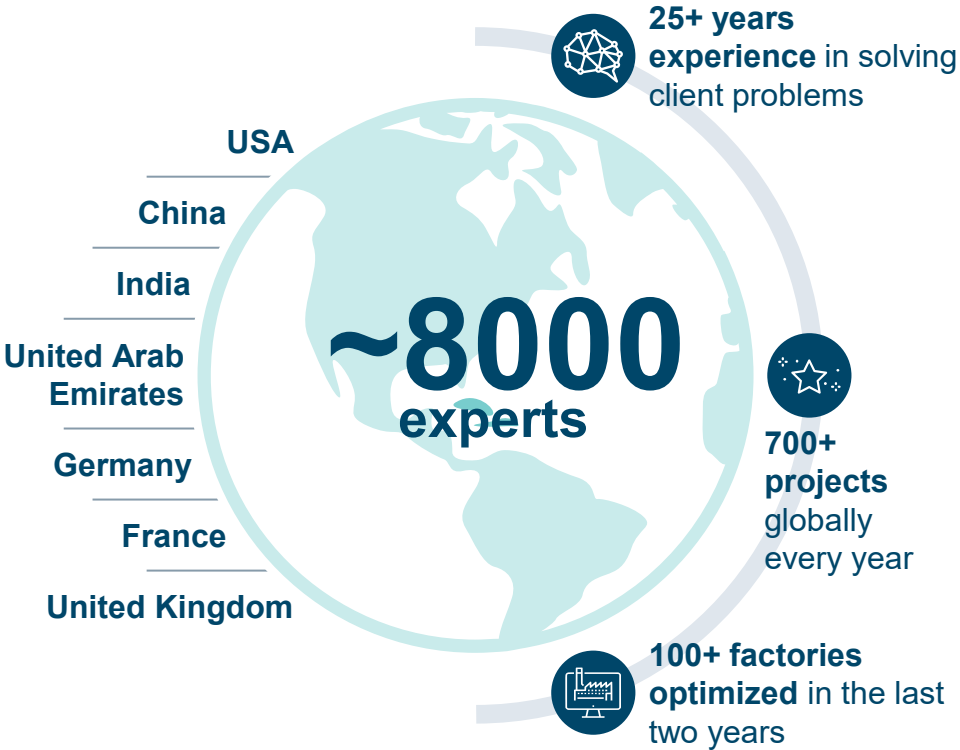


Solution Design & Prototyping



Solution Implementation

Interdisciplinary, agile teams to deliver end-to-end consulting and solutions



Our team is complementing the ADV Consulting strategy & digital portfolio with Data Science expertise – focusing on Enterprise & Industrial AI applications

Our Team



We employ **40+** data analytics consultants in **3** countries.



Our highly diverse team comes from **11** countries and is fluent in **13** languages.



Over the past two years, we have successfully **delivered >50** data science & AI projects.



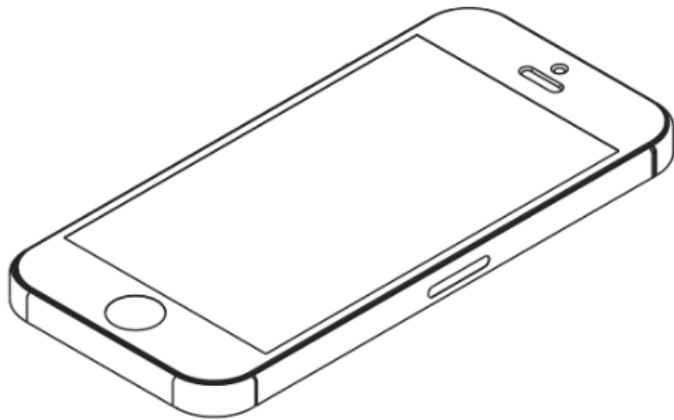
Our **interdisciplinary** team has a strong background in academia and a broad data science and data engineering toolset.



**Out of 300.000+ Jobs at
Siemens, there is not one
that is not impacted by
Data Science**

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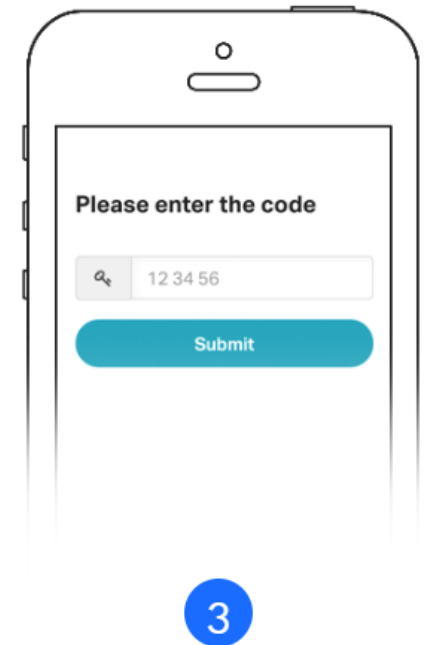
What do you estimate is the percentage of Data Science projects that never go into production?



1

www.menti.com|

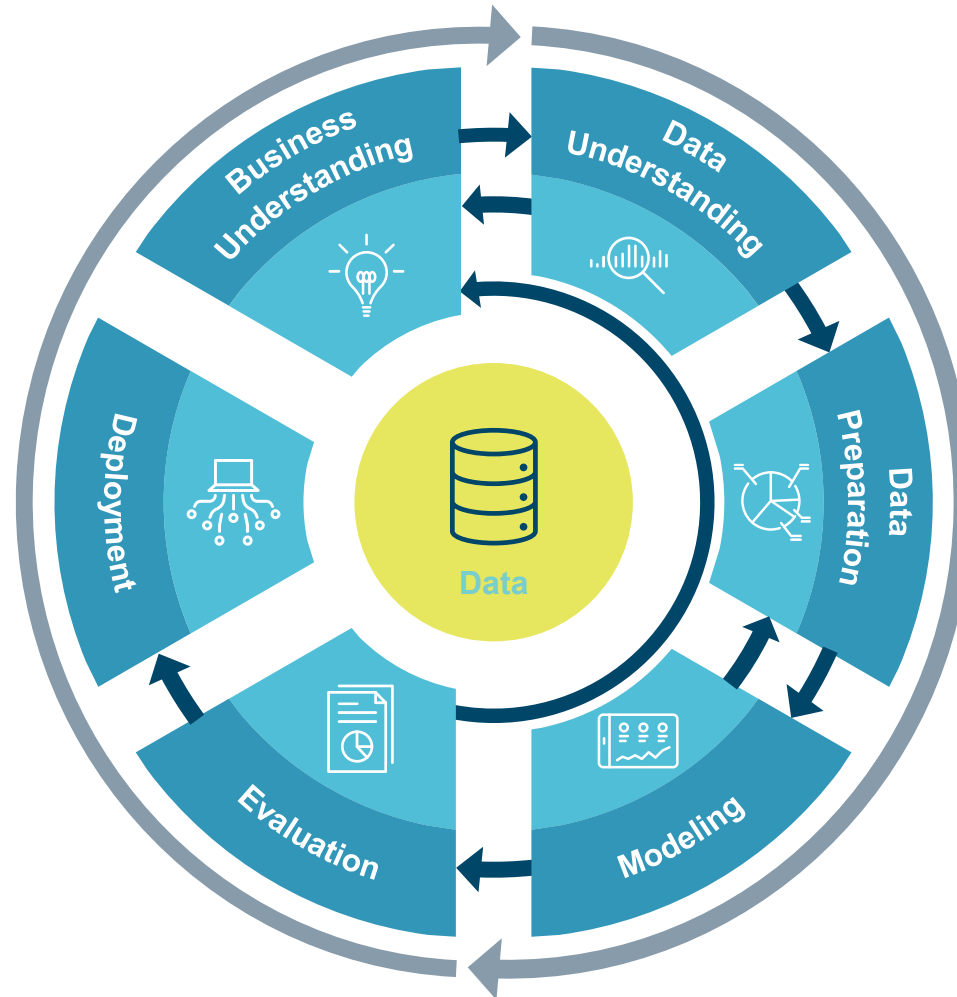
2



3



The reasons for Data Science projects to fail are manifold – the Cross Industry Standard Process for Data Mining (CRISP-DM) is one lever to mitigate the risk of failure



The framework is industry agnostic and has helped us in deploying Data Science solutions across a wide range of industries & customers

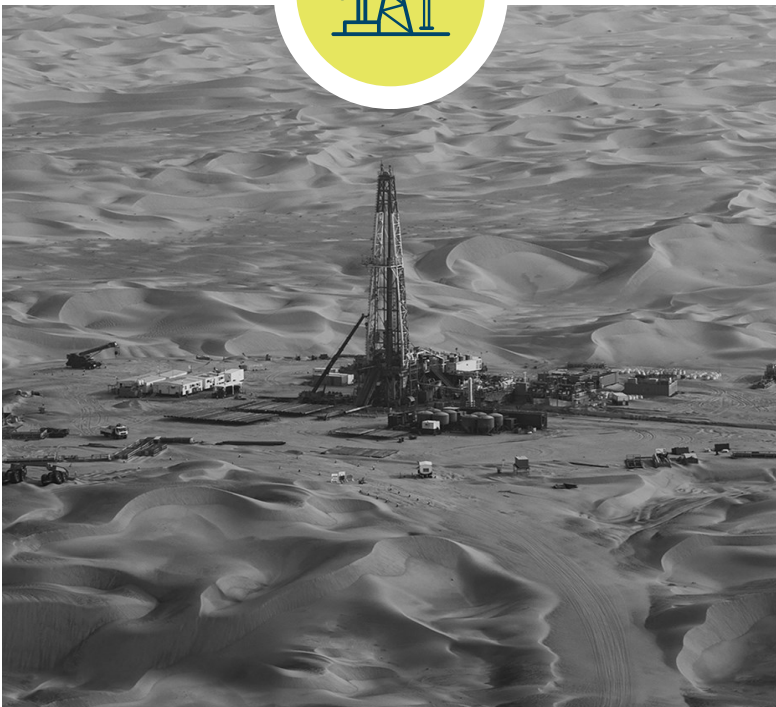
Healthcare

Remote condition monitoring of medical devices



Oil & Gas

Prediction of sand accumulation in the desert



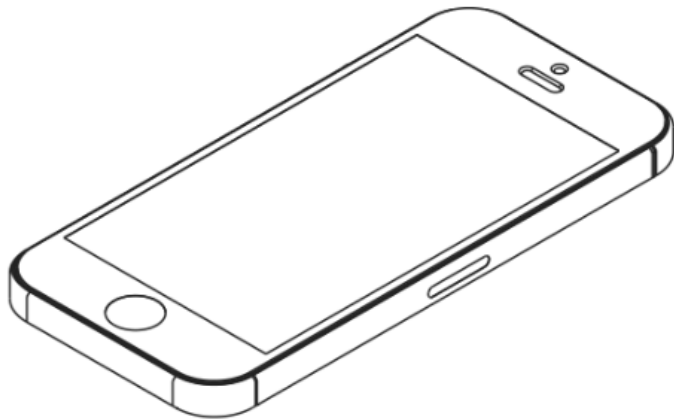
Smart Infrastructure

Decision support for sales representatives to determine the optimal sales price



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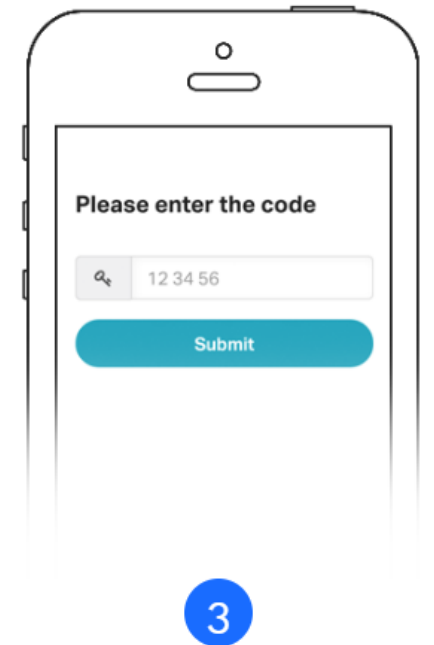
In which of the three use cases are you most interested in?



1

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2



3

Healthcare

Remote condition monitoring
of medical devices



Oil & Gas

Prediction of sand accumulation
in the desert



Smart Infrastructure

Decision support for sales representatives to
determine the optimal sales price



Data-driven solution for condition monitoring to improve patient experience



CLIENT

Multi-national corporation in MedTech



CLIENT CHALLENGES

Lack of data-based insights to improve patient experience and identify business potentials from IoT integration



OUR SOLUTION

Connecting an optimized field device set-up and advanced algorithms to derive condition monitoring solution with underlying business case

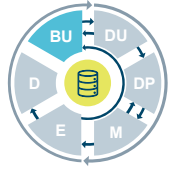


CLIENT BENEFITS

Delivery of technical proof-of concept with annual profit potential of several Mio \$

Several Mio \$
annual profit potential

Home dialysis devices have several pain points due to high returns and undetected faults



Patient

*Why is the device so noisy?
Am I safe?*

*Why do so many patients complain
about the noise? Is the device really
broken? Better return it to be safe!*



**Technical
Support**



**Failure
Analyst**

*Can't say if the device is really
broken, it seems fine.*

*Why do we have so much buffer
capacity, capital expenditures and
refurbishment costs?*



**Business
Analyst**



Approach

- Remotely detect faulty devices
- Classify which part of the device is broken
- Faster fault identification
- Improve patient retention
- Reduce refurbishment and related costs



Lessons Learned

- All stakeholders must be involved at the beginning of a project and their different challenges must be taken seriously.

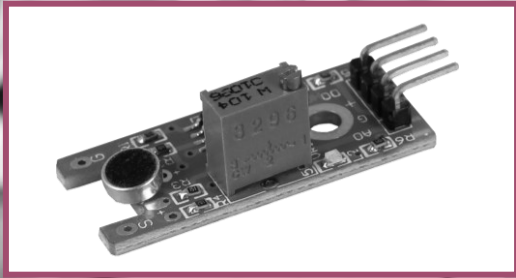
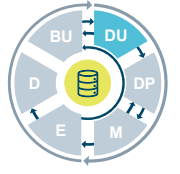


Challenges

- High return rates with undetected faults
- 15-18k return/year, but 40% of the returns with no problem found



Acoustic sensors could lead to privacy concerns – Vibration is a good alternative that can solve that problem



Approach

- Use of vibration data from sensor attached to the device
- Training data collected and labeled by technicians in laboratory setting
vibration file from broken machine?
(yes / no)



Lessons Learned

- The data quality directly depends on the selection of a suitable sensor and its setup.

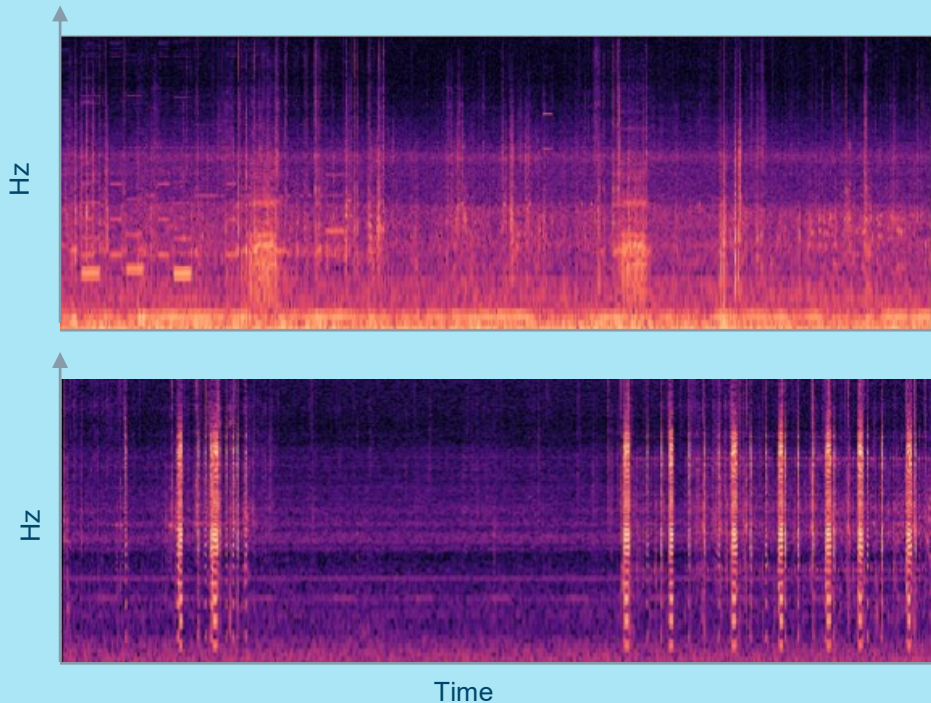
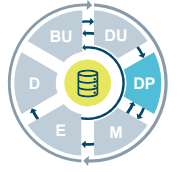


Challenges

- Collect high-quality data, while taking patients privacy into account (e.g., acoustic data would be inappropriate)



The vibration data can be used to identify different mechanical components of the machine and their faults



Spectrogram of **healthy** machine



Spectrogram of **faulty** machine

Approach

- External company takes over the data labeling
focus on individual parts of the vibration file



Lessons Learned

- Correct labeling is key and often involves people.
- Data augmentation can help with limited positive examples by creating slightly different copies of the original file.

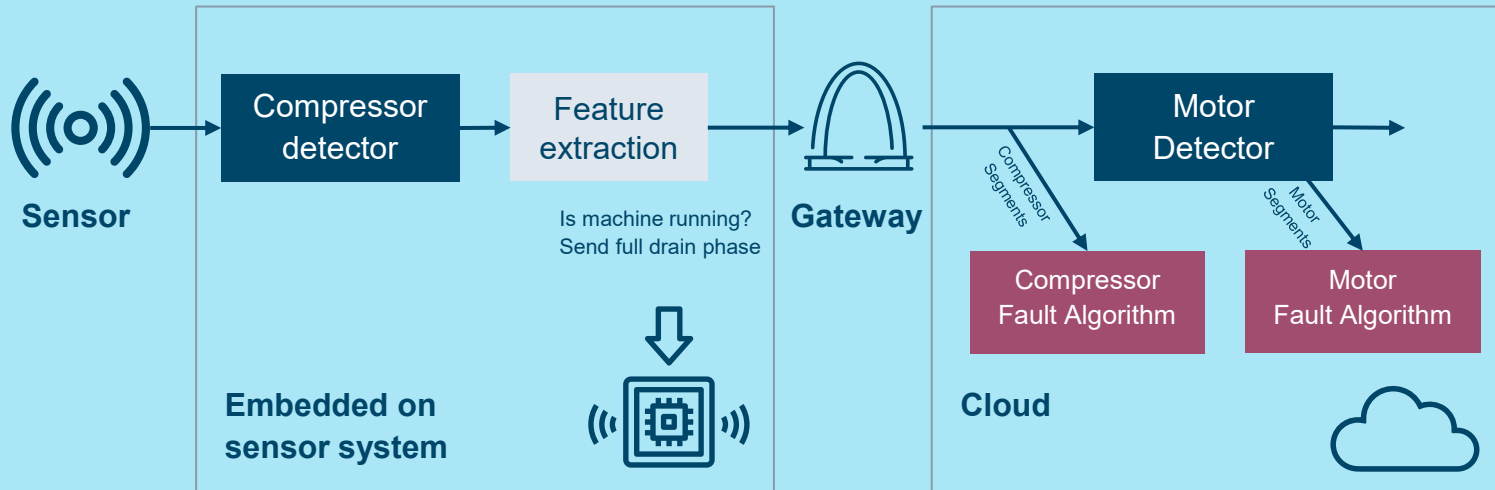
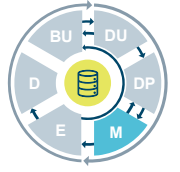


Challenges

- Identify which part of the vibration file indicates that the machine is defective
- Deal with limited positive examples



Three models are used to classify the faulty nature of different parts of the machine



Approach

- Model: Random Forest
- 3 models for different parts of the machine



Lessons Learned

- By applying the same model several times and combining their results, a detailed statement about the condition of the device can be made.

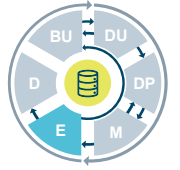


Challenges

- Classification problem
- Is machine faulty? (yes / not) > if yes: which part is broken?



Data drift in a real environment can lead to the need for adjusting the entire pipeline



Approach

- Feasibility tests
- Understand how data is collected in real-world environment and how this affects the model trained with data from a test environment
- Adjustment of the model and pipeline based on the results of the evaluation



Lessons Learned

- An unsatisfactory result does not always have to be caused by the model itself, but can also result from poor data quality or data labeling.
- Often the models need to be adjusted after testing in the real environment, and sometimes even the initial target.

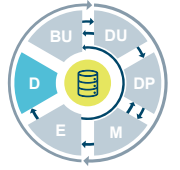


Challenges

- Models should also provide good predictions for new data



Pains and gains of all stakeholders must be considered in the rollout to create a value proposition



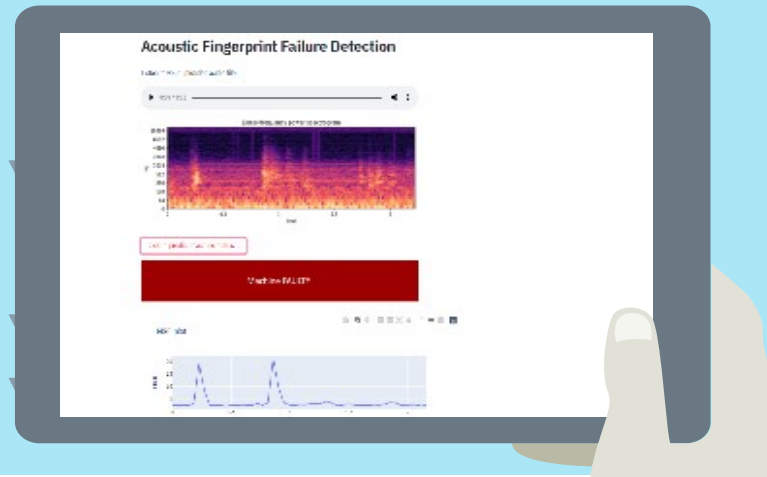
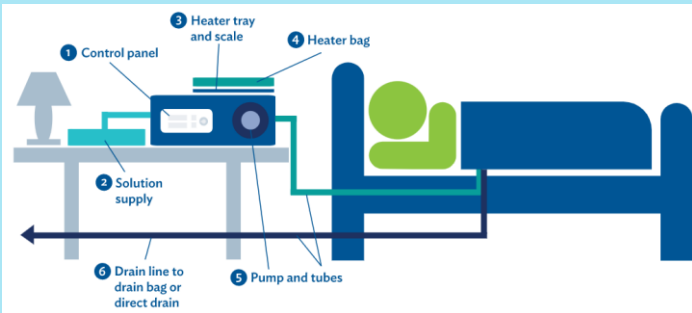
Patient



Technical Support



Failure Analyst



Approach

- Demonstrating proper installation of the device with the new sensor in patients' premises
- Condition monitoring app for technical support and failure analyst that provides a full fleet overview and pro-active alert monitoring



Lessons Learned

- The best model is worth nothing if it is not accepted and used. Usability is key!



Challenges

- Proper set up of the device at the patient's home
- Overview on the functionality of the devices for the company



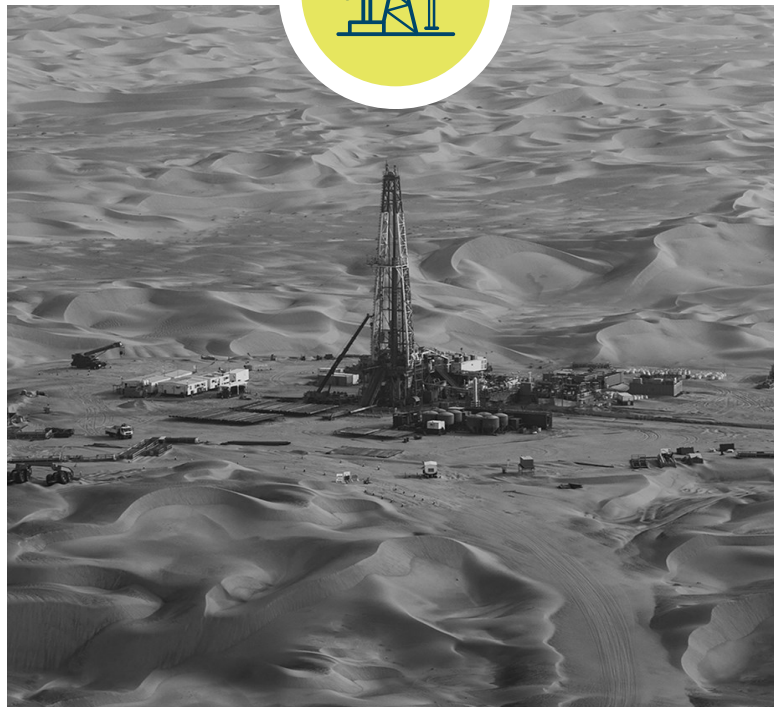
Healthcare

Remote condition monitoring
of medical devices



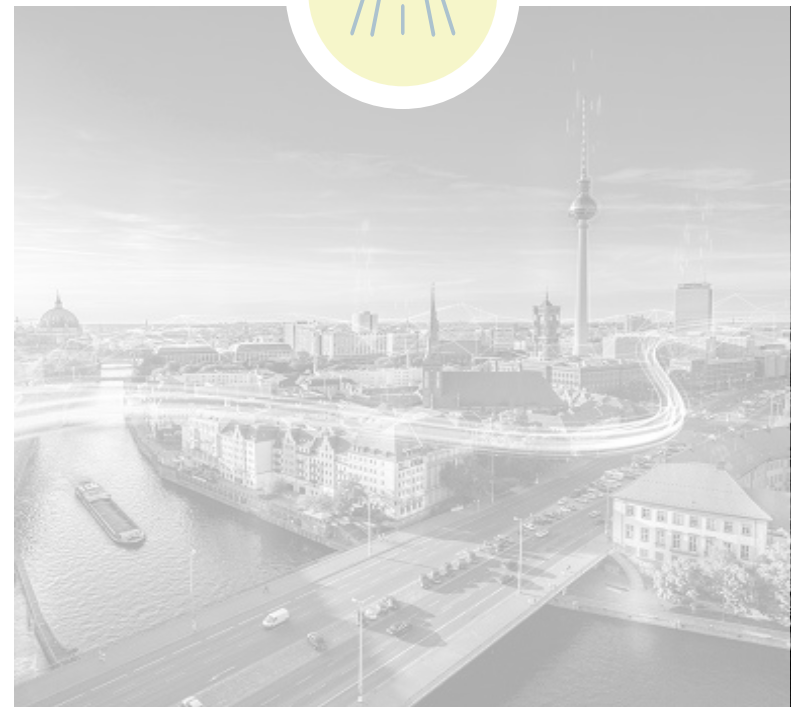
Oil & Gas

Prediction of sand accumulation
in the desert



Smart Infrastructure

Decision support for sales representatives to
determine the optimal sales price



We predicted sand accumulations to help a major oil & gas customer reduce cost of operations while improving plant availability



CLIENT

Leading O&G company in the Middle East



CLIENT CHALLENGES

Disruption of operations due to accumulation of sand and dust on critical infrastructures.



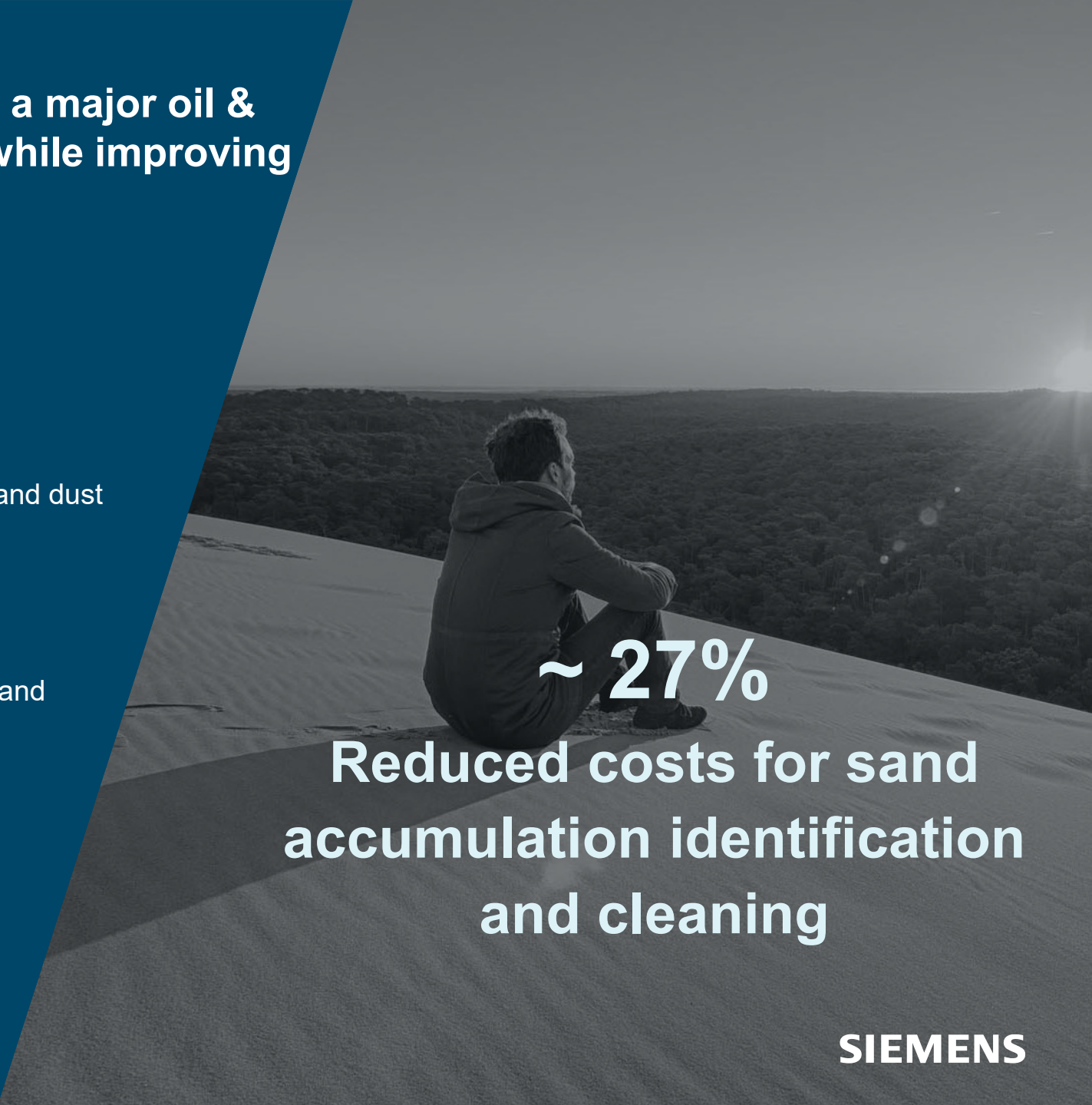
OUR SOLUTION

Leveraging existing data and its potentials to predict sand accumulates based on a machine-learning algorithm.



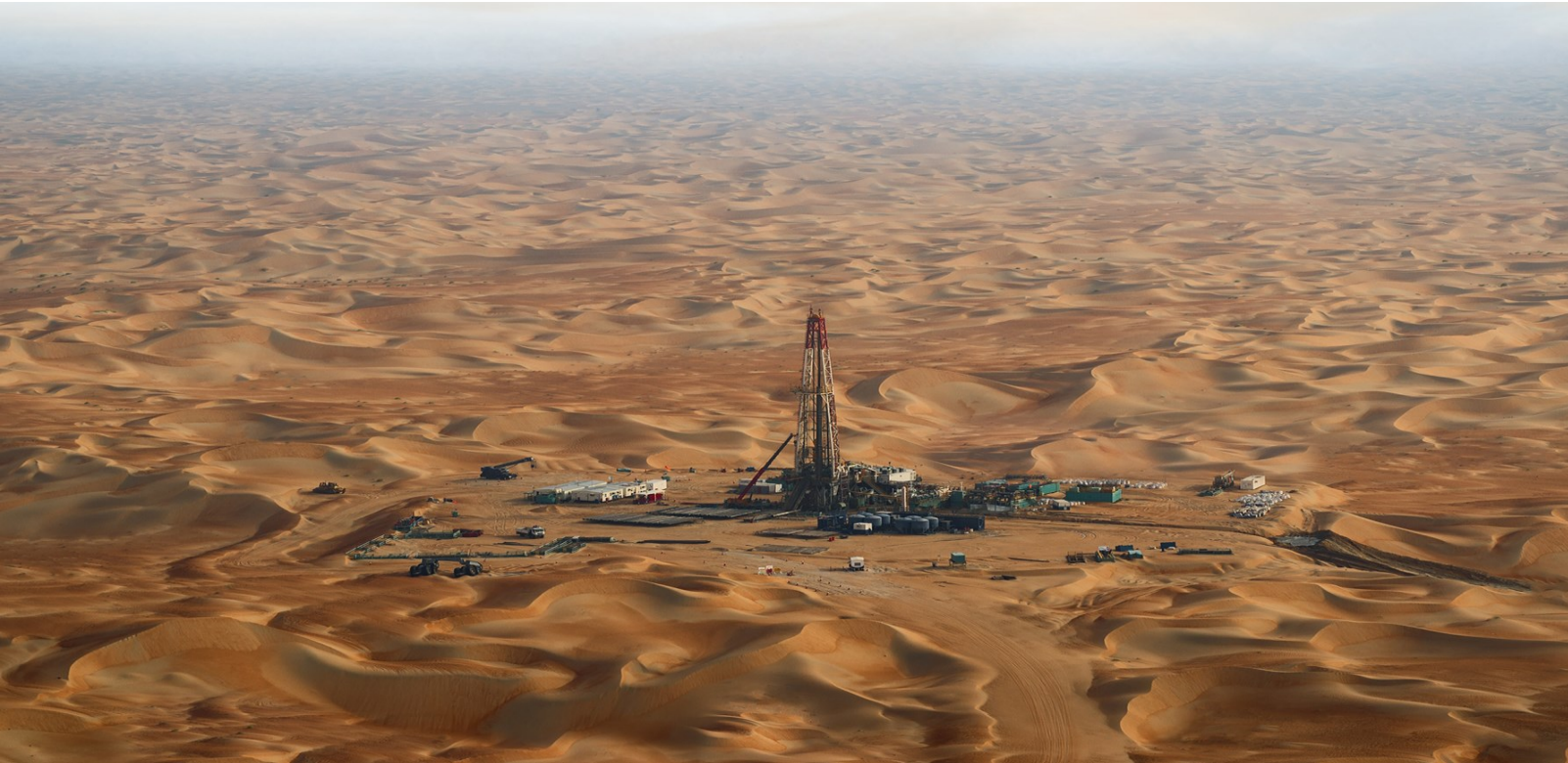
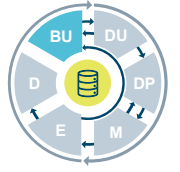
CLIENT BENEFITS

Reduced risk of activity disruptions and cost of sand cleaning, based on prediction of sand accumulation in industrial areas.



~ 27%
Reduced costs for sand
accumulation identification
and cleaning

Sand accumulations and dust block and cover access to roads and disrupt operations



Challenges

- Disruption of operations due to accumulation of sand on critical infrastructure
- Manual checks through patrols required



Approach

- Leveraging existing data and its potentials to predict sand accumulates based on a machine-learning algorithm

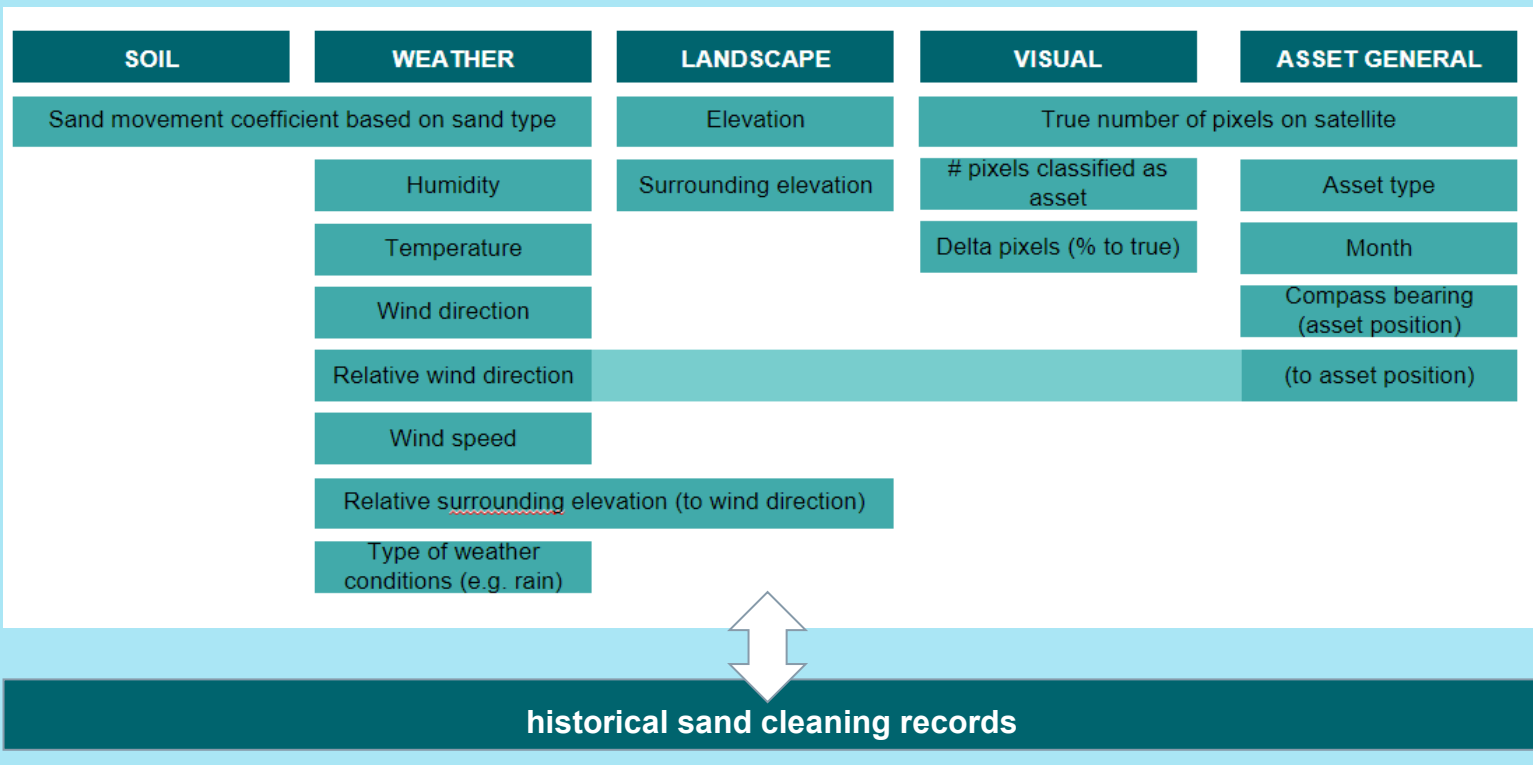
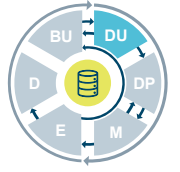


Lessons Learned

- The development of an efficient solution starts from the deep understanding of customer's pain points. First, the problem must be formulated from a business perspective, then it can be determined what data is needed to answer the question.




We take five different groups of factors influencing sand movement and accumulation




Approach

- We combined available data on clearing activities for every geo-point and trained an algorithm to predict accumulation
- We identified five high influencing factor groups



Lessons Learned

- Various internal and external data is often needed to explain a situation.
- If the data sources are not well selected or some are missing, the model will not produce the intended results.

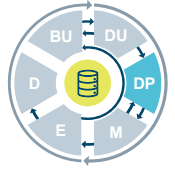


Challenges

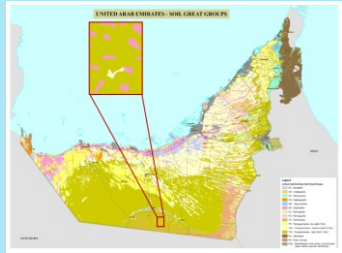
- Identify factors that have high influence on sand movement



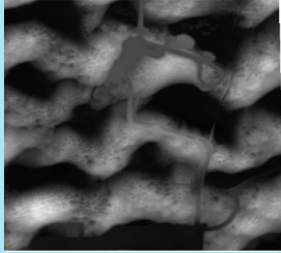
Pre-processing procedures highly depend on the input data and are key for the final predictive model



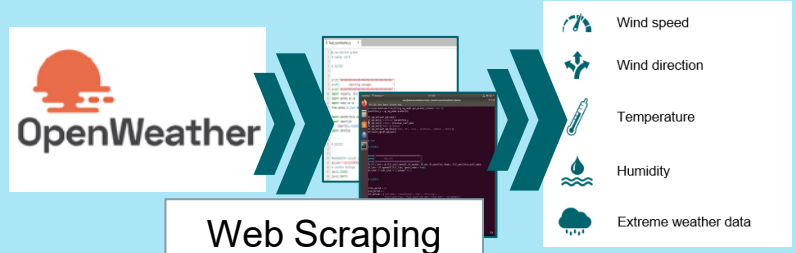
SOIL



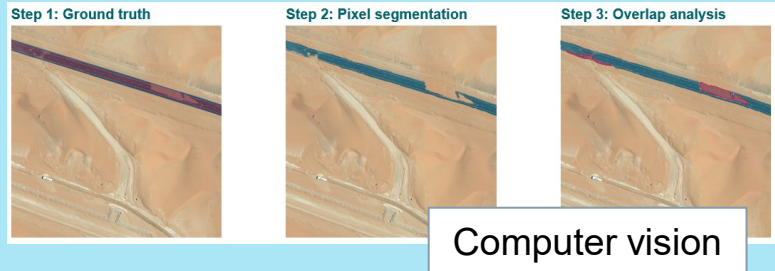
LANDSCAPE



WEATHER



VISUAL



HISTORICAL SAND CLEANING RECORDS



Approach

- Different input data requires different handling to be used in the final prediction model.
- Text mining, computer vision, web scraping were used for the pre-processing steps.



Lessons Learned

- Often data cannot be used directly in the final prediction model.
- Different data types require different preprocessing, for this, ML algorithms are applied, for instance.

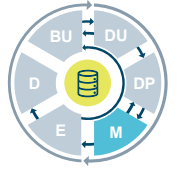


Challenges

- Different data sources and pre-processing procedures for the input factors are required

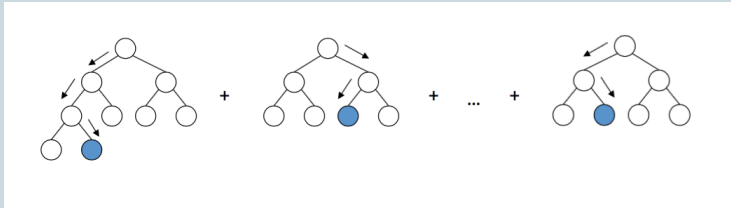


We used two conceptually different modelling architectures to capture specifics of the data



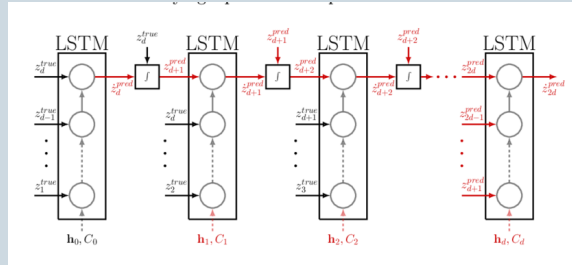
Gradient Boosting

- Partitioning non-linear model
- Explainable on global and local levels
- Tuning with Bayesian grid search
- Consecutive predictions of daily values



LSTM

- Multiplicative non-linear model
- Low-to-no explainability of results
- Random grid search tuning
- Simultaneous prediction for the complete horizon



Approach

- Predict sand movement with two conceptually different modelling architectures: LSTM and XGBoost



Lessons Learned

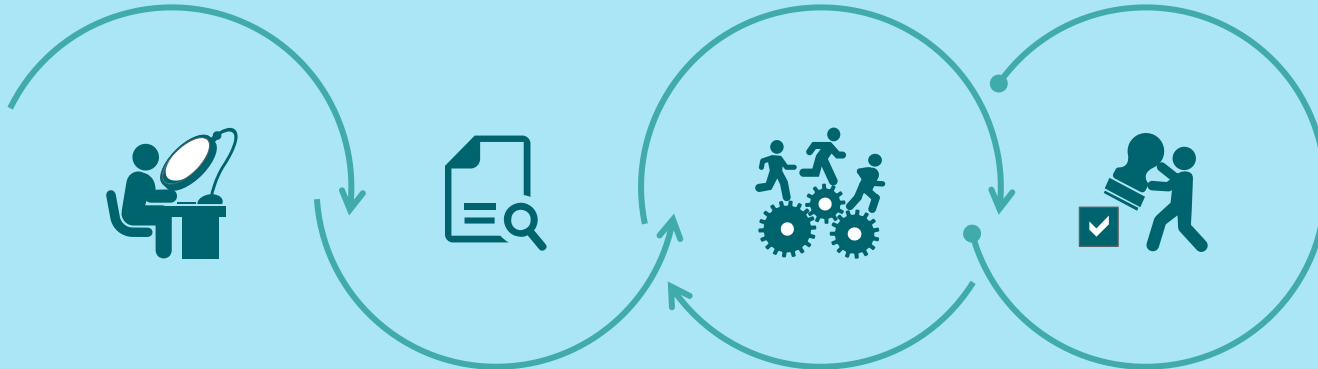
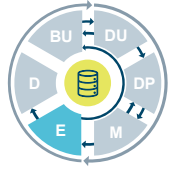
- A single model is often not sufficient to address the different types of data and derive knowledge from them.

Challenges

- Development of a suitable model that allows good conclusions to be drawn from the various input data



User Acceptance Test will be carried out to test scenarios that emulate the user's journey



Identify test scenarios

Identify and prepare test scenarios based on the high-level business process and create test cases with clear steps

Prepare test data

Define and prepare the data required for testing

Execute testing

Run the test cases, identifying bugs (if any) and retesting the bugs once solved

Sign off UAT

Conclude and sign off UAT, ensuring that no critical defects are open and that business objectives are met

Challenges

- Verifying whether the developed solution meets the needs of the end user or whether there is room for improvement



Approach

- Test the model's ability to identify sand accumulation areas and volume
- Test the ability of users to identify the location and amount of accumulated sand

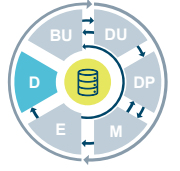


Lessons Learned

- It is not only important for the model to perform well, but also to get the users on board early.
- User acceptance tests are key!



The results of data analysis and predictions are delivered to the end users through a tailor-made mobile app



Approach

- Develop different interfaces for different end users
- Reduce risk of activity disruptions through early detection of sand blocking
- Optimized tracking and scheduling process of sand removal activities



Lessons Learned

- Different interfaces of the final solution help to meet the needs of different end users.



Challenges

- Address the needs of different end users



Healthcare

Remote condition monitoring
of medical devices



Oil & Gas

Prediction of sand accumulation
in the desert



Smart Infrastructure

Decision support for sales representatives to
determine the optimal sales price



Bring the experience of the organization to the fingertips of the sales representatives



CLIENT
Smart Infrastructure



CLIENT CHALLENGES
Lack of consistency and transparency about quotes to customers



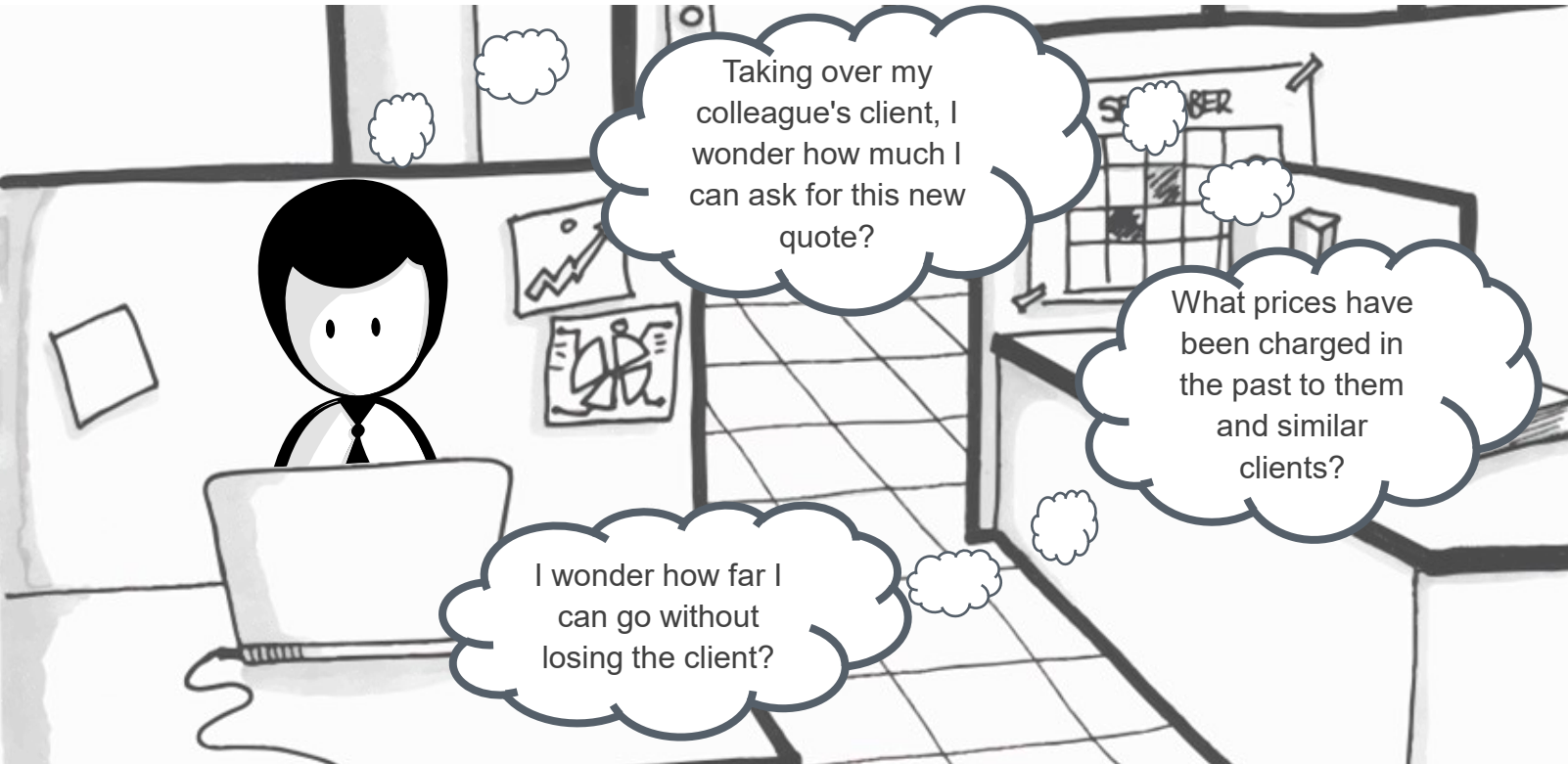
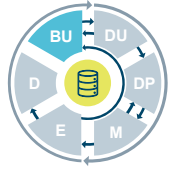
OUR SOLUTION
Automated price recommendation based on the history of sales offers via machine learning



CLIENT BENEFITS
Transparency enhancement and improvement of sales margin



Need for decision support in offering prices to new as well as to existing customers based on historic data



Approach

- Get commitment by top management and identify key contacts
- Conduct workshop(s) with business process and governance experts to understand the business process and its deviations

Lessons Learned

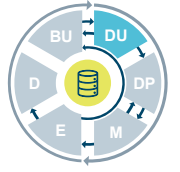
- Business processes can vary across country even within the same organization. Pay attention to details and include the major stakeholders.

Challenges

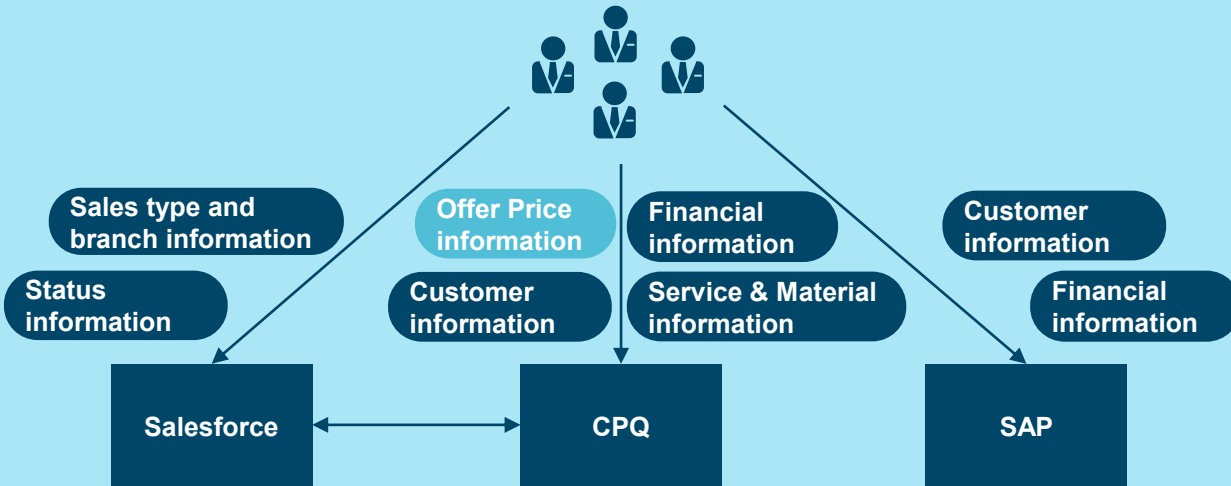
- Understand the business process and its deviations (per country & business type)
- Identify key contacts to learn about these processes and drive them



Identifying and connecting the right data sources is one thing, assessing and ensuring high data quality another



Sales experts' usability of systems varies by country & branch



Approach

- Conduct workshop(s) with business / data experts to have a holistic view and identify influencing factors on the price / to data sources
- Compare expert insights with own data analysis

Lessons Learned

- A good business understanding helps significantly to make sense of data. Clarify questions on data early with business to make sure you do have the right understanding.
- Well defined processes are key for consistent data quality.

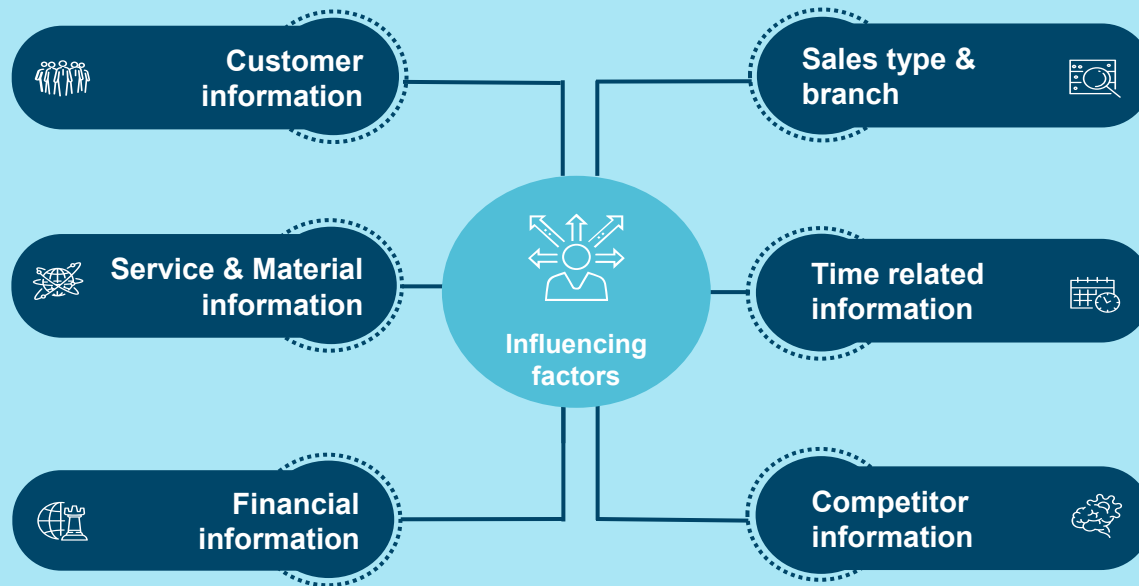
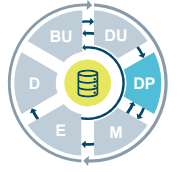


Challenges

- Process differences result in data differences
- Identify data experts and get access to data sources



Conduct workshops with business and data experts to identify and map influencing factors



Approach

- Ensure data quality
- Ensure explainability of features



Lessons Learned

- Data preparation has to be done in co-creation with the business and data experts of the customer. Due to data availability & quality, many potential features have to be filtered out.

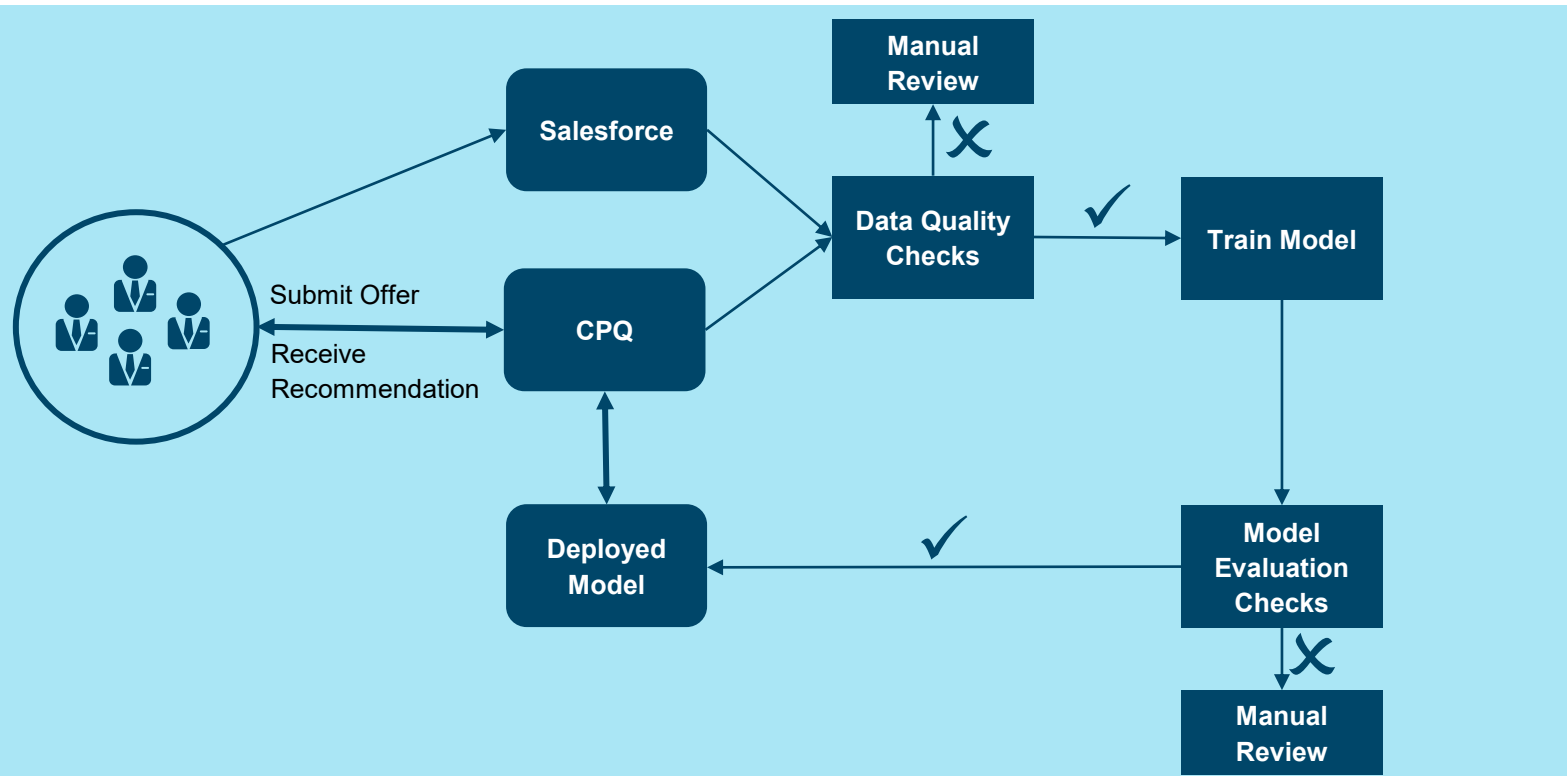
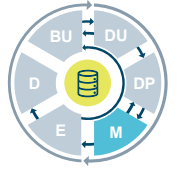


Challenges

- Map identified features to data sources and ensure data quality
- Explainability of features



The trained model must be thoroughly evaluated from a technical and business perspective



Approach

- Train model
- Conduct technical evaluation (analyze for systematic model errors)
- Conduct business evaluation (feature importance vs. business understanding)



Lessons Learned

- Create models as individualized as necessary, but as standardized as possible! Adjustments for individual countries only if it results in high impact.

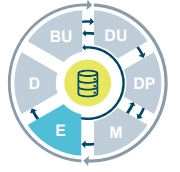


Challenges

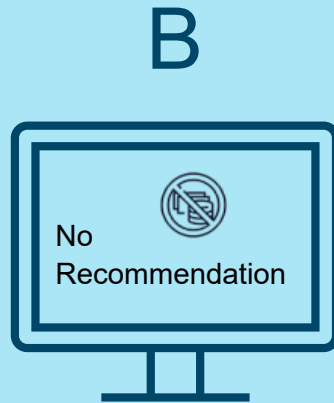
- Standardization vs. individualization
- Consideration of concept drifts (e.g. inflation and Covid)



The calculated impact expectations have been validated via an AB Testing based on real business transactions



Test group
Price Recommender



Control group
No Price Recommender

Approach

- Conduct AB Testing together with selected countries to prove business impact of AI solution



Lessons Learned

- Before the final deployment of a new solution, the expected benefits for the end users must be tested, validated and proven.

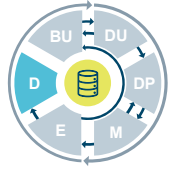


Challenges

- Model evaluation
- Validation of business impact

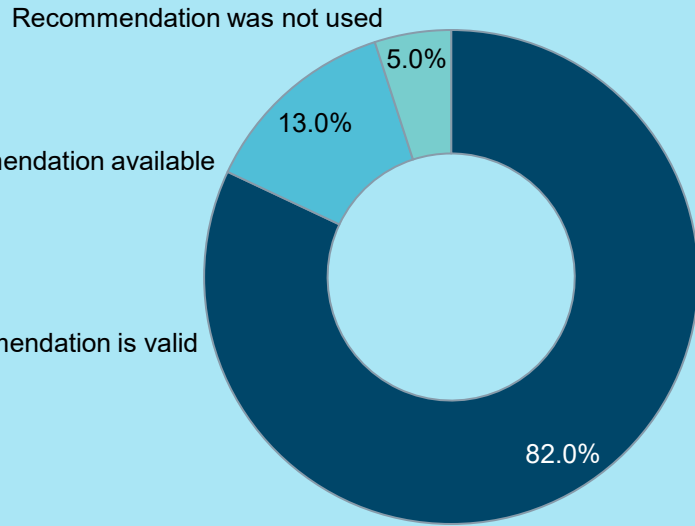


Development of a central reporting cockpit to ensure user acceptance and long-term business impact

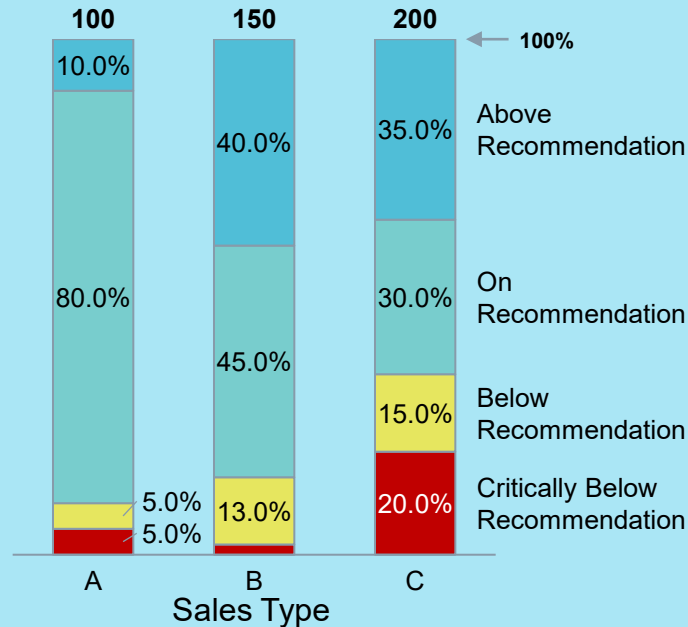


Price Recommender Dashboard*

Submitted Offers



Submitted Offers



* Simplified extracts of parts of the dashboard using dummy data

Challenges

- Integration into existing system
- User centric reporting cockpits



Approach

- Customer co-creation via user stories
- Definition with meaningful KPIs that can track user acceptance and long-term business impact

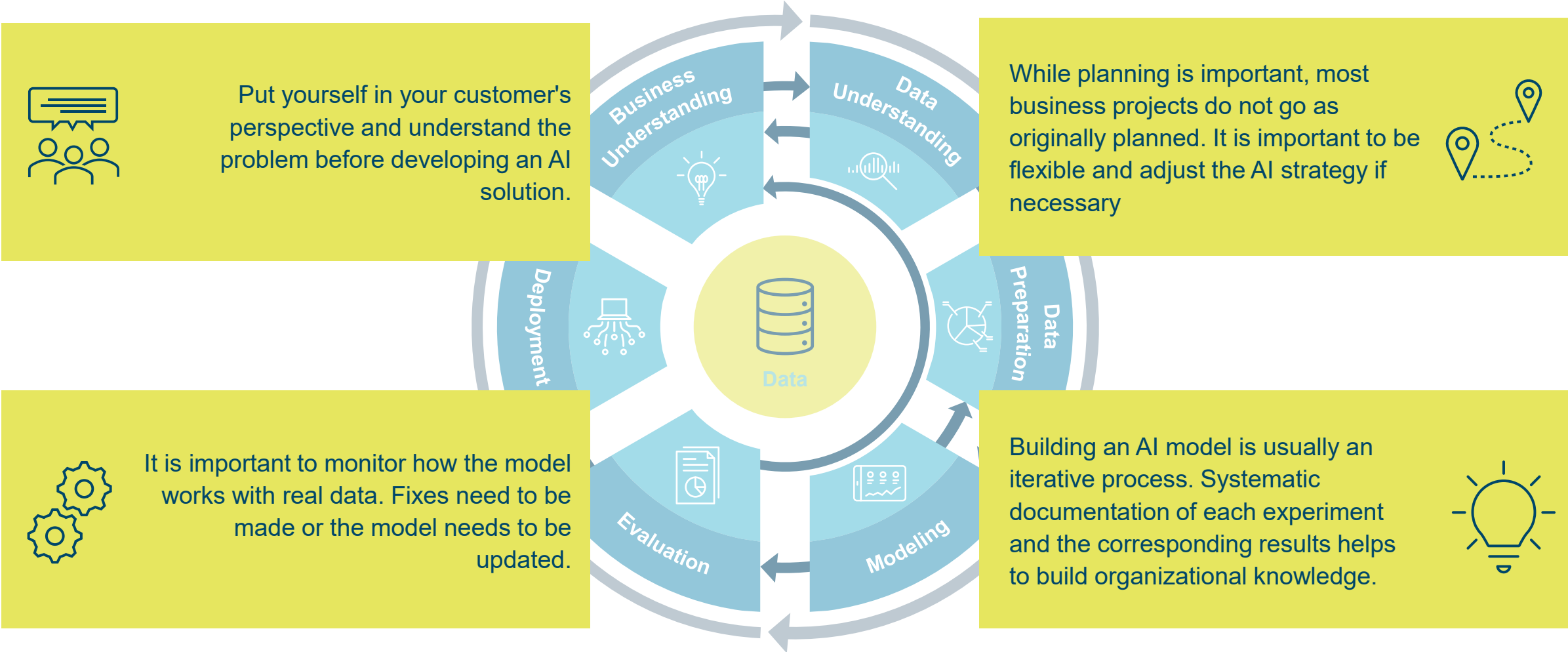


Lessons Learned

- Only if you can ensure user acceptance and business impact, you create value with the AI solution.



The reasons for Data Science projects to fail are manifold – the CRISP-DM framework is one lever to mitigate the risk of failure



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Thanks for listening

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