What Is Next?



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Overview



Handling features

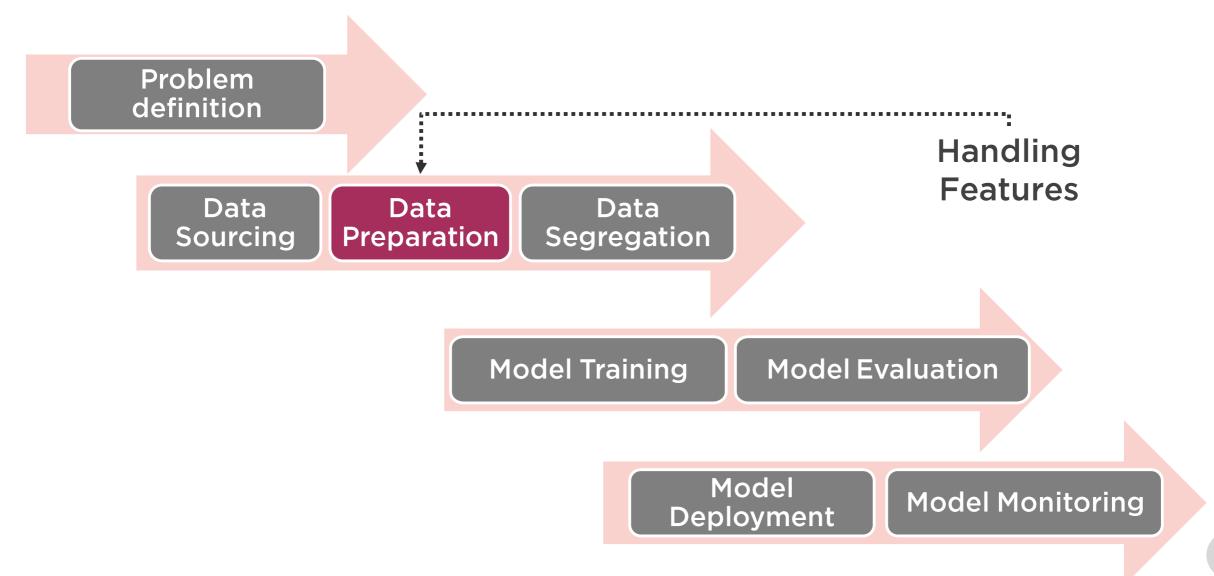
Model improvement

Al as a Service

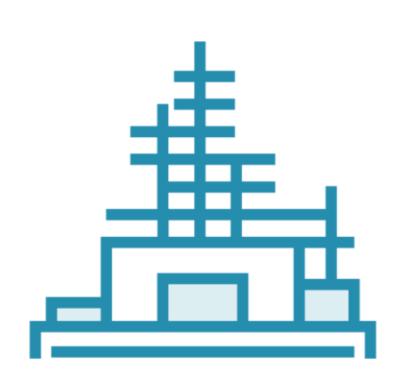
Machine Learning DevOps



More on ML: Handling Features



Handling Features

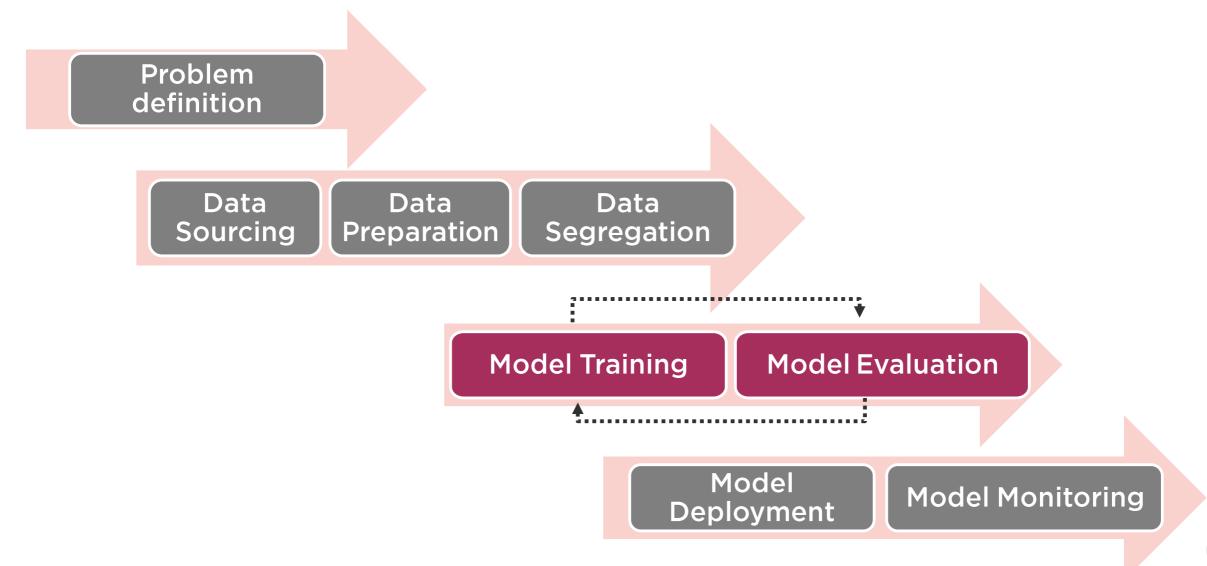


Why we need to handle features?

Steps

- Feature engineering
 - Converting birth date to age
 - Converting weekdays to numbers
- Feature scaling
- Feature selection

More on ML: Model Improvement



Ensemble Methods

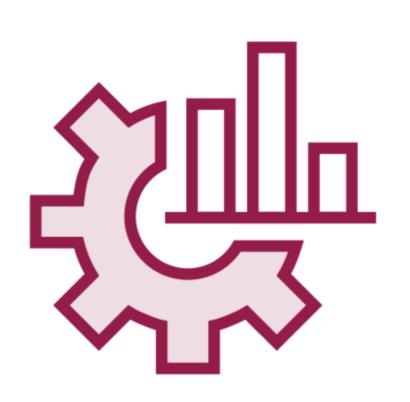


Combines different ML techniques

Common categories

- Bagging
- Boosting
- Voting

Hyperparameters Optimization

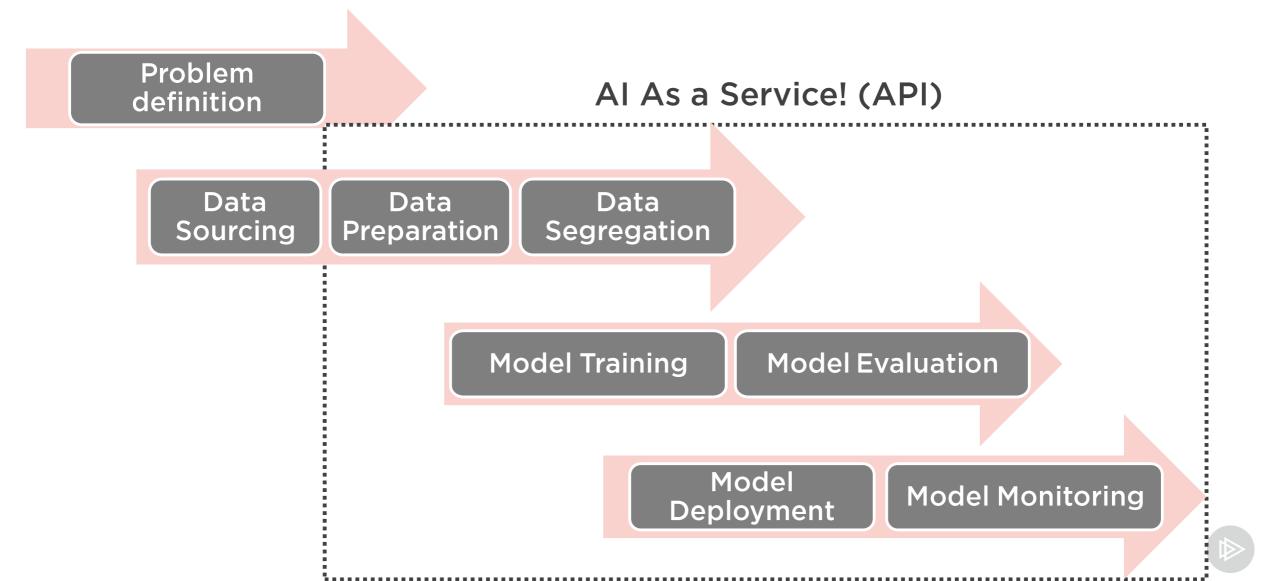


Blackbox "terminals"

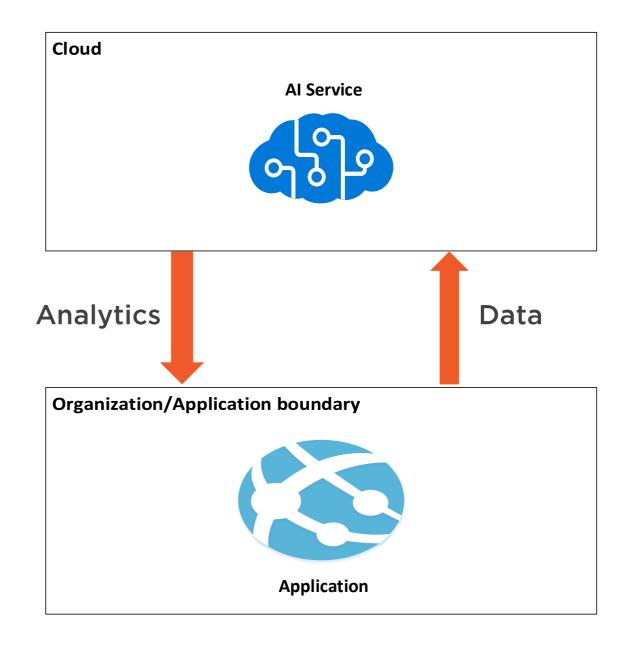
Types

- Grid Search
- Random Search

More on ML: Automated ML

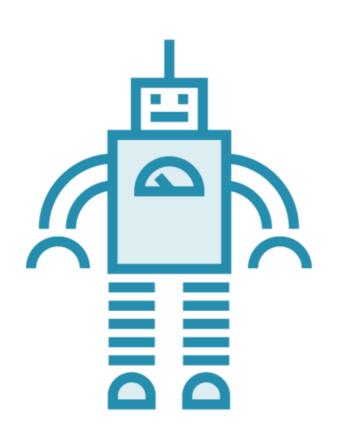


What is that black boxed API?





Al as a Service (AlaaS)



Abstracted as APIs

Works well with standardized problems

- Vision
- Text to Speech
- Sentiment Analysis

Many players in the market

- Microsoft/Amazon/Google

Challenges

- Performance
- Compliance
- Strict/Lock-In



Democratized AI – AI as a Service for all!

January 6, 2019 | 8 Comments



Background ...

Few of years ago, writing a software application that performs any sort of "intelligence" was not the easiest task to do. Considerable know-how around certain frameworks and libraries need to be established, which means learning curve, dollars money, and a risk that many were unwilling to take. I remember 8 years ago, when I was planning to implement some object tracking algorithm for a university project. I had to read around optical flows. Lucas-Kanade (LK) algorithm

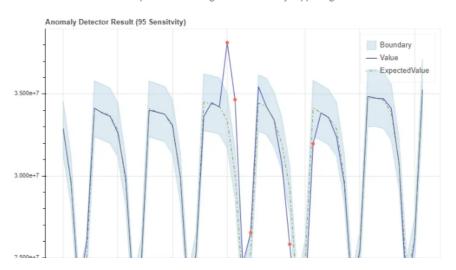


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    false
    false
],
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    false,
    false
],
    "isNegativeAnomaly":[
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],
    "isNegativeAnomaly":[
    false,
    false
],
    "period":12
}
```

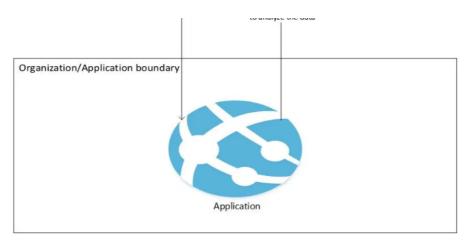
Table 3 Batch point detection API response sample - Note that suggested window does not exist

Explanation through Visualization

Since A picture worth a thousand words, I believe it would be simpler to have an image that explains how anomaly detection works so that we have a deeper understanding of what is actually happening.

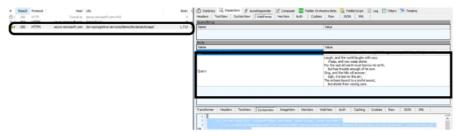






Fiddler:

If we intercept the request using fiddler, we can see that the poetry text is transmitted to azure.microsoft.com domain. The elapsed request time was 2.454 seconds.



Case 2: Al service on docker image

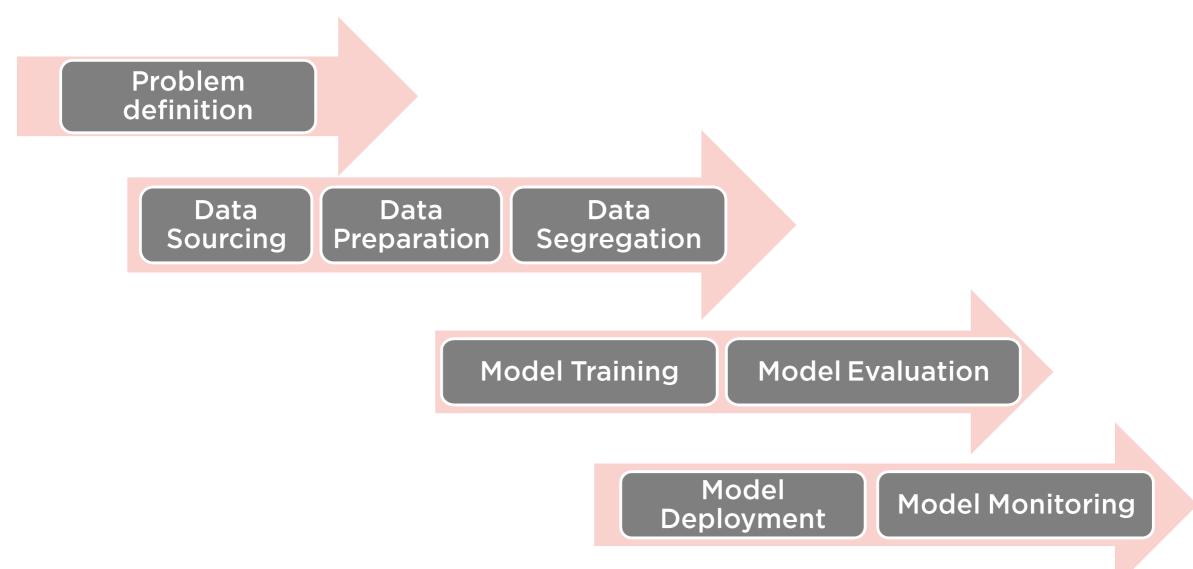
The following picture gives a high-level description of how the AI as a Service operation occurs when deployed as Azure cognitive service container:

- The application sends data to Azure cognitive service container on- prem (confidential data does not leave the organization boundary).
- 2. Azure cognitive service container responds with analytics results to the application.
- 3. Azure cognitive service reports back billing information to Azure to debit the subscription (Periodic billing).





More on ML: Operationalization



Operationalization



Deploying our model

- Serialization and Deserialization: Python (Pickle, Joblib)
- Web services (REST)

Monitoring our model

- Logging
- Auto-healing



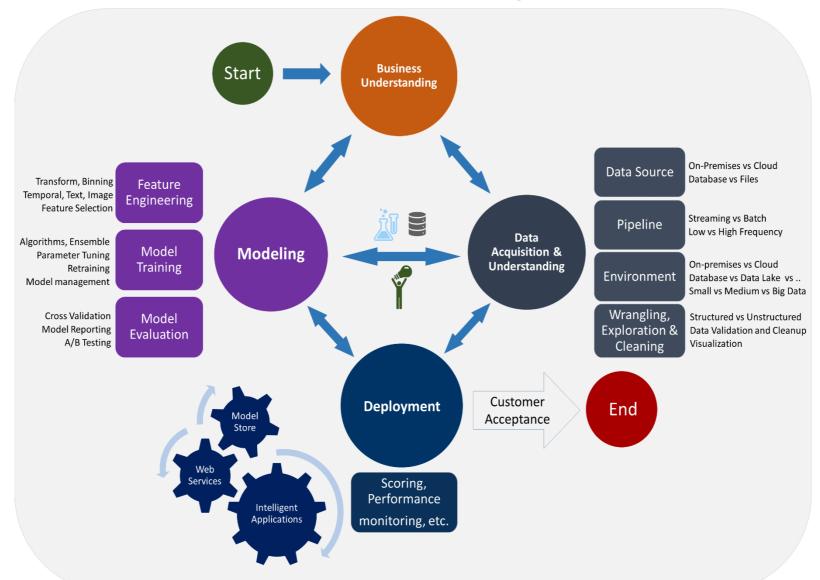
More on ML: Data Science Process

Team Data Science Problem definition **Process** Data Data Data Sourcing Preparation Segregation **Model Evaluation Model Training** Model **Model Monitoring** Deployment

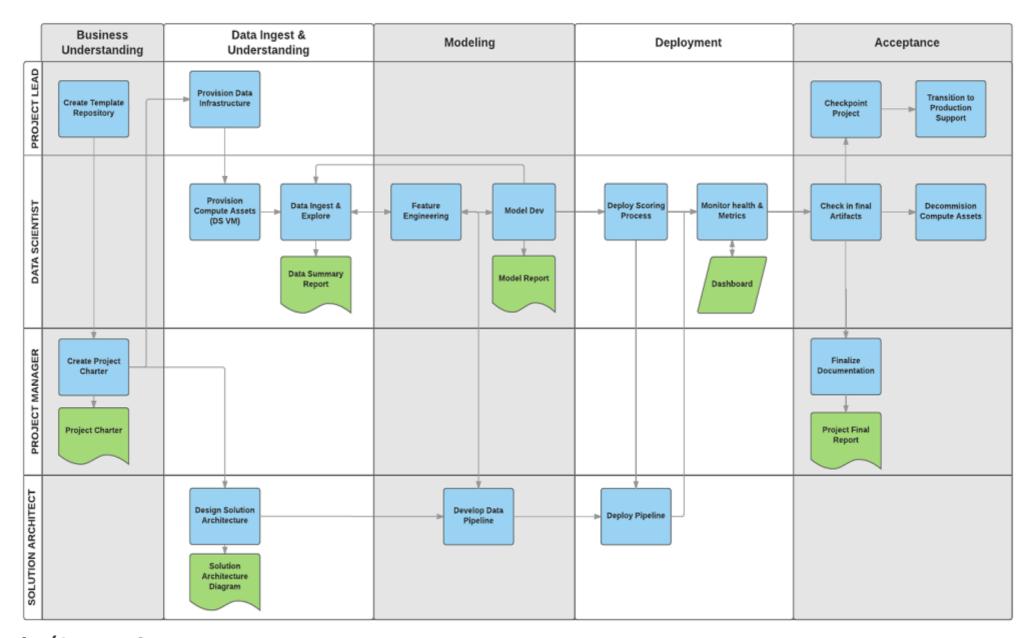
The Team Data Science Process (TDSP) is an agile, iterative data science methodology to deliver predictive analytics solutions and intelligent applications efficiently. TDSP helps improve team collaboration and learning. Microsoft: http://bit.ly/2R5WiHq



Data Science Lifecycle







Summary



Handling features

- Feature engineering
- Feature scaling
- Feature selection

How to make ML model even better

- Ensemble methods
- Hyperparameters optimization

Al as a Service

- What is it?
- Benefits and challenges

Machine Learning DevOps

- Operationalization
- Team Data Science Process

