



Tackling Fleet Test Data with MATLAB

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2 July | Europe

MathWorks

**AUTOMOTIVE
CONFERENCE 2020**

A Fleet is a Collection of Resources that Generate Data You Want to Act on



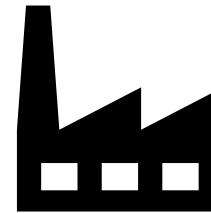
Automotive

- Vehicles
- Engines
- Controllers



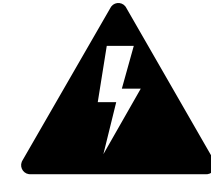
Agriculture

- Harvesters
- Tractors
- Mining



Manufacturing

- Pick & Place machines
- Welding robots
- Material handling systems

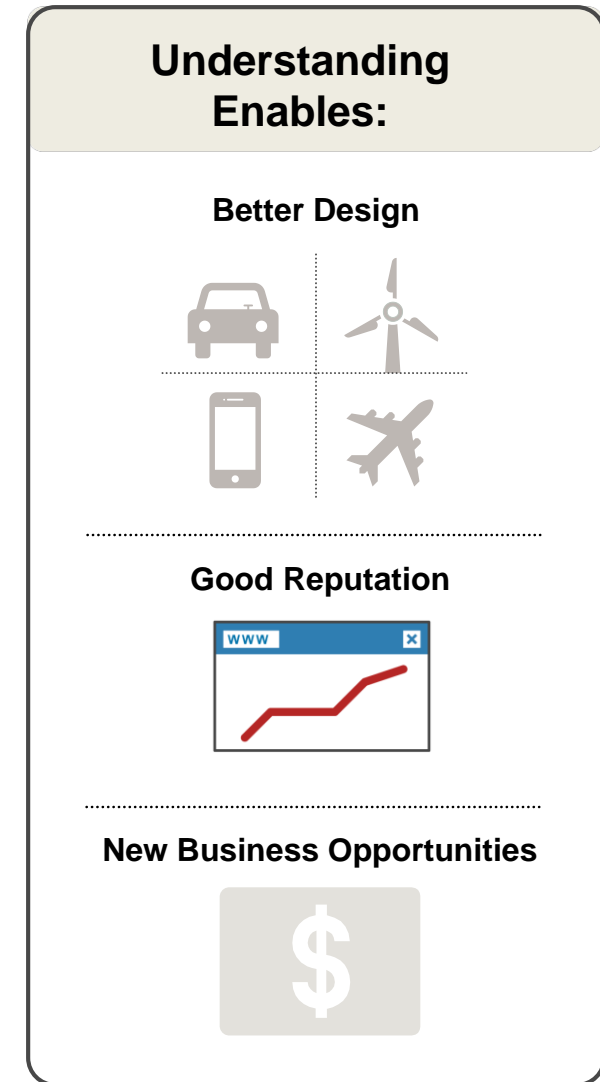


Infrastructure

- Charging stations
- Parking spaces
- Electronic toll collection

Benefits of Fleet Data Analytics

- Faster time to insights (design, testing)
- Warranty / Transparency / Reputation
- New business opportunities in the Mobility Age



Fleet Analytics in Practice: Volkswagen Data Lab

Develop technology building block for tailoring car features and services to individual

- Driver and Fleet Safety
- Driver Coaching
- Driver-Specific Insurance

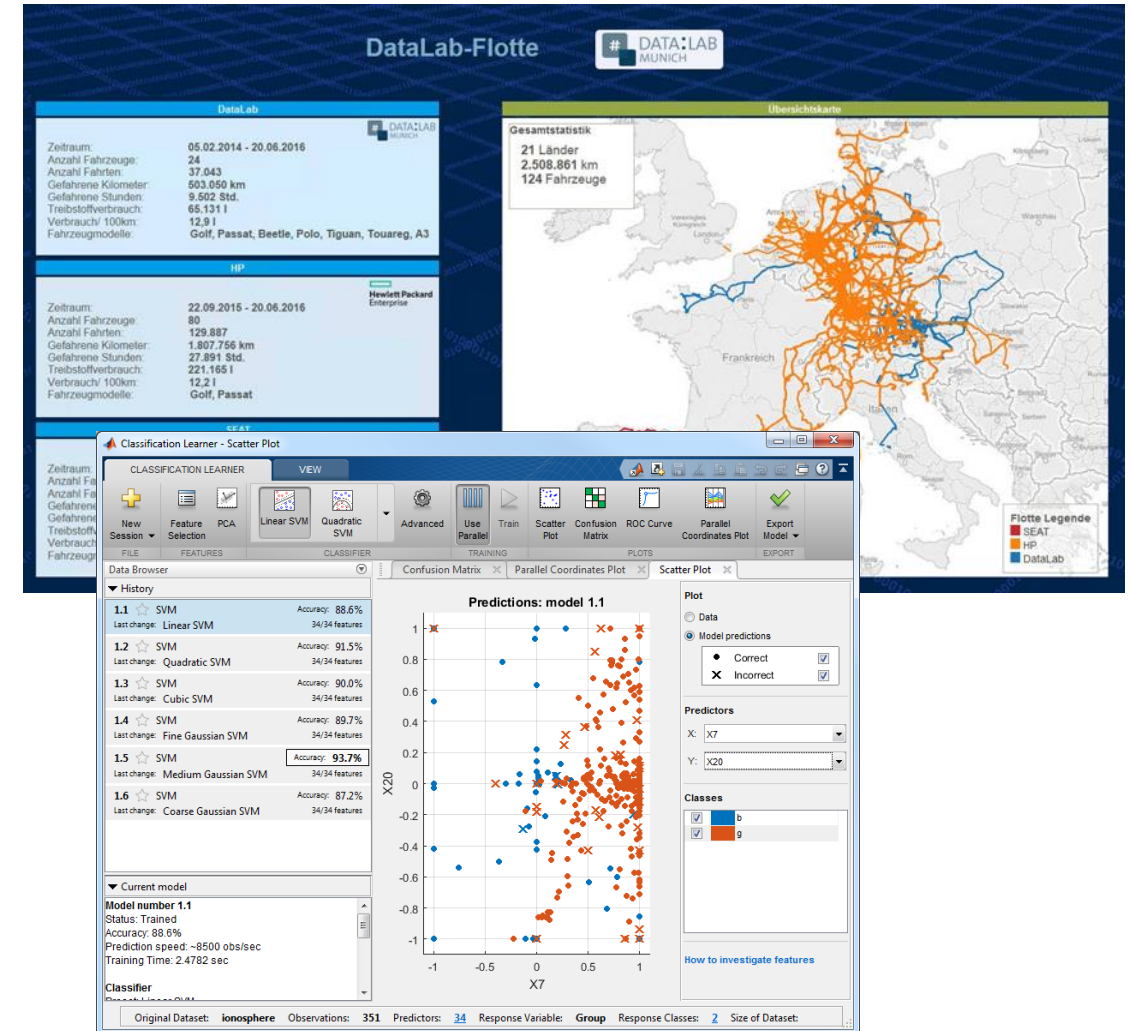
Data sources

- Logged CAN bus data and travel record

Results

- Proof-of-concept model for “telematic fingerprint”
- Basis for the “pay-as-you-drive” concept

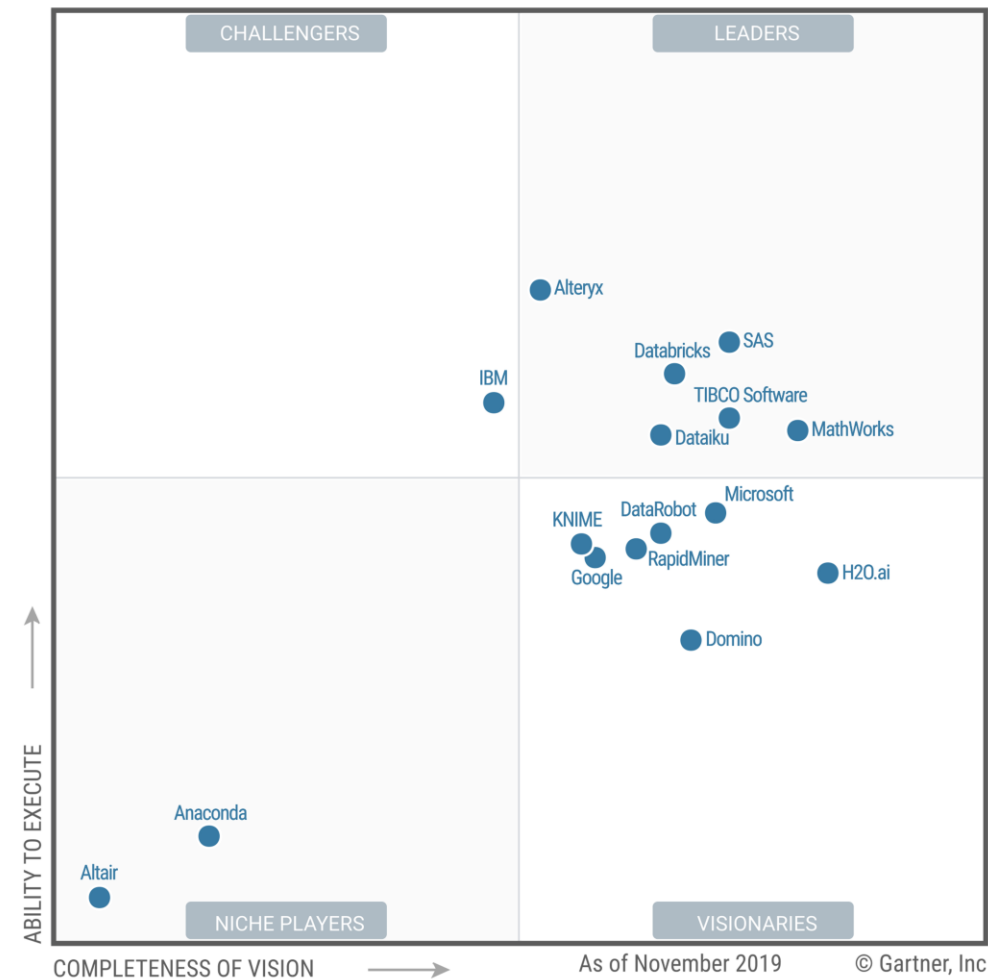
Source: [“Connected Car – Fahrererkennung mit MATLAB”](#)
 Julia Fumbarev, Volkswagen Data Lab
 MATLAB EXPO Germany, June 27, 2017, Munich Germany





is a **Leader** in the Gartner Magic Quadrant for 2020 Data Science and Machine Learning Platforms

Figure 1. Magic Quadrant for Data Science and Machine Learning Platforms

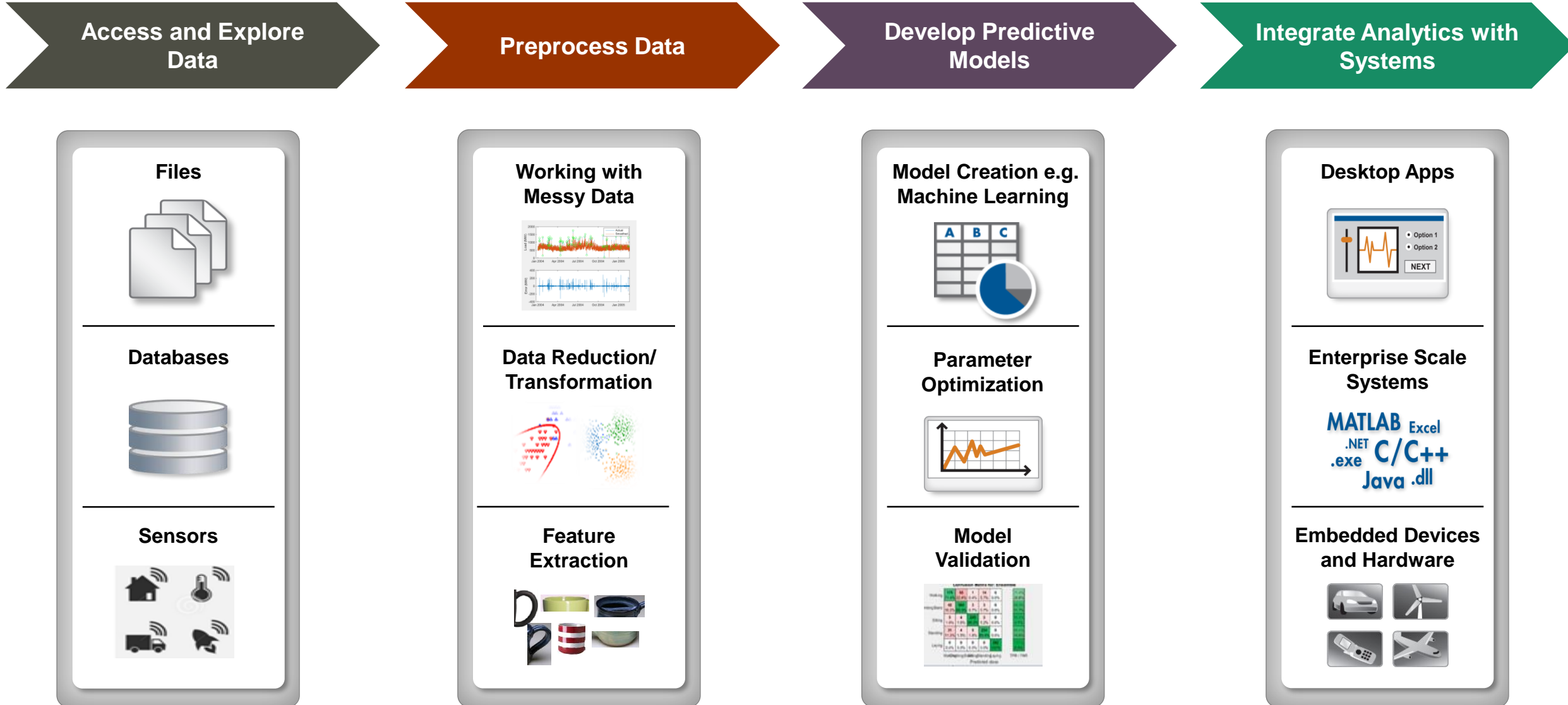


Source: Gartner (February 2020)

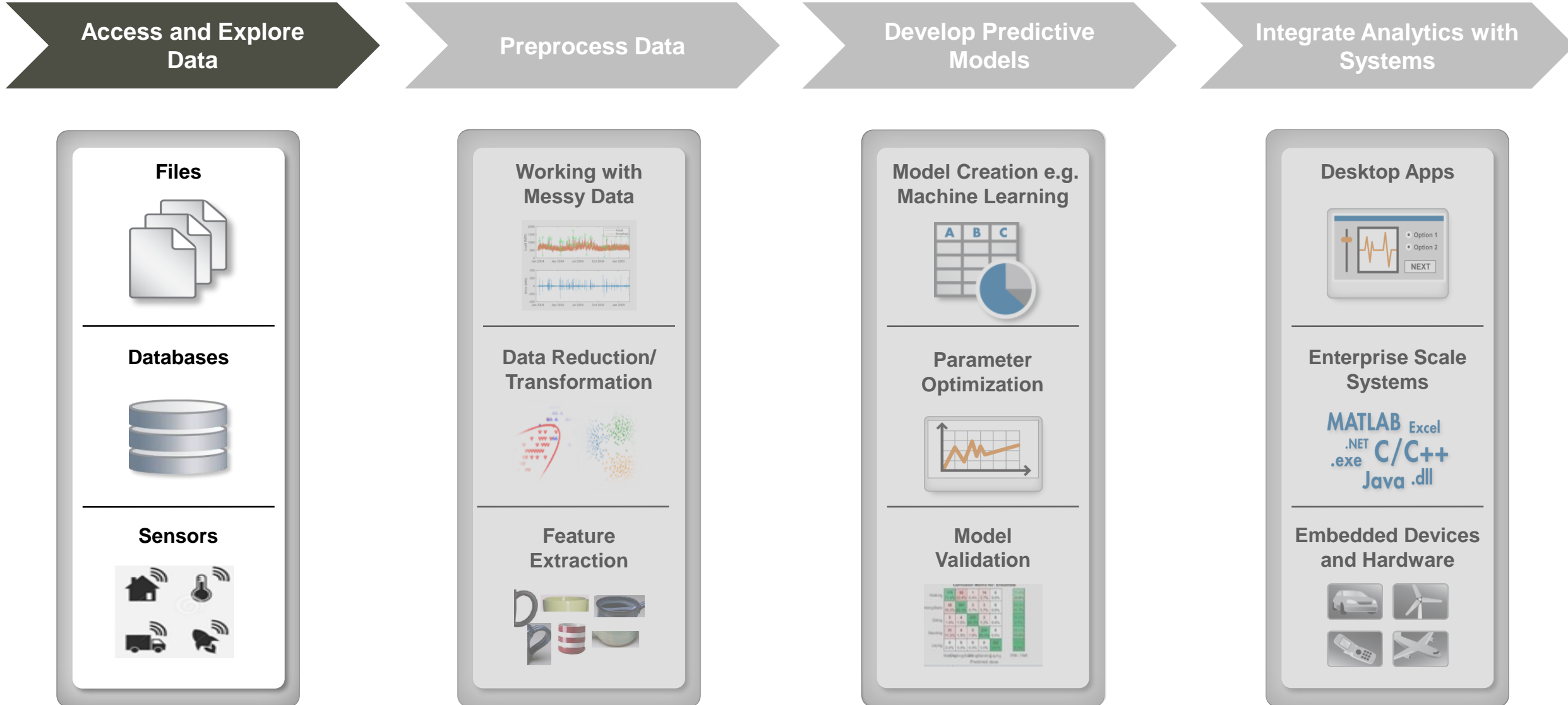
*Gartner Magic Quadrant for Data Science and Machine Learning Platforms, Peter Krensky, Erick Brethenoux, Jim Hare, Carlie Idoine, Alexander Linden, Svetlana Sicular, 11 February 2020 .

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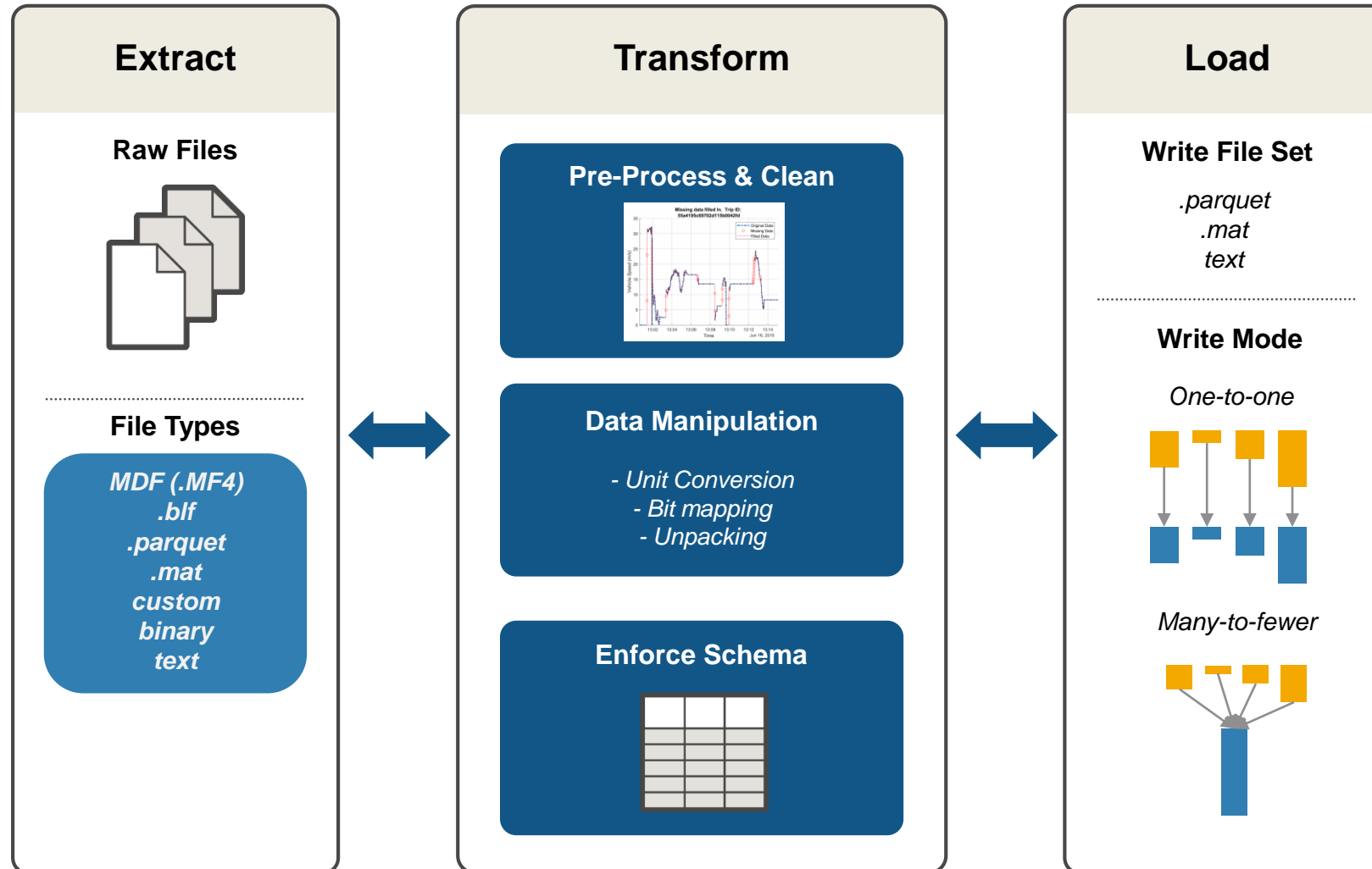
Data Analytics Workflow



Data Analytics Workflow



Extract, Transform, and Load (ETL) Workflow Considerations



Access Data in Many Formats From Many Locations

- Type
- Structure
- Location



Access Data From Anywhere With Minimal Changes



```
setenv("AWS_ACCESS_KEY_ID",id)
```

```
setenv("HADOOP_HOME",hadoopPath)
```

```
fileLoc = "hdfs://hadoop01glnxa64:123/datasets/FoodPic.jpg"
```

```
img = imread(fileLoc);
```

Scale to Large Collections of Data with Datastore

Create a datastore from all CSV files

```
ds = datastore('*.*csv')
```

Read a single file of data

```
data = read(ds);
```

Reset the datastore back to the first file

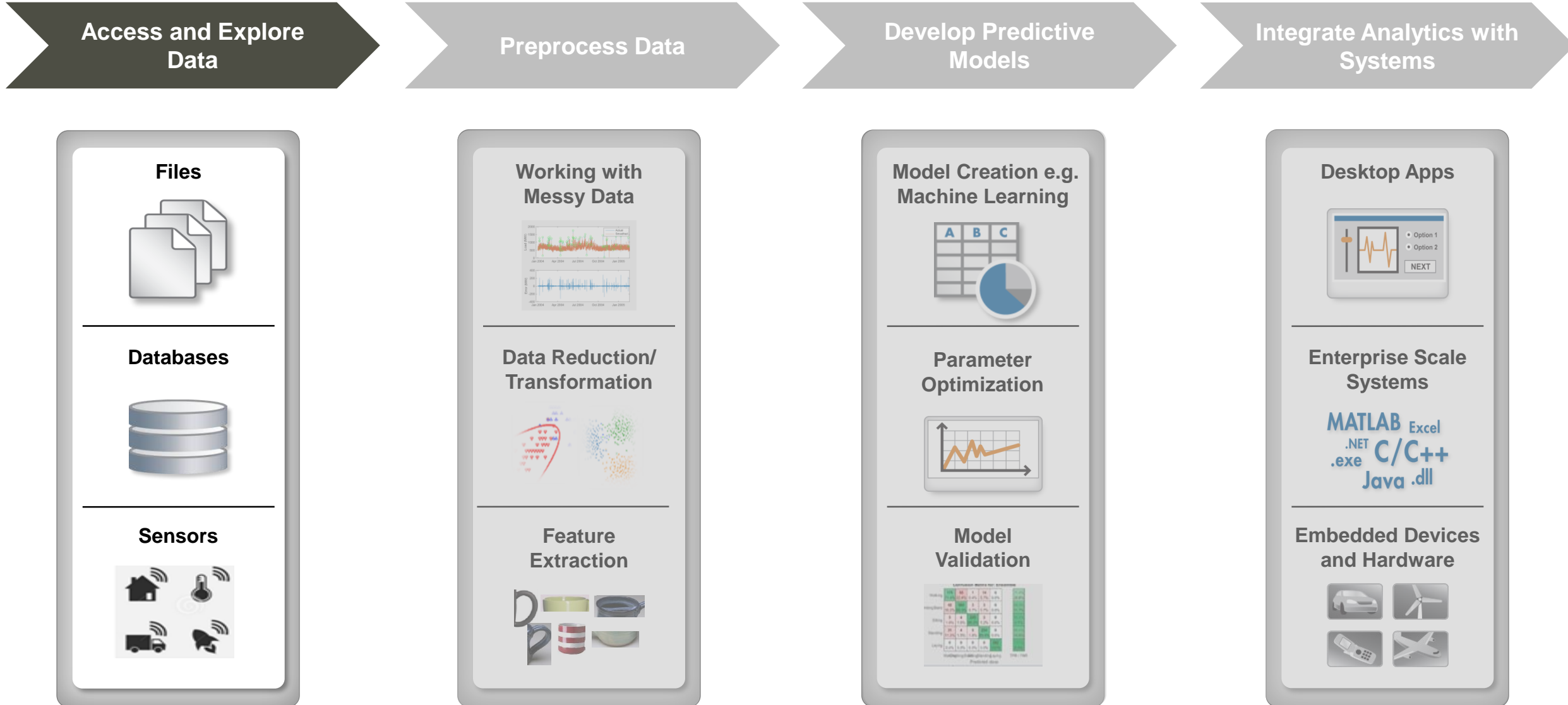
```
reset(ds);
```

Find the maximum value of “Y” in each file

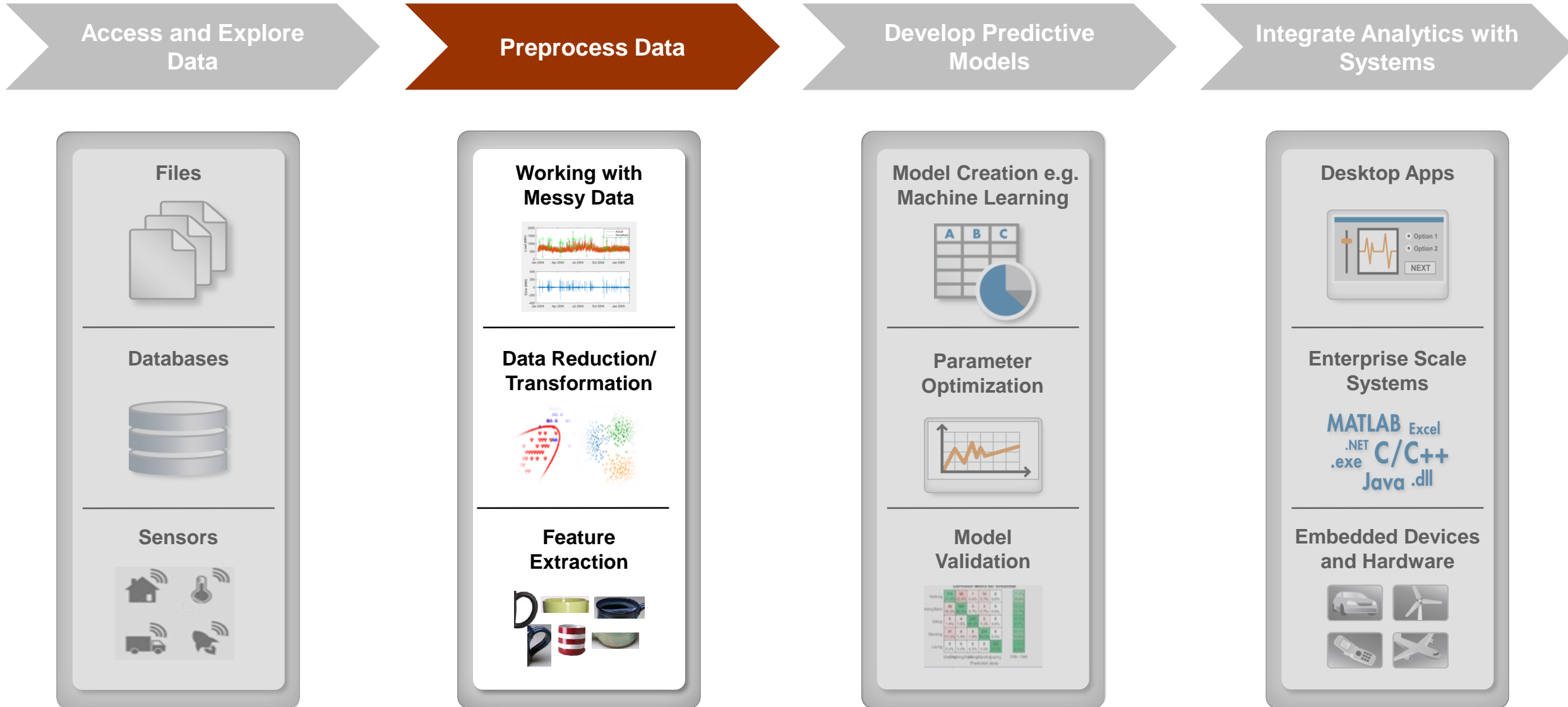
```
X = [];
while hasdata(ds)
    data = read(ds);
    X(end+1) = max(data.Y);
end
```

Available Datastores	
General	datastore
	spreadsheetDatastore
	tabularTextDatastore
	fileDatastore
Database	databaseDatastore
Image	imageDatastore
	denoisingImageDatastore
	randomPatchExtractionDatastore
	pixelLabelDatastore
	augmentedImageDatastore
Audio	audioDatastore
Predictive Maintenance	fileEnsembleDatastore
	simulationEnsembleDatastore
Simulink	SimulationDatastore
Automotive	mdfDatastore
Custom	subclass matlab.io.Datastore
Transformed	transform an existing datastore

Data Analytics Workflow



Data Analytics Workflow



Preprocess and Explore Data in Few Lines of Code

Import

```
t1 = readtimetable("s3://bucket_name/file.txt");
```

Preprocess

```
t = synchronize(t1,t2,t3);  
t = fillmissing(t,"linear");  
t = rmoutliers(t);  
t = smoothdata(t,"movmedian");  
t = normalize(t);
```

Explore

```
top5 = topkrows(t,5,"RH");  
byTime = groupsummary(t,"Time","year","mean");  
scaled = grouptransform(t,"State","rescale");  
chgpts = ischange(t,"variance","Threshold",20);
```

Visualize

```
stackedplot(t);  
geoplot(t.Lat,t.Lon,t.RH);  
heatmap(t,"State","AQILabel");  
scatterhistogram(t.RH,t.DP);
```


Use Dedicated Functions for Common Preprocessing Tasks

- Synchronize by time
- Find, fill, and remove missing
- Work with outliers
- Smooth noisy data
- Normalize, rescale data

Preprocess

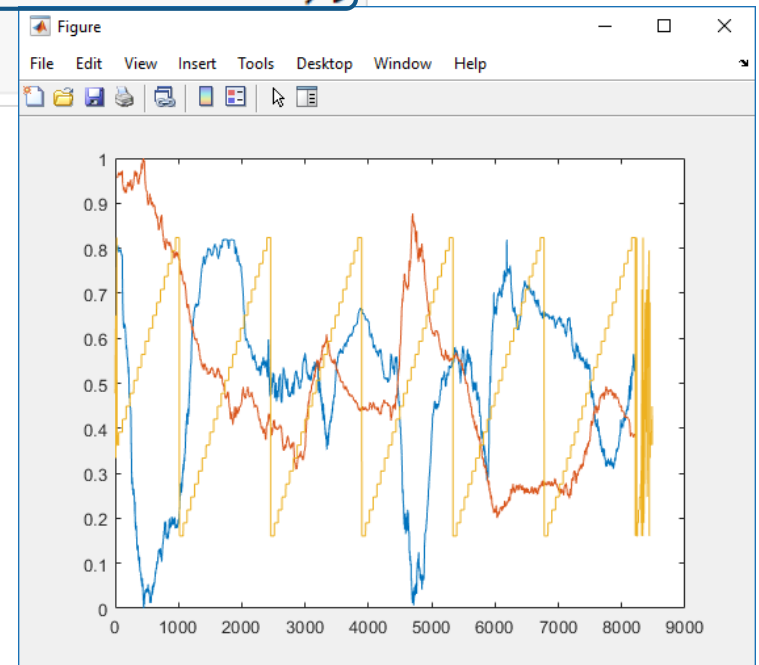
```
t = synchronize(t1,t2,t3);
```

```
t = fillmissing(t, "linear");
```

```
t = rmoutliers(t);
```

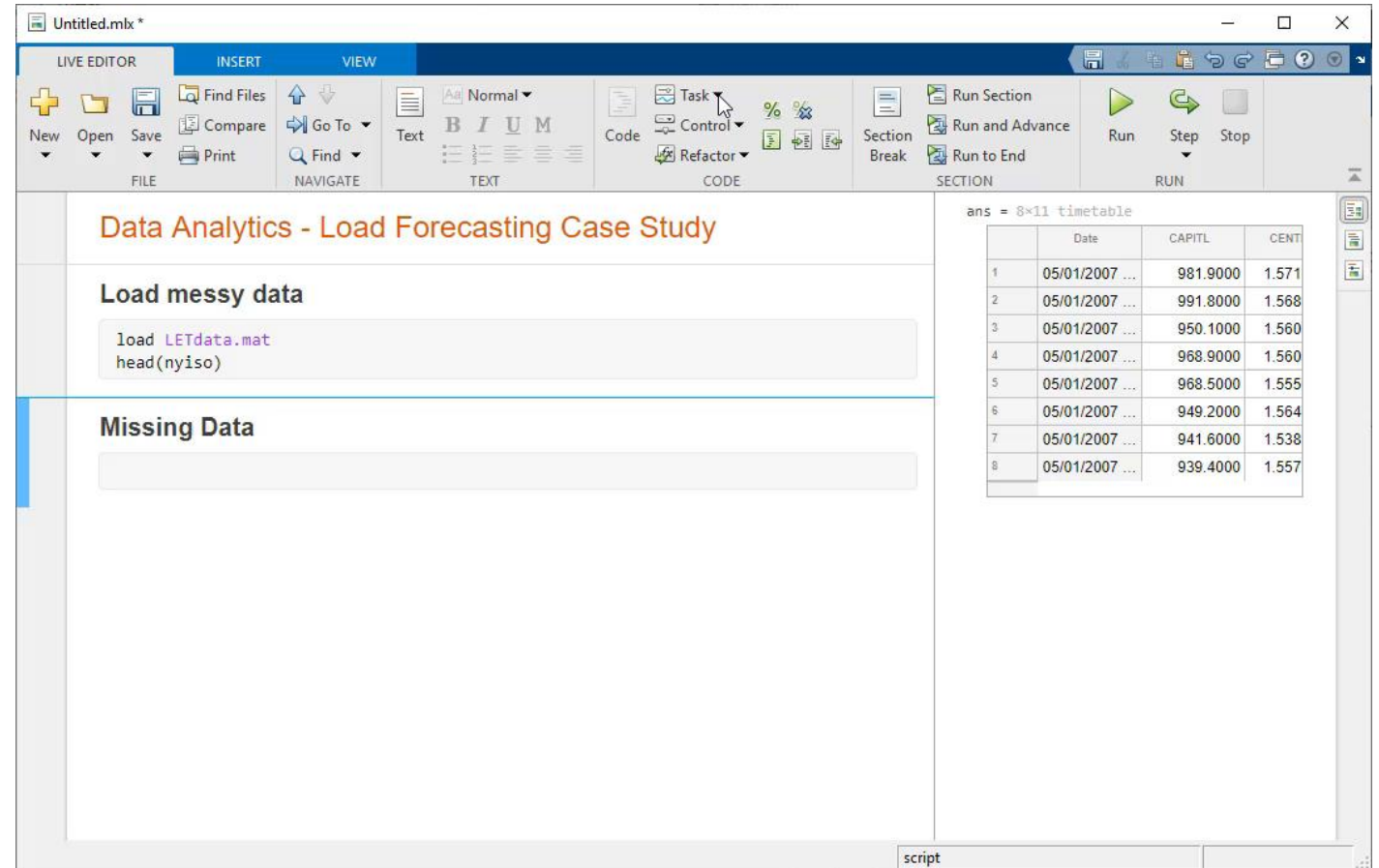
```
t = smoothdata(t, "movmedian");
```

```
t = normalize(t);
```



Explore Settings Quickly With Live Tasks

- Tasks are apps that can be included in scripts
- Preprocessing tasks allow you to:
 - Interactively explore parameters and options
 - Preview results based on those parameters and options
- Automatically generate the corresponding MATLAB code
- Save the task as part of the script for subsequent use by others

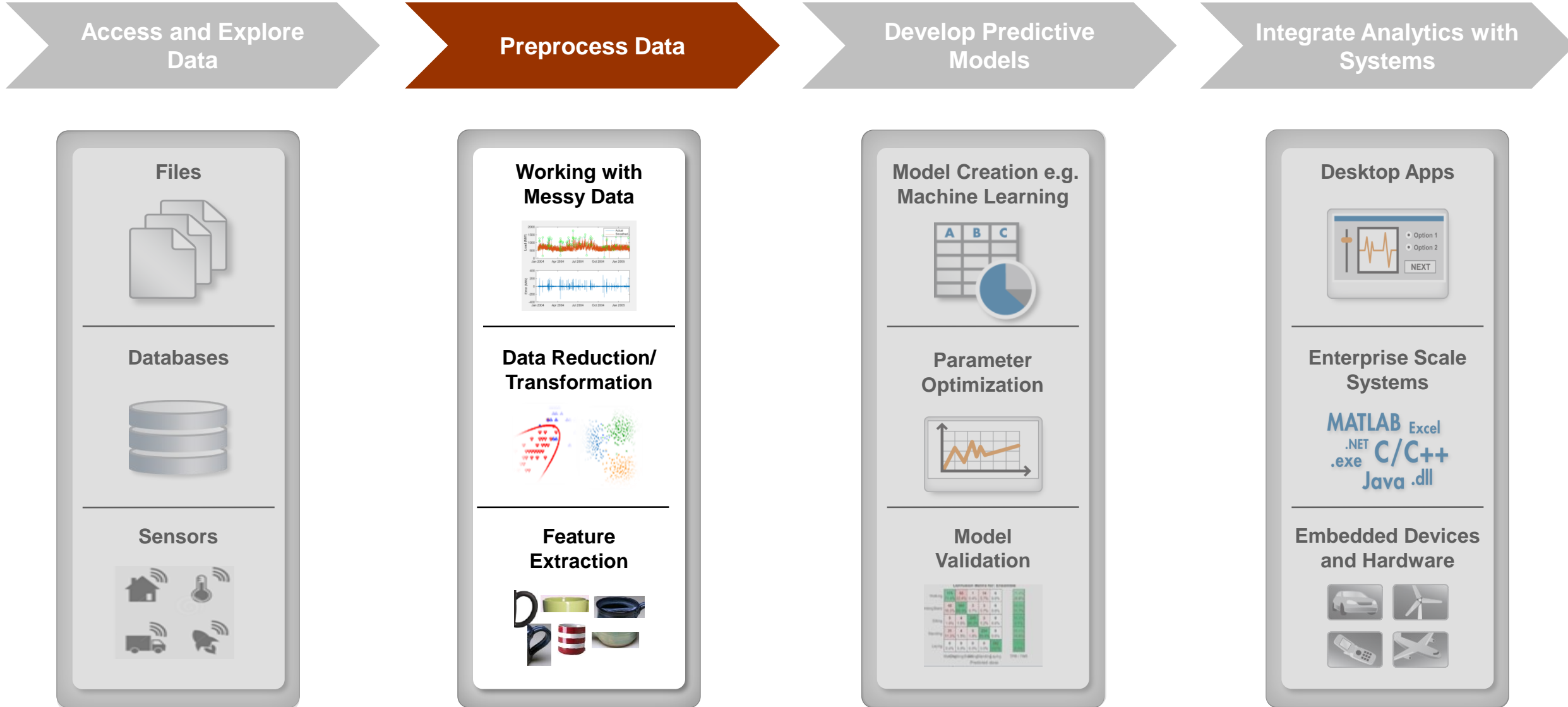


The screenshot displays the MATLAB Live Editor interface for an 'Untitled.mlx' file. The interface is divided into several sections:

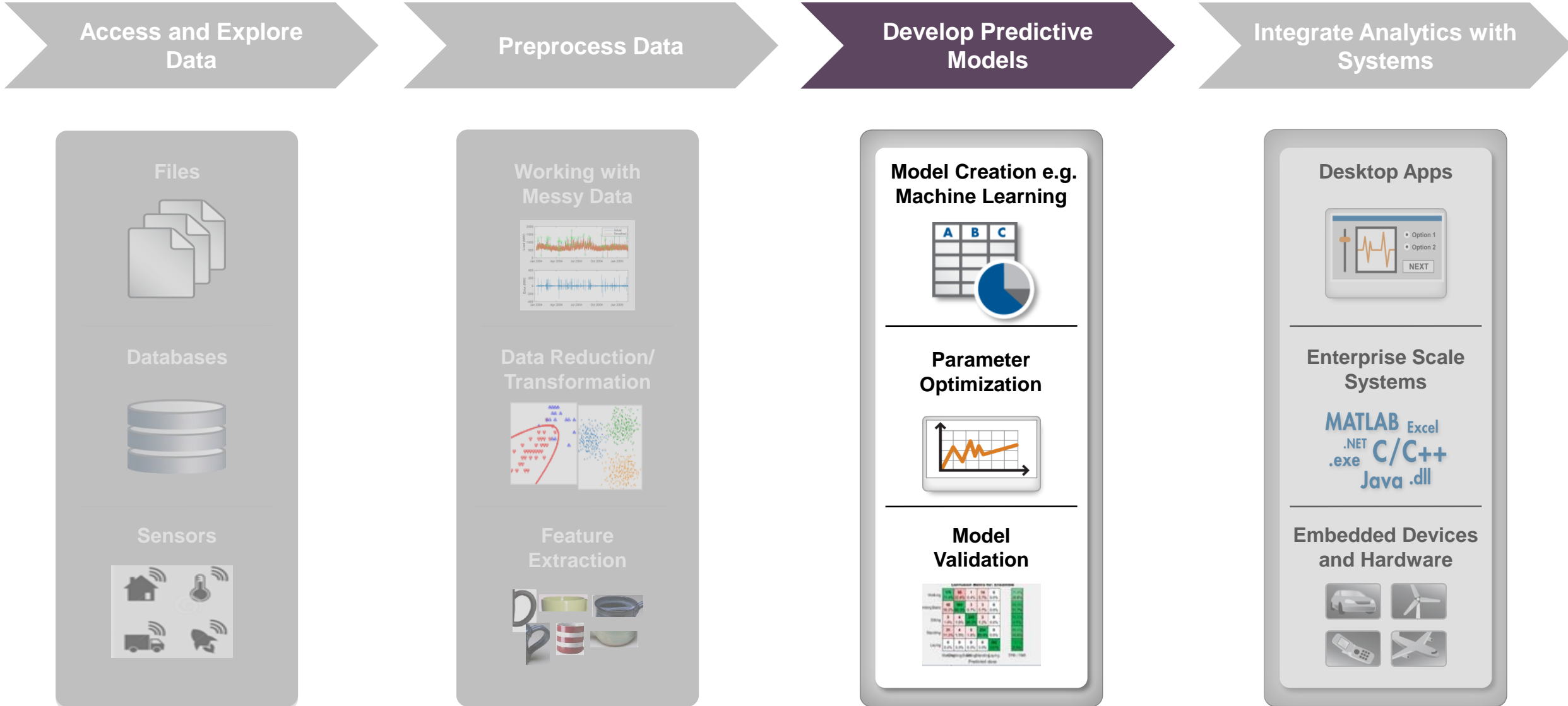
- Toolbar:** Includes 'LIVE EDITOR', 'INSERT', and 'VIEW' tabs. The 'LIVE EDITOR' tab is active, showing icons for 'New', 'Open', 'Save', 'Compare', 'Print', 'Go To', 'Find', 'Text', 'Code', 'Task', 'Control', 'Refactor', 'Section Break', 'Run Section', 'Run and Advance', 'Run to End', 'Run', 'Step', and 'Stop'.
- Task Title:** 'Data Analytics - Load Forecasting Case Study'.
- Task Content:**
 - Load messy data:** Contains the MATLAB code `load LETdata.mat` and `head(nyiso)`.
 - Missing Data:** A section for handling missing data, currently empty.
- Output Window:** Shows the results of the 'Load messy data' task, displaying an 8x11 timetable with columns 'Date', 'CAPITL', and 'CENT'. The data is as follows:

	Date	CAPITL	CENT
1	05/01/2007 ...	981.9000	1.571
2	05/01/2007 ...	991.8000	1.568
3	05/01/2007 ...	950.1000	1.560
4	05/01/2007 ...	968.9000	1.560
5	05/01/2007 ...	968.5000	1.555
6	05/01/2007 ...	949.2000	1.564
7	05/01/2007 ...	941.6000	1.538
8	05/01/2007 ...	939.4000	1.557

Data Analytics Workflow



Data Analytics Workflow

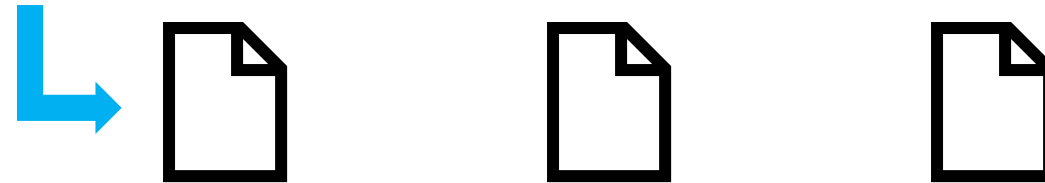


Automotive Vehicle Fleet – Intrinsic Hierarchy

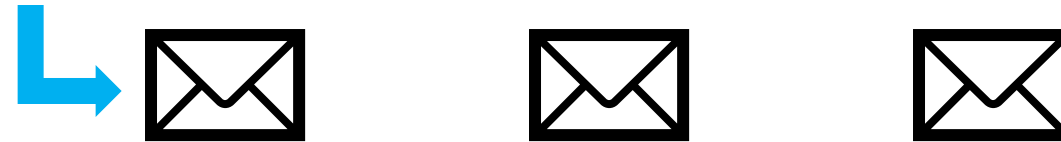
Vehicles



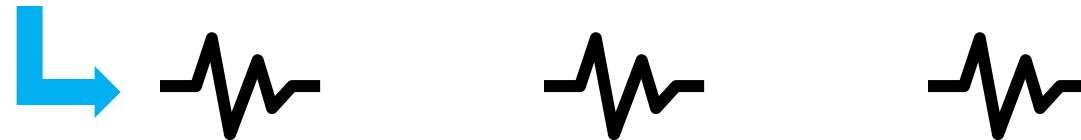
Trips (files)



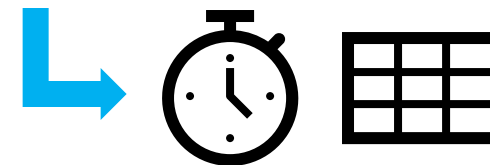
Messages



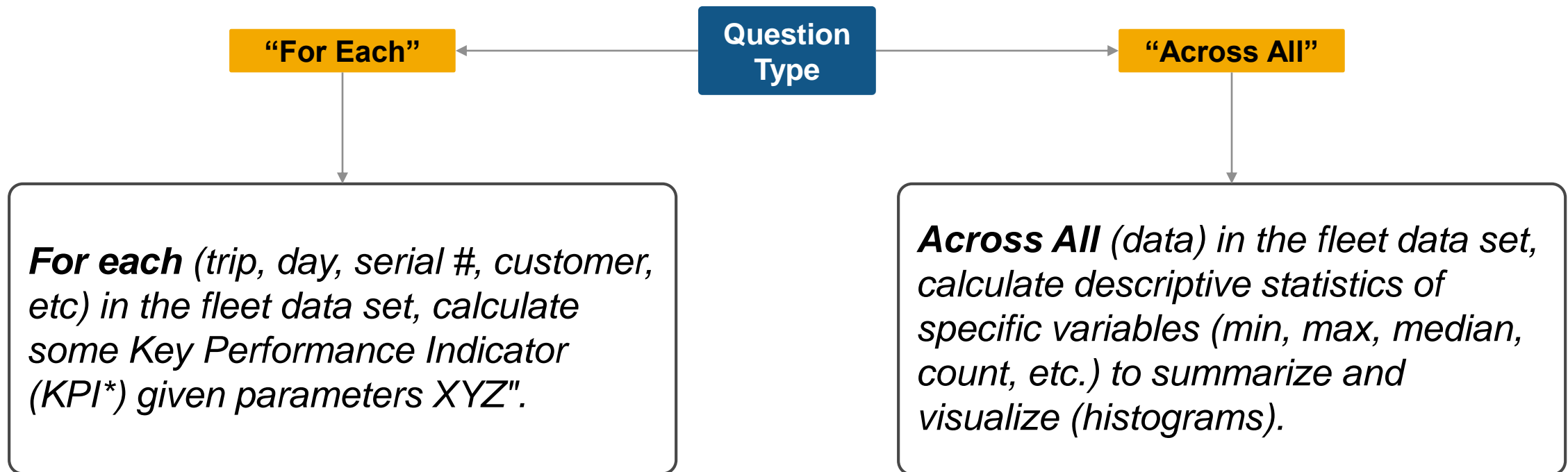
Signals



Time – Value pairs



Fleet Analytics Questions Fall Into Two Broad Categories



Scale Up to Big Data Without Big Changes

One file

Access Data

```
measured = readtable('PumpData.csv');
measured = table2timetable(measured);
```

Preprocess Data

Select data of interest

```
measured = measured(timerange(seconds(1),seconds(2)), 'Speed')
```

Work with missing data

```
measured = fillmissing(measured, 'linear');
```

Calculate statistics

```
m = mean(measured.Speed);
s = std(measured.Speed);
```

One hundred files

Access Data

```
measured = datastore('PumpData*.csv');
measured = tall(measured);
measured = table2timetable(measured);
```

Preprocess Data

Select data of interest

```
measured = measured(timerange(seconds(1),seconds(2)), 'Speed')
```

Work with missing data

```
measured = fillmissing(measured, 'linear');
```

Calculate statistics

```
m = mean(measured.Speed);
s = std(measured.Speed);
```

```
[m,s] = gather(m,s);
```

Big Data File-Based Workflows

```
function events = detectEvents(t)
<fill in analytics code here>
end
```



File set location
(C:\, NAS/NFS, S3, Blob, hdfs://...)

Datastore

```
ds = parquetDatastore("hdfs://myData/Test")
```

Question Type

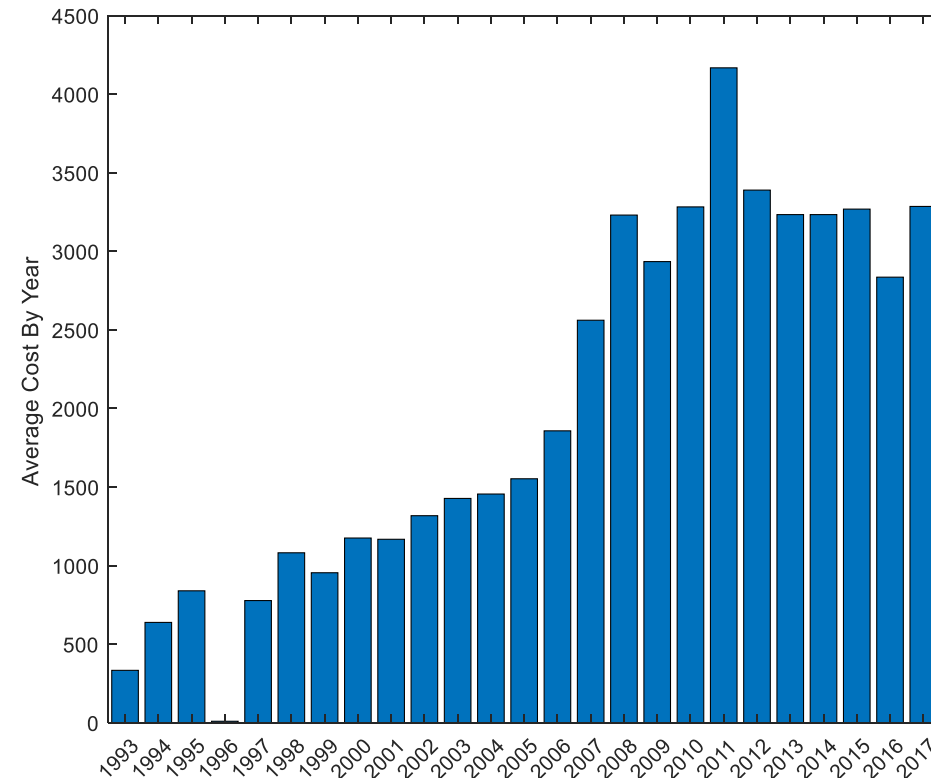
“For Each”

```
tds = transform(ds, @detectEvents);
TT = tall(tds);
EventsSummary = gather(TT);
```

Compute Groupwise Metrics and Detect Events

- Explore range
- Grouped calculations
- Detect local minima and maxima
- Detect abrupt changes in data with `ischange`

```
top5 = topkrows(t,5,"RH");  
byTime = groupsummary(t,"Time","year","mean");  
scaled = grouptransform(t,"State","rescale");  
chgpts = ischange(t,"variance","Threshold",20);
```



Big Data File-Based Workflows

```
function events = detectEvents(t)
<fill in analytics code here>
end
```



File set location
(C:\, NAS/NFS, S3, Blob, hdfs://...)

Datastore

```
ds = parquetDatastore("hdfs://myData/Test")
```

Question Type

“For Each”

“Across All”

```
tds = transform(ds, @detectEvents);
TT = tall(tds);
EventsSummary = gather(TT);
```

```
TT = tall(ds);
MedianValSigABC = gather(median(TT.SignalABC));
```

Perform “Across All” Calculations with Tall Tables

Create a datastore from a collection of CSV files, and select the "Time" and "EngineSpeedRPM" variables.

```
ds = datastore('EngineData*.csv',...  
             "SelectedVariableNames",["Time", "EngineSpeedRPM"]);
```

Create tall table:

```
t = tall(ds);
```

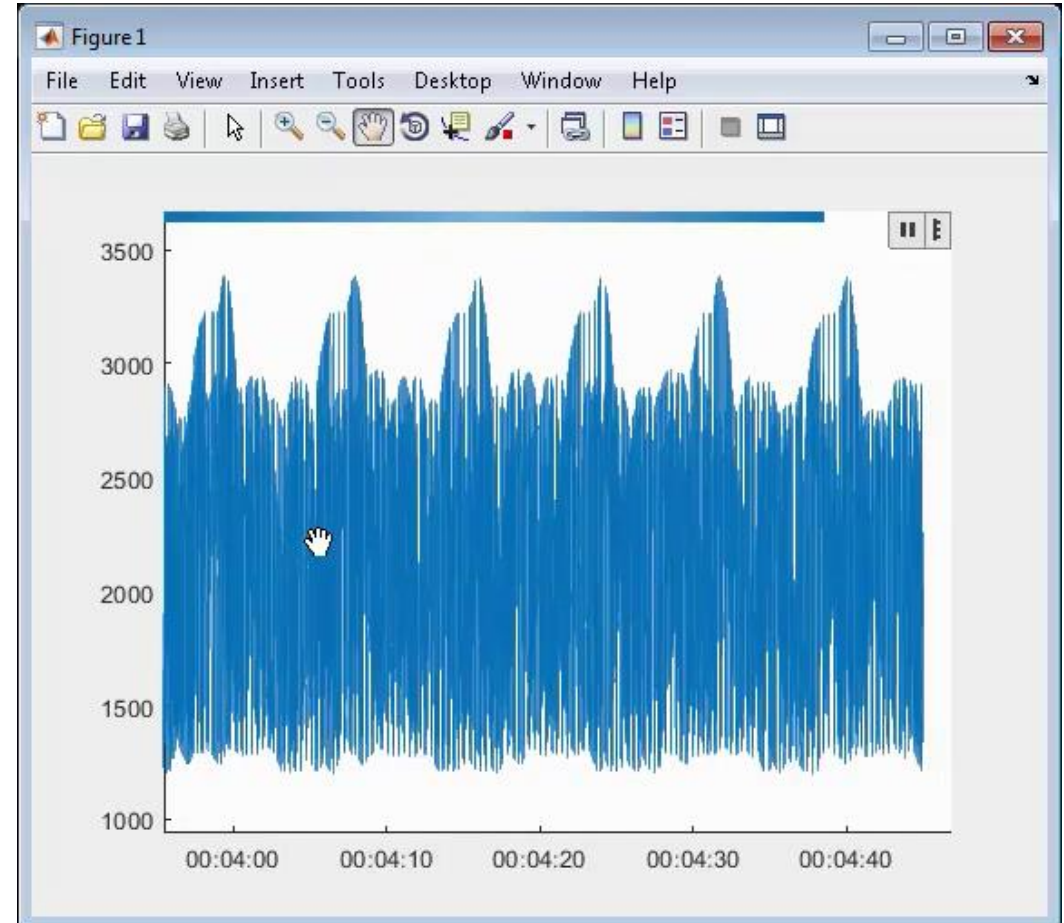
Convert to tall timetable:

```
tt = table2timetable(t);
```

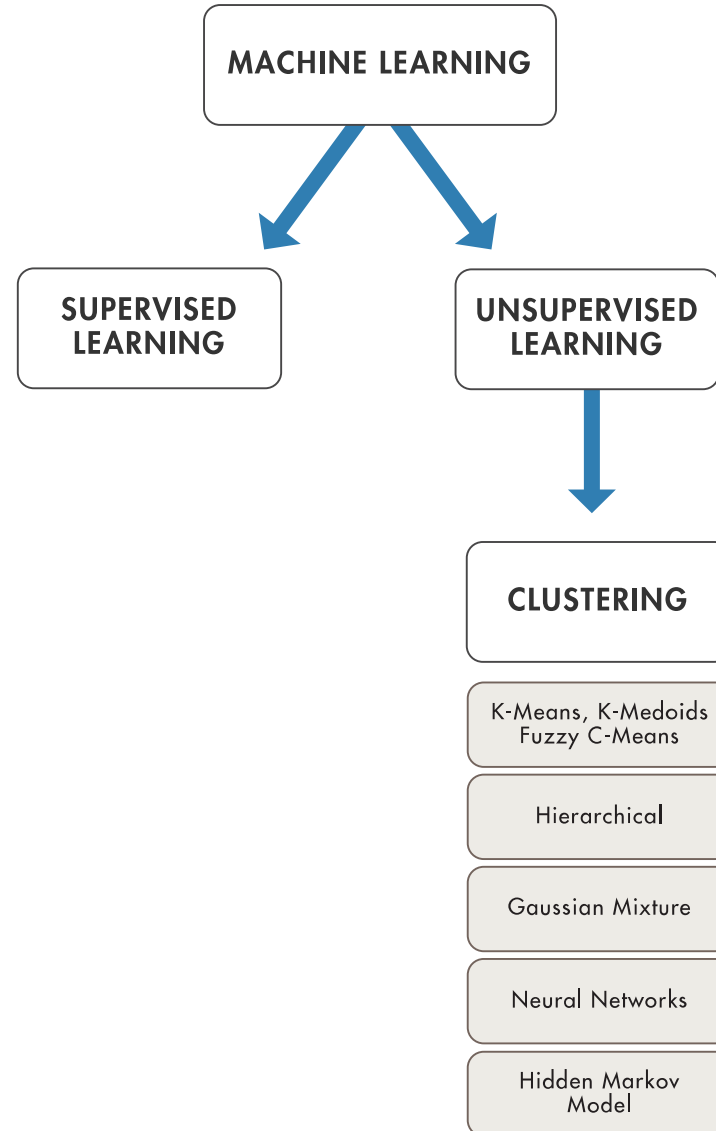
Plot EngineSpeedRPM vs. Time:

```
plot(tt.Time,tt.EngineSpeedRPM)
```

- Visualizations
- Data preprocessing
- Machine Learning



Explore Fleet Data with Unsupervised Learning



Unsupervised Learning for Operational Mode Clustering

Plot the raw data:

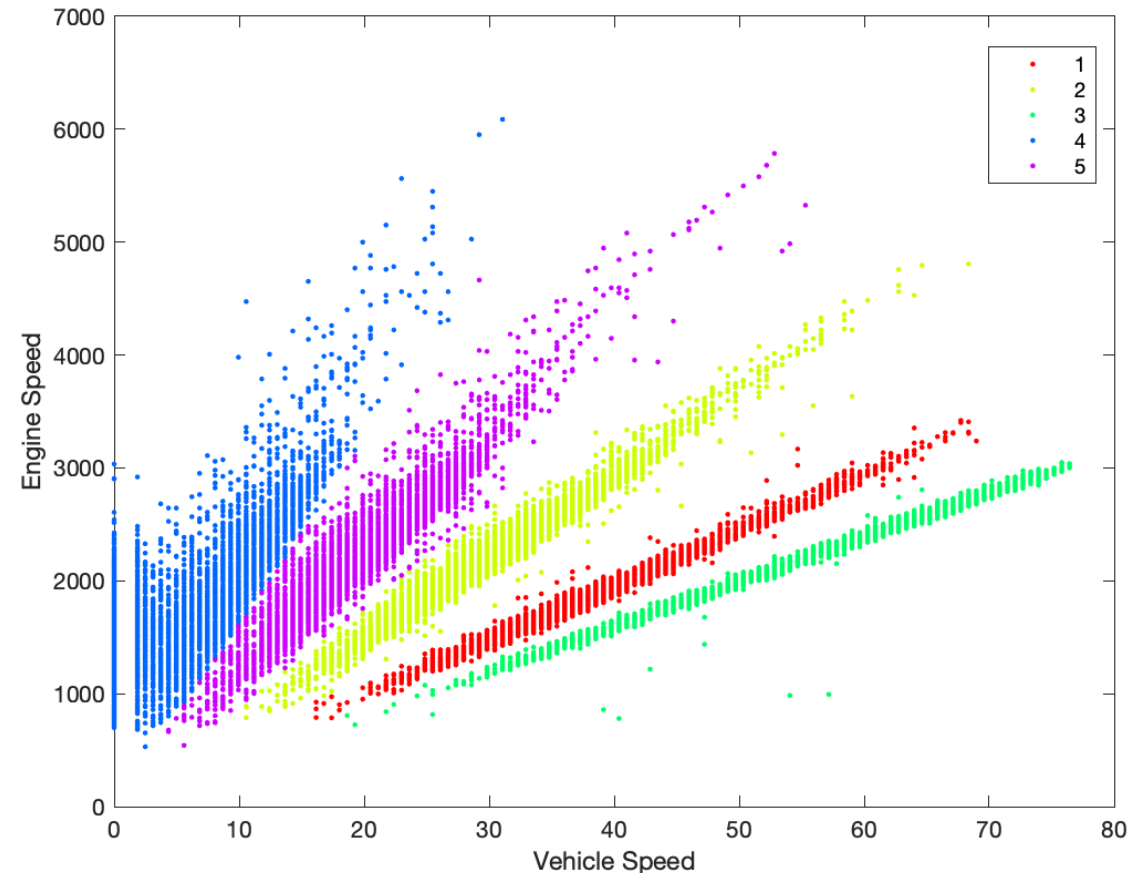
```
figure;  
plot(t.Speed_OBD_,t.EngineRPM, '.k')  
xlabel('Vehicle Speed');  
ylabel('Engine Speed');
```

Cluster the data with the K-Means algorithm:

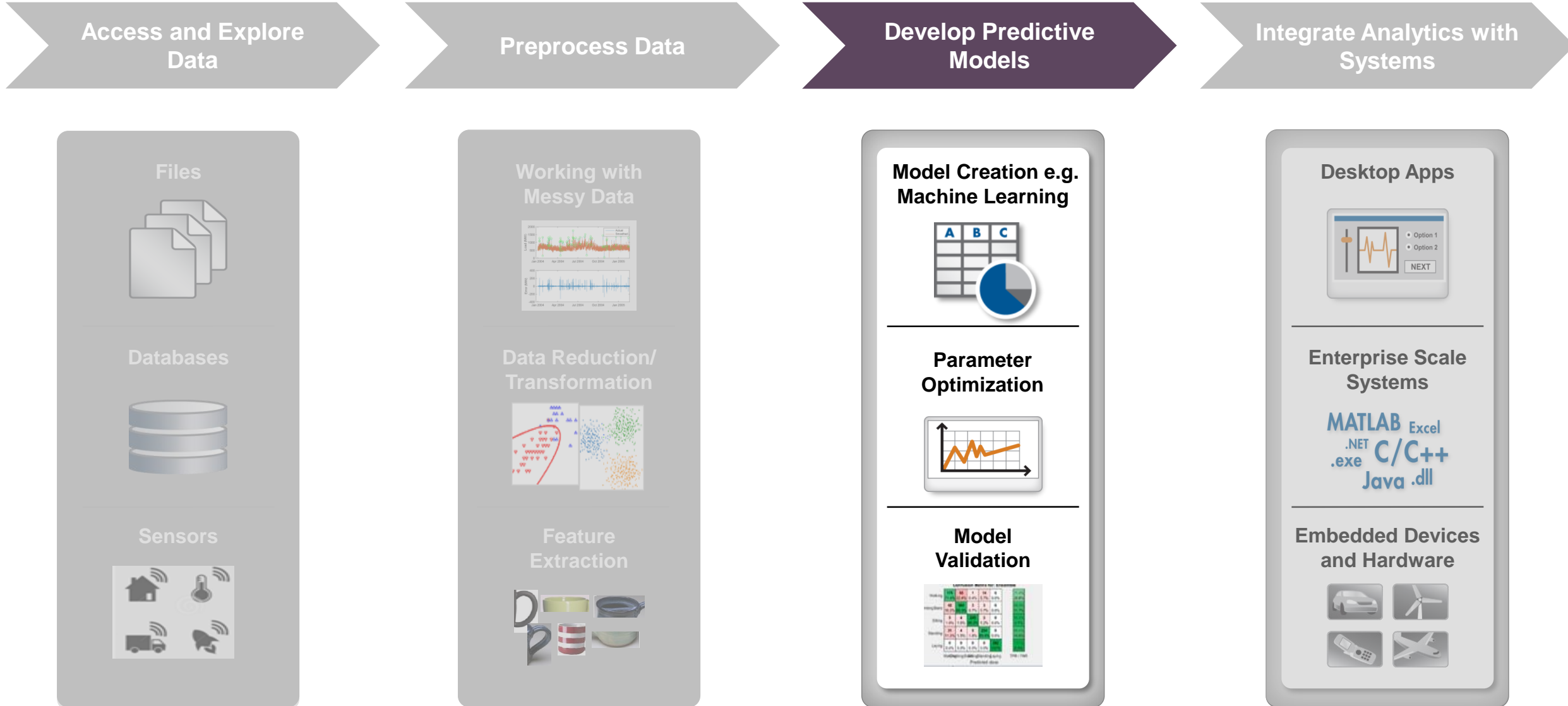
```
X = [t.Speed_OBD_,t.EngineRPM];  
IDX = kmeans(X,5,"Distance","cosine");
```

Plot results of the clustering:

```
gscatter(t.Speed_OBD_,t.EngineRPM,IDX);  
xlabel('Vehicle Speed');  
ylabel('Engine Speed');
```



Data Analytics Workflow



Data Analytics Workflow

Access and Explore Data

Preprocess Data

Develop Predictive Models

Integrate Analytics with Systems

Files



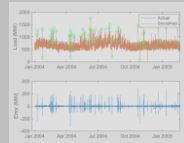
Databases



Sensors



Working with Messy Data



Data Reduction/Transformation



Feature Extraction



Model Creation e.g. Machine Learning



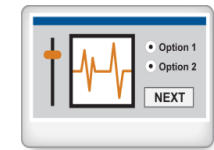
Parameter Optimization



Model Validation



Desktop Apps



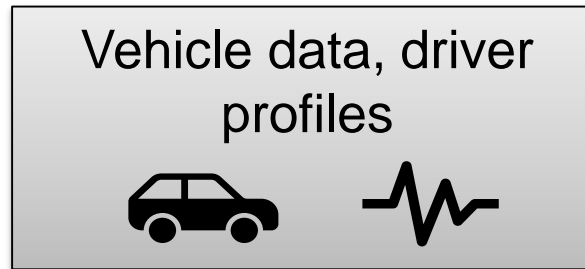
Enterprise Scale Systems

MATLAB Excel
.NET C/C++
.exe Java .dll

Embedded Devices and Hardware



Deploying Fleet Analytics



“Cold Storage”

Historic data:

- **Batch processing**
- Large data on cluster
- Explore long term trends
- Build models



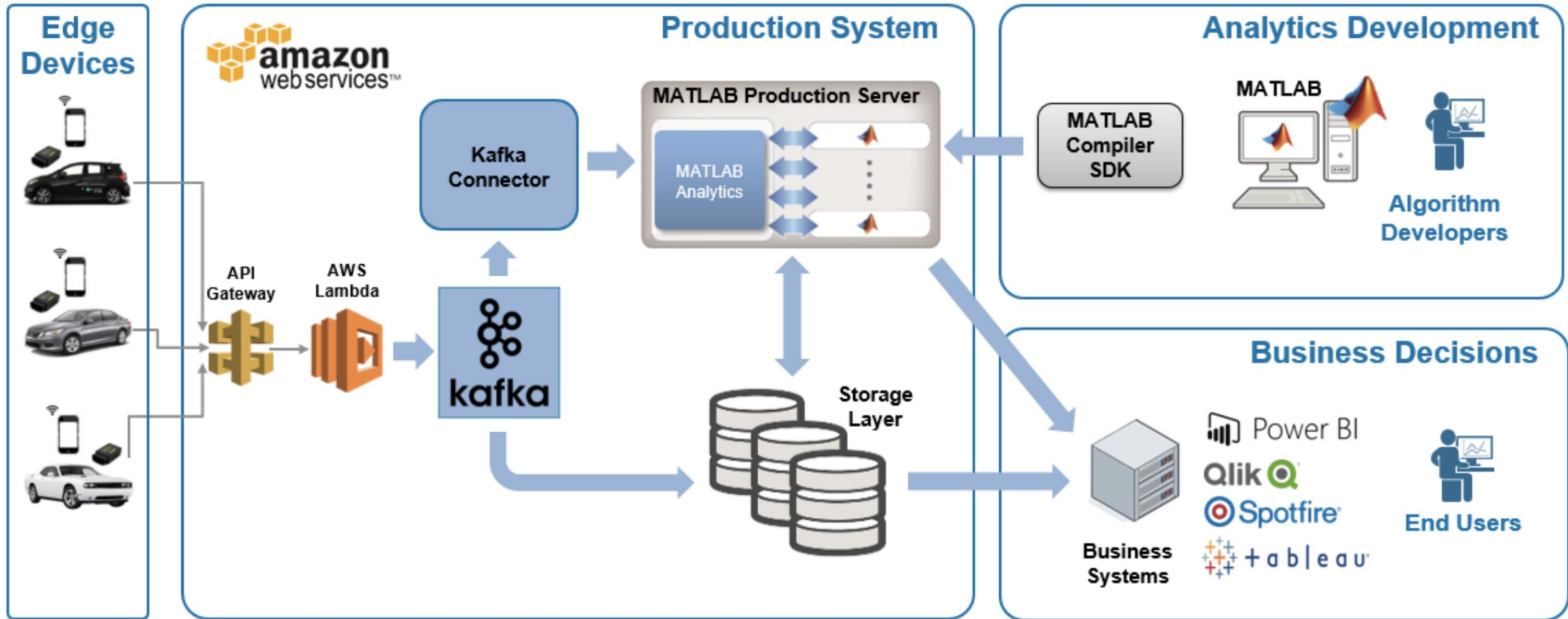
“Hot Storage”

Streaming data:

- **Near real-time**
- Test and implement model for new data
- Stream processing

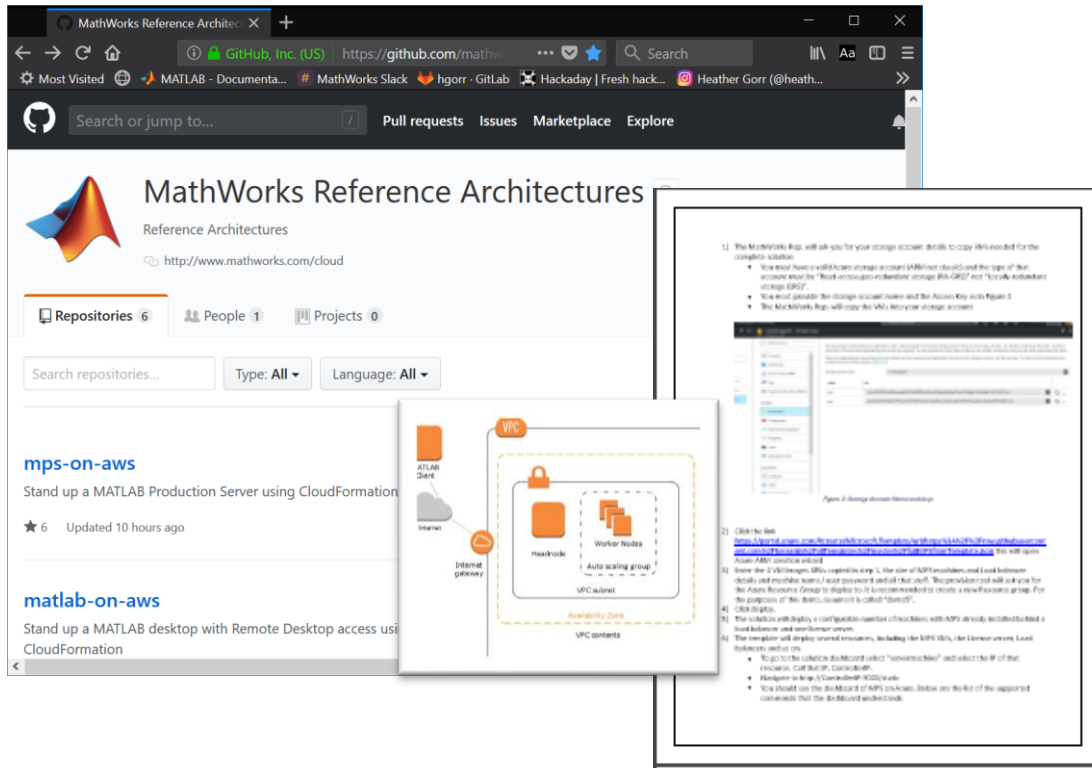


Fleet Analytics Streaming Architecture



Use Reference Architectures to Run MATLAB on Cloud Platforms

GitHub



<https://github.com/mathworks-ref-arch/>

Data Analytics Workflow



Files

Databases

Sensors

Working with Messy Data

Data Reduction/Transformation

Feature Extraction

Model Creation e.g. Machine Learning

Parameter Optimization

Model Validation

Desktop Apps

Enterprise Scale Systems

MATLAB Excel
 .NET C/C++
 .exe Java .dll

Embedded Devices and Hardware

MATLAB Training Courses for Data Science

Get Started for Free



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Get started quickly with the basics of MATLAB®.

» Details and launch



Machine Learning Onramp

An interactive introduction to practical machine learning methods for classification problems.

» Details and launch



Deep Learning Onramp

Get started with deep learning techniques to perform image recognition.

» Details and launch

Training Courses

MATLAB Fundamentals (3 days)

MATLAB for Data Processing and Visualization (1 day)

Processing Big Data with MATLAB (1 day)

Statistical Methods in MATLAB (2 days)

Machine Learning with MATLAB (2 days)

Signal Preprocessing and Feature Extraction with MATLAB (1 day)

Deep Learning with MATLAB (2 days)

Accelerating and Parallelizing MATLAB Code (2 days)

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★★★★★ 4.9 14 ratings

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Financial aid available

- Exploratory Data Analysis
- Data Processing and Feature Engineering
- Predictive Modeling and Machine Learning
- Data Science Project



is a **Leader** in the Gartner Magic Quadrant for 2020 Data Science and Machine Learning Platforms

MathWorks has the furthest completeness of vision in the Leaders quadrant

We believe this recognition demonstrates our ability to:

- Empower your team, including those with limited AI or data science experience
- Provide complete workflows for data preparation, AI modeling, system design, and production
- Deploy AI models on embedded devices, edge, enterprise systems, and the cloud
- Use Simulink to tackle integration challenges and reduce risk in designing AI-driven systems



*Gartner Magic Quadrant for Data Science and Machine Learning Platforms, Peter Krensky, Erick Brethenoux, Jim Hare, Carlie Idoine, Alexander Linden, Svetlana Sicular, 11 February 2020 .

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Q&A

On which part of the data analytics workflow do you spend most time?

- a Data access
- b Data preprocessing
- c Data analysis and modeling
- d Enterprise integration

Please contact us with questions



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sbomberg@mathworks.com