



A data-driven prognostics for an assembly machine for automatic transmissions

Prof. Hyunbo Cho, Ph.D.

hcho@postech.ac.kr

Industrial & Management Engineering

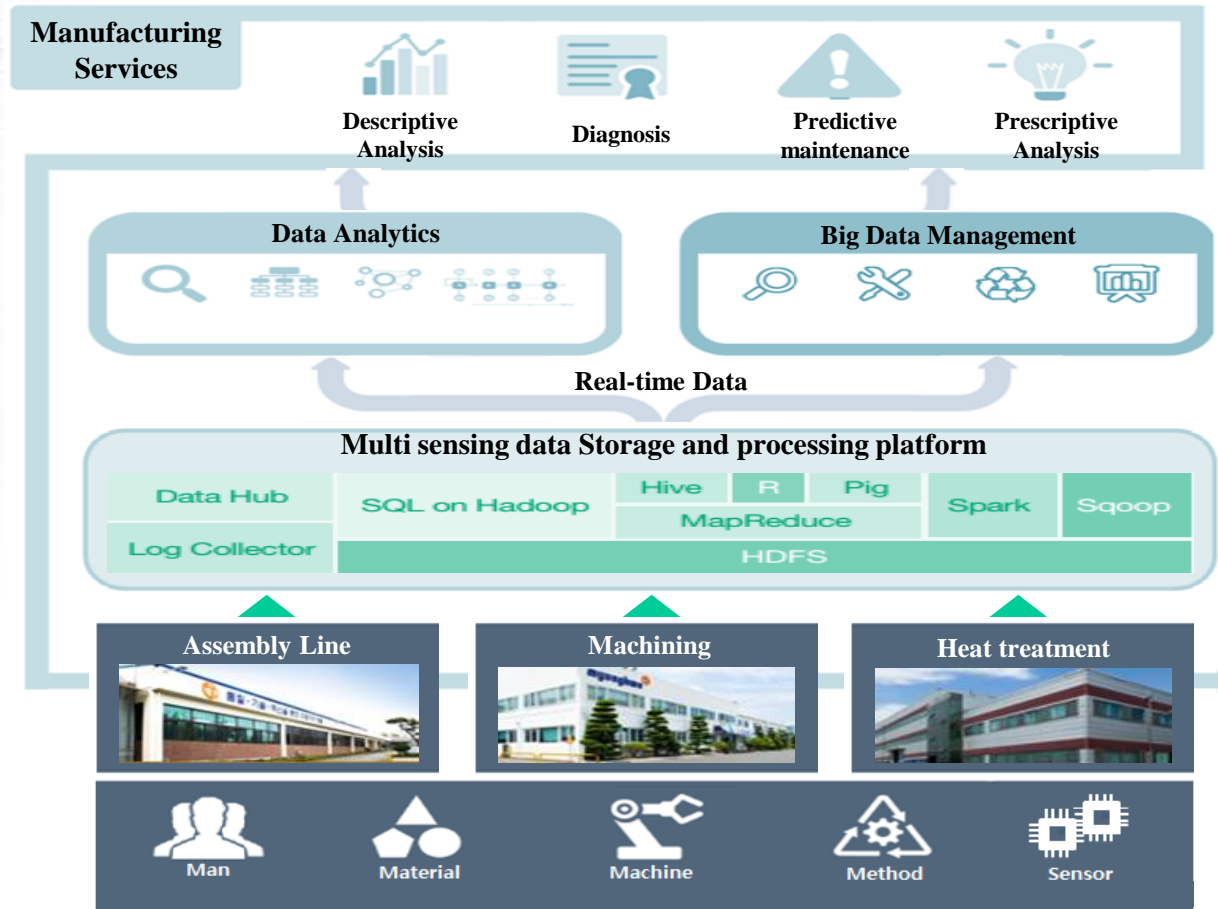
POSTECH



Introduction to Project

Goal

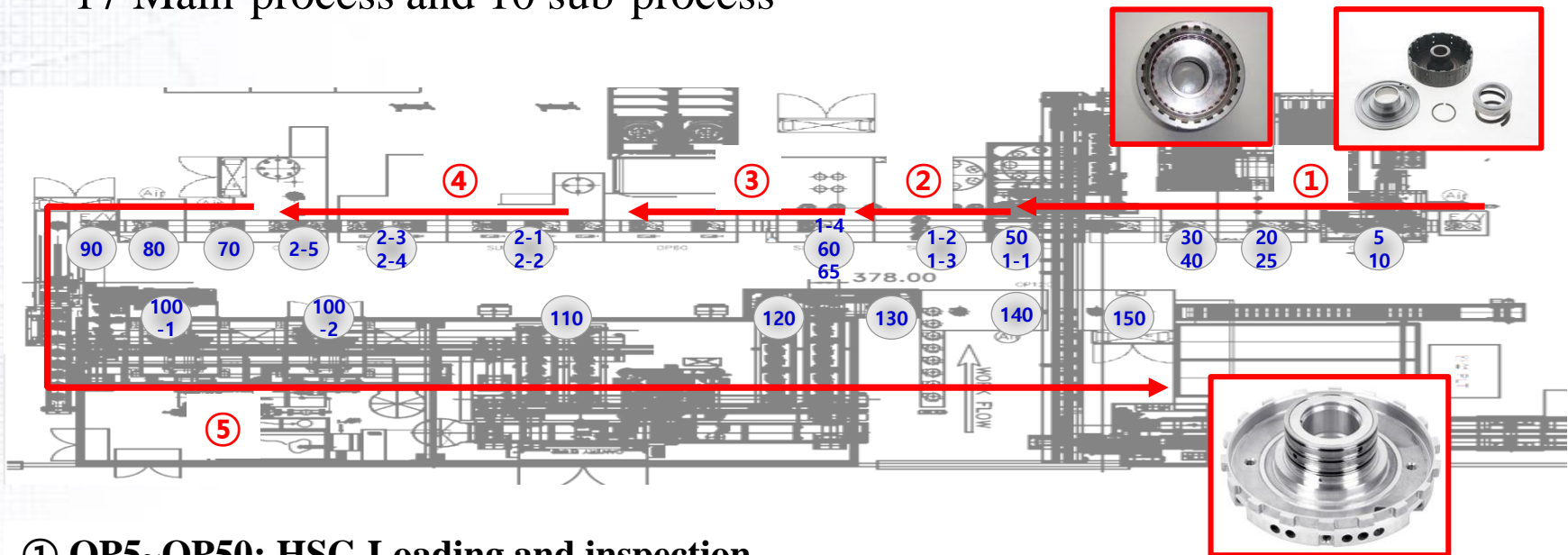
Development of artificial intelligence-based diagnosis / prediction system for the **Seamless Manufacturing**



Layout for Assembly Line

Automatic Assembly Line for Transmission Drum

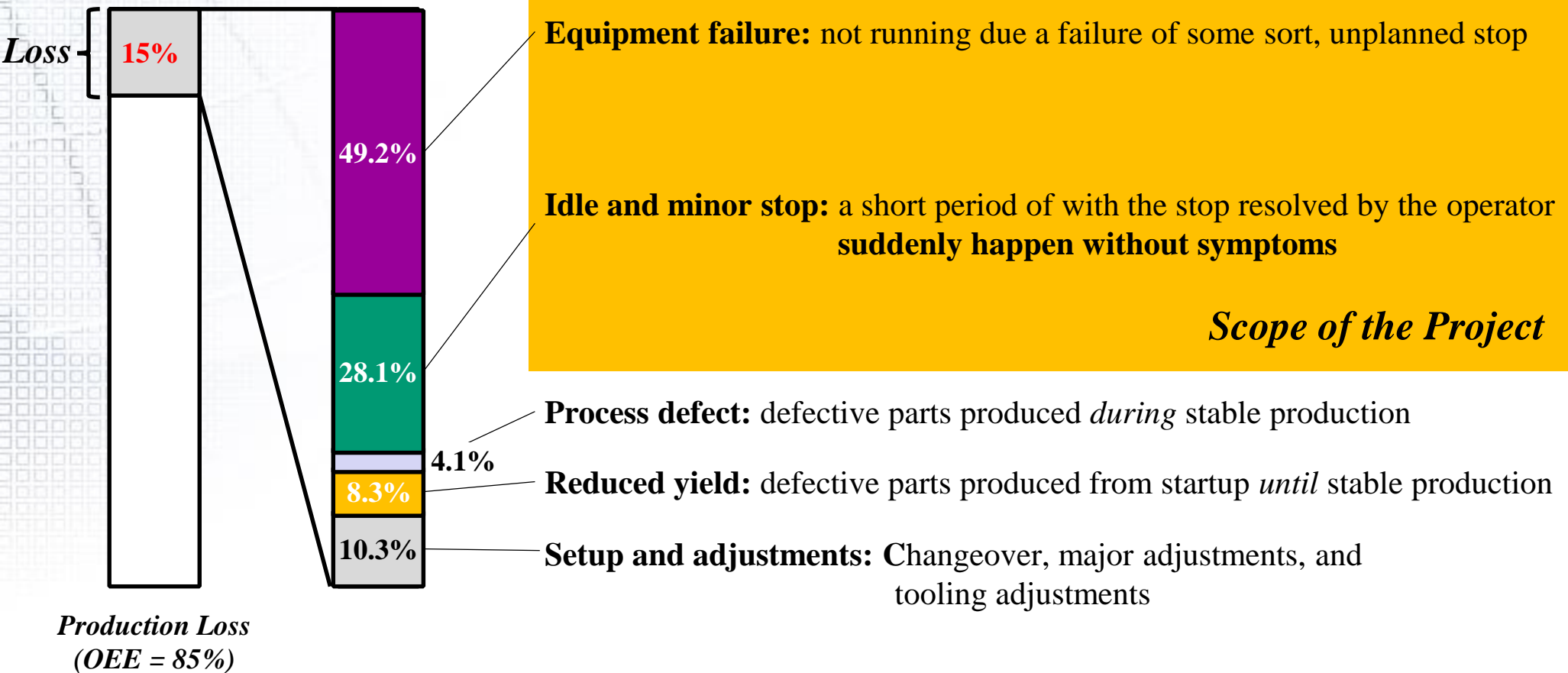
- 17 Main-process and 10 sub-process



- ① OP5~OP50: HSG Loading and inspection
- ② SUB1-1~SUB1-4: Piston, Cover, HSG Assembly
- ③ OP60~OP65: Fastening bolt and allowance inspection
- ④ SUB2-1~SUB2-5: SEAL Ring, O-RING assembly and operation check
- ⑤ OP70~OP150: Attaching barcode, Oil leak inspection, Oil remove, Package



Losses in Assembly Line





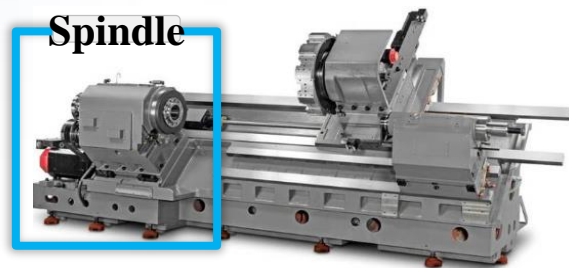
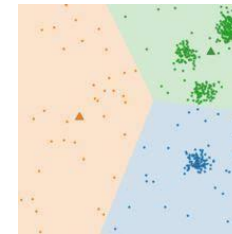
Efforts for Avoiding Equipment Failure

- ❑ Predictive maintenance for motors and machine tools



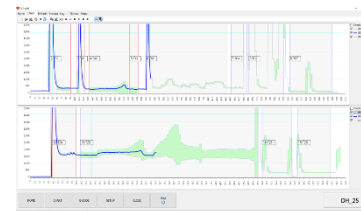
Servo motor

Vibration



Spindle

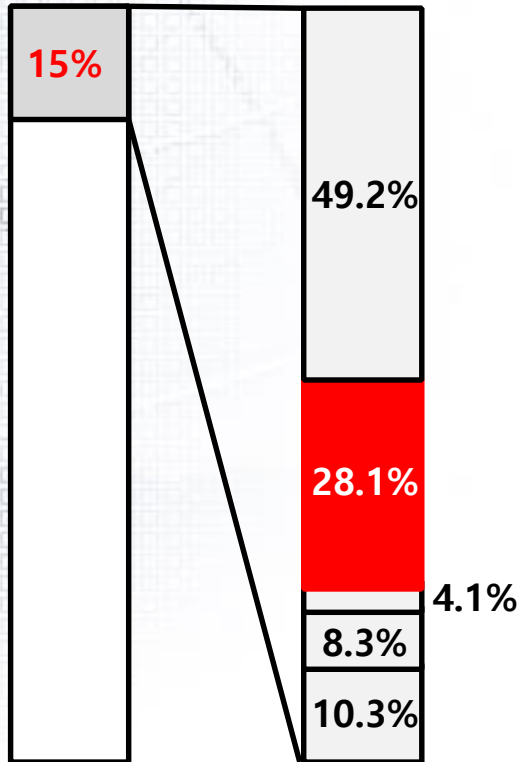
Vibration



Theses are not Today's Topic



Goal: Reduce Minor Stops



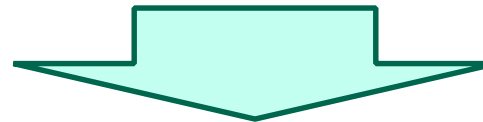
Definition: Unlike equipment failures, machine is stopped by the machine itself or operators for a time shorter than threshold due to transient errors

Cases of Minor Stops (difficult to measure or expensive)

- Screw jam
- Breakdown of proximity sensors
- Misaligned sensors
- Disconnection of harness

Problems on worksite

- (1) Stops without symptoms → Cannot predict when it will happen
- (2) Cannot detect them all automatically → Operators cannot recognize
- (3) Cannot find reasons without experts → Need time to fix



Data-driven Minor Stops detect and Reasoning



Data Exploration

- Summary of Data
 - 06/2017~07/2017 Data (> 800,000 rows)
 - 400 kinds of sensor data
 - 3 types of Minor stops on data

SerialNo	Cycle Time	Minor Stop	...	Torque	ScrewLoading	Air Pressure	Temperature	Proximity sensor_1	...
A0001	201	449		6.2	1	12.5	24.3	1	
A0001	201	449		6.2	1	12.5	24.3	1	
A0002	28.9	0		6.2	1	12.4	24.3	1	
A002	28.9	0		6.3	1	12.5	24.3	0	
...					

< Example of data >



Problems on Data

SerialNo	Cycle Time	Minor Stop	...	Torque	ScrewLoading	Air Pressure	Temperature	Proximity sensor_1	...
A0001	201	449		6.2	1	12.5	24.3	1	
A0001	201	449		6.2	1	12.5	24.3	1	
A0002	28.9	0		6.2	1	12.4	24.3	1	
A0002	28.9	0							
...									
A1210	177	0							
A1210	177	0		6.3				0	
A1210	177	0		6.2				1	
A1210	177	0		6.3	1	12.5	24.3	0	
A1211	28.9	0		6.2	1	12.4	24.3	1	
A1211	28.9	0		6.3	1	12.5	24.3	0	

- About 32% of data longer than Cycle time of the line (29s)
- Don't have error codes of minor stops

Low Reliability of the Data



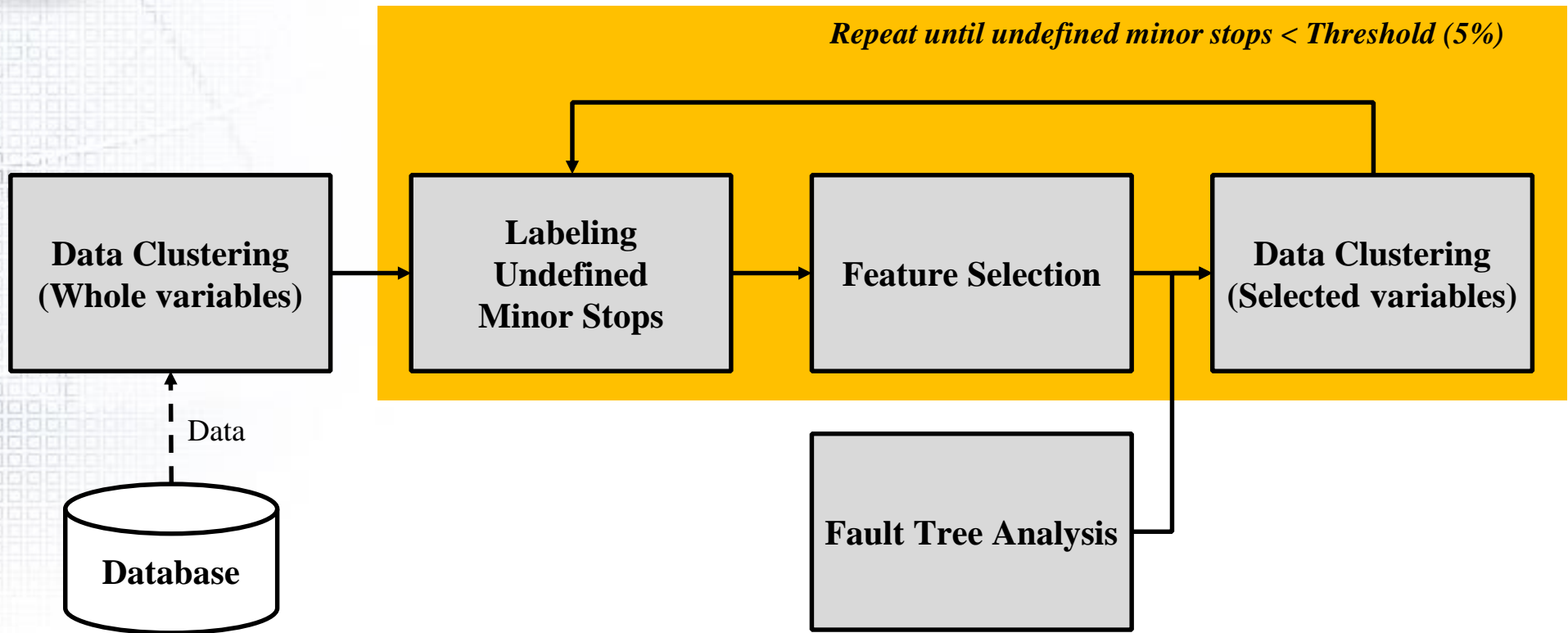
Define that processing time **more than 30s is Minor Stop**
Divide undefined minor stop data from normal condition data based on clustering



*Data-driven diagnosis model for **detecting** and **reasoning** minor stops*

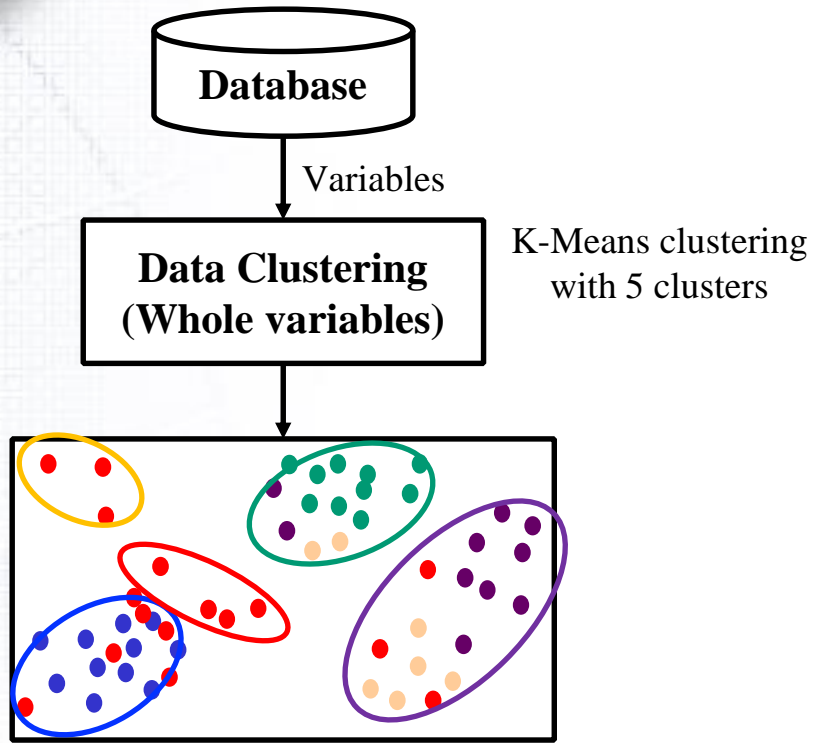


Modeling Approach

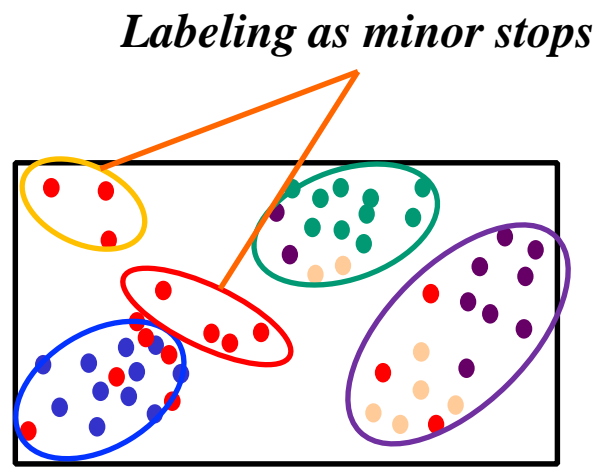




Data Clustering with Whole Data and Labeling undefined Minor Stops



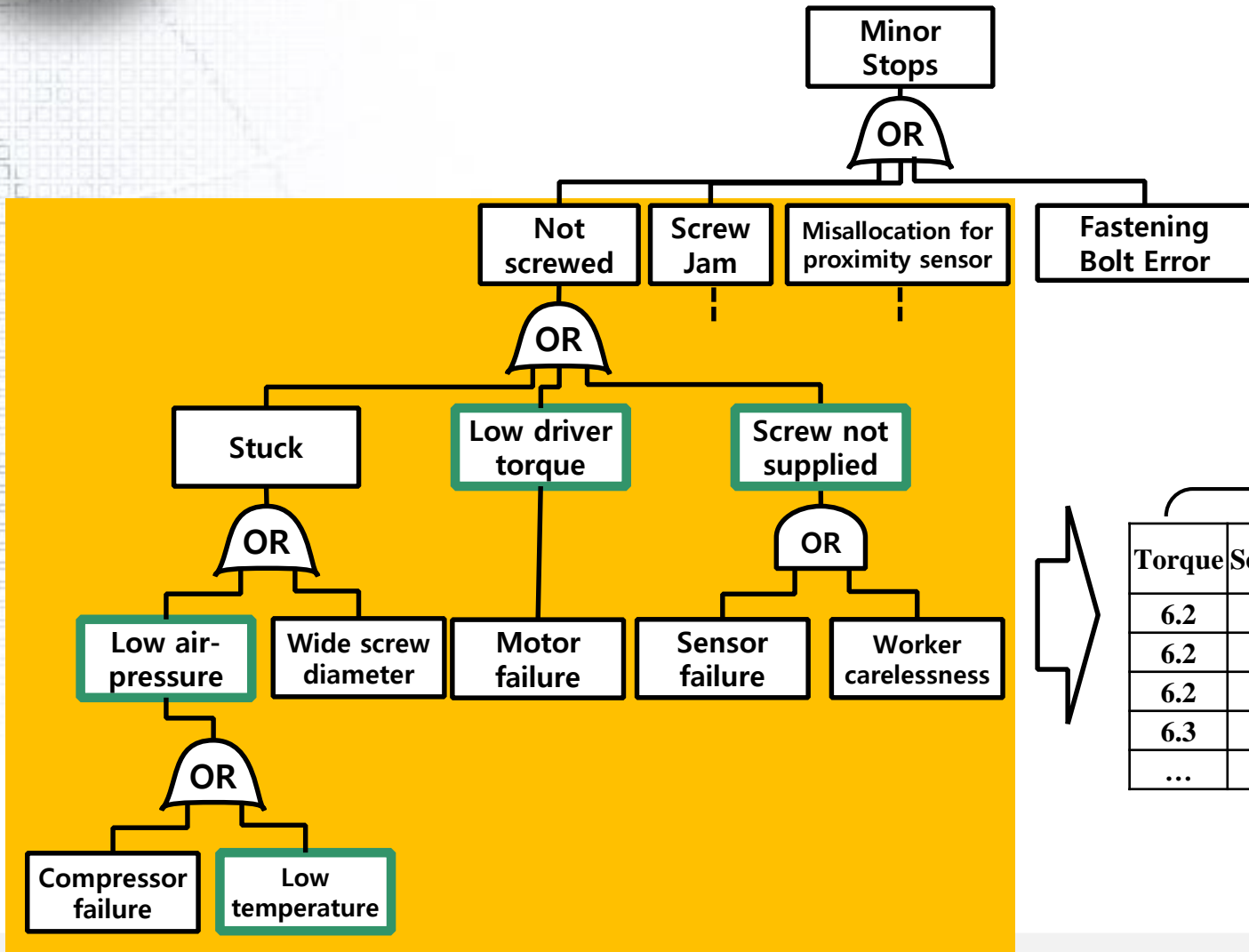
- Normal condition
- undefined minor stops
- Labeled minor stop #1
- Labeled minor stop #1
- Labeled minor stop #1





Fault Tree Analysis

Develop fault tree with worksite experts to find meaningful variables



Fixed variables



Torque	ScrewLoading	Air Pressure	Temperature	...
6.2	1	12.5	24.3	
6.2	1	12.5	24.3	
6.2	1	12.4	24.3	
6.3	1	12.5	24.3	
...	





Feature Extraction

- Variables from FTA are subjective
- Select meaningful and objective variables using ANOVA

Fault Code	X0B	X0F	X10	X11	X12	X13	X14	...
0	11.7	4.3	0	0	1	1	0	
0	11.6	4.4	0	0	1	1	0	
0	11.6	4.3	0	0	1	1	0	
0	11.7	4.4	0	0	1	1	0	
321	11.6	3.9	1	1	1	1	0	
331	11.7	3.9	1	1	1	1	0	
0	11.7	4.4	1	1	1	1	0	
0	11.6	4.3	1	1	1	1	0	
0	11.7	4.4	1	1	1	1	0	
181	11.6	4.3	1	0	1	1	0	
0	11.7	4.4	1	1	1	1	0	
0	11.6	4.3	1	1	1	1	0	
0	11.6	4.4	1	1	1	1	0	
0	11.6	4.3	1	1	1	1	0	
0	11.6	4.4	1	1	1	1	0	
...	



Variables	p-value
X0B	0.00
X0F	0.01
X10	0.00
X12	0.13
X13	0.35
X14	0.00
...	...

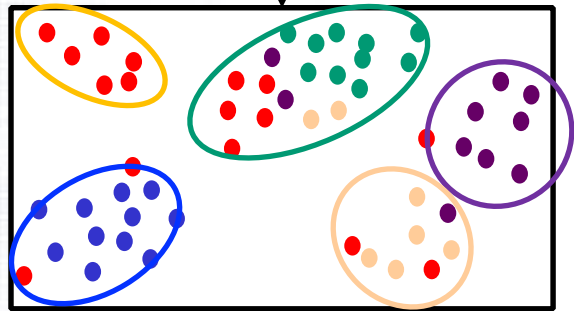
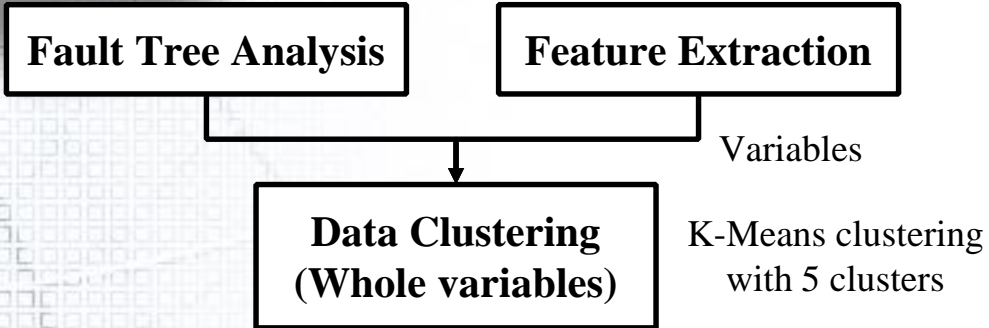


Variables	p-value
X0B	0.00
X0F	0.01
X10	0.00
X12	0.13
X13	0.35
X14	0.00
...	...



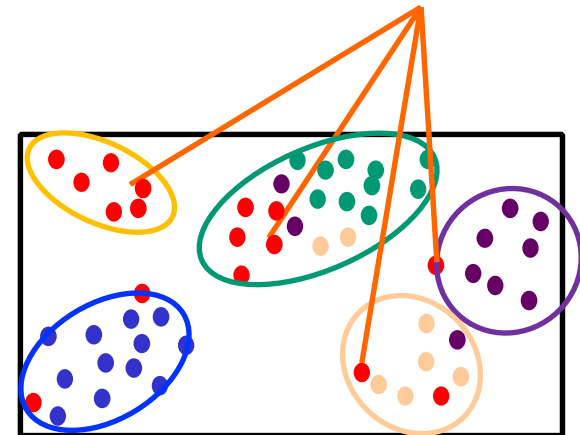


Data Clustering with Selected Variables and Labeling undefined Minor Stops



- Normal condition
- undefined minor stops
- Labeled minor stop #1
- Labeled minor stop #1
- Labeled minor stop #1

Labeling as minor stops



$$\frac{\text{Undefined minor stops}}{\text{Normal condition data} + \text{Undefined minor stops}} = 32\% \rightarrow 14.8\%$$



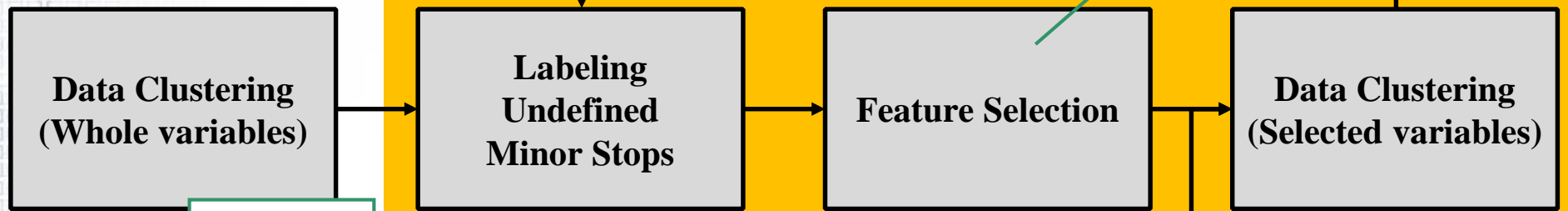


Result

$$\frac{\text{Undefined minor stops}}{\text{Normal condition data} + \text{Undefined minor stops}} = 32\% \rightarrow 14.8\% \rightarrow 5.5\% \rightarrow 4.1\%$$

Feature = Candidate for causing minor stops

Repeat until undefined minor stops < Threshold (5%)



400 variables

43 Fixed variables from worksite experts



Lessons Learned (1/2)

- ❑ Importance of data reliability
 - Developed model without considering cycle time → low recall and precision
 - Sensor data is not always accurate
 - Machines do not make a mistake but humans do
- ❑ Need more researches for reducing minor stops
 - Many researches for equipment failures but a few for minor stops
 - Minor stops happen more frequently than equipment failures
 - Generally repaired by operators and is not easy to identify and solve problems (Office staff couldn't identify minor stops for 2 hours during strike)



Lessons Learned (2/2)

- ❑ Convert tacit knowledge into explicit knowledge
 - Reducing feature space of data from Fault Tree Analysis
 - Knowledge should be part of the model
(Data talks. But can operators.)

- ❑ Increasing productivity is the key because of the regulation
 - Korean new labor regulation = no more than 52 hours work in a week
 - Cost ↑, Production capacity ↓
 - Data analytics will play an important role for productivity



Future Works

Data Analytics (5-tics) Roadmap

