

**Predict Plants Outage or Machine Failure** 

#### **Client Profile**

During a workshop at client site a use case came up on top of the list. How to reduce or even predict. Manufacturing plants outage hours or days in advance.

#### The Business Challenge

In plants there is a need of Predictive maintenance (PdM) that monitors the performance and condition of equipment while it is in use to reduce the risk of failure. Predictive maintenance, also known as condition-based maintenance, has been used in the industrial sector since the 1990s.

#### **Canspirit Predictive Analytics Approach**

Canspirit Team uses Google Cloud to implement the solution

- This model comprises of another Manufacturing unit's dataset, sources are telemetry, machine failure, error logs, maintenance, and machines configuration(age).
- Canspirit team will built a predictive model using this dataset, leveraging statistical techniques to identify the influence of certain errors, machine unique ID and machine components which are predicted to fail.

#### Telemetry OEM Automated Data Capture Software(s) e.g. ROCKWELL,REDZO Predictions on – Relative Influence on Machine outage Maintenance Component(s) causing failure Voltage, model and age leading failure Certain errors in gi en plant a NE, TRIHEDRAL, W Machine ONDERWARE Failures e.t.c Error Logs Machine Configuration superset 피

#### **Architecture of Predict Plants Outage or Machine Failure**

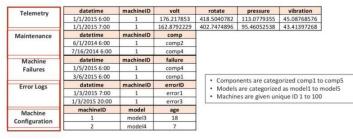
Nature of Data

Data Source

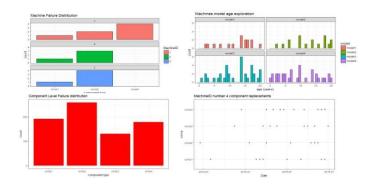
Predictive Model hosting on Cloud Computing

# A sample data of Predict Plants Outage or Machine Failure to decide whether the Model is effective:

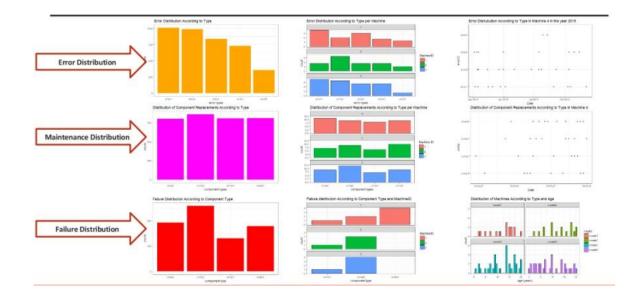
Machine Learning Model



#### **A Dataset View:**

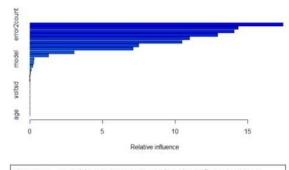


### **Exploration on Dataset**



## **Insights to Predictions :**





<u>Summary</u> – Variable error2count is predicted to influence most to cause a certain machine to fail. On the other hand vibrationsd\_24hrs and pressuremean are predicted not to cause any failure in next 24 hours (as per the dataset used).

	variable	Relative Influence
error2count	error2count	17.55478
error5count	error5count	14.92207
voltmean_24hrs	voltmean_24hrs	14.2271
vibrationmean_24hrs	vibrationmean 24hrs	12.93978
error3count	error3count	11.06276
pressuremean_24hrs	pressuremean_24hrs	9.595632
rotatemean_24hrs	rotatemean_24hrs	7.292141
error1count	error1count	6.60328
error4count	error4count	3.716549
sincelastcomp1	sincelastcomp1	1.364124
model	model	0.3032902
sincelastcomp3	sincelastcomp3	0.2124723
vibrationmean	vibrationmean	0.05653781
rotatemean	rotatemean	0.03527521
pressuresd_24hrs	pressuresd_24hrs	0.02871147
sincelastcomp4	sincelastcomp4	0.02720787
age	age	0.02229394
voltmean	voltmean	0.01600519
pressuresd	pressuresd	0.0131631
vibrationsd	vibrationsd	0.006823297
pressuremean	pressuremean	p
voltsd	voltad	þ
rotatesd	rotatesd	þ
voltsd_24hrs	voltsd_24hrs	p
rotatesd 24hrs	rotatesd 24hrs	þ
vibrationsd 24hrs	vibrationsd 24hrs	p
sincelastcomp2	sincelastcomp2	D

#### **Technologies and Tools**

- Python
- ROCKWELL
- REDZONE
- TRIHEDRAL
- WONDERWARE
- Google Cloud

#### **Challenges and Solutions**

- The machines and Components were categorized manually:
  - Components are categorized comp1 to comp5
  - Models are categorized as model1 to model5
  - Machines are given unique ID 1 to 100
- Data set and training of model was done using the company's requirement and Different OEM Automated Data Capture Software's were used.

#### **Business Benefits**

- The Company was able to predict manufacturing plants outage hours or days in advance.
- Prevention of Machine Failure reduced up to 40%.

#### **Bottom Line**

**Canspirit** leveraged its Data Science, machine learning and Cloud Computing skills to Predict Manufacturing Analytics to Predict Outage and Machine Failure.

For a discussion on how we can help you grow your business, email us now at: info@canspirit.ai