



# PSG INSTITUTE OF TECHNOLOGY AND APPLIED RESEARCH

Neelambur, Coimbatore – 641 062

## Smart Drainage, Effluent and Sewage Pumping System



**TEAM:**

PSG iTech – 2

**CATEGORY:**

Smart  
Machines

# Team Details



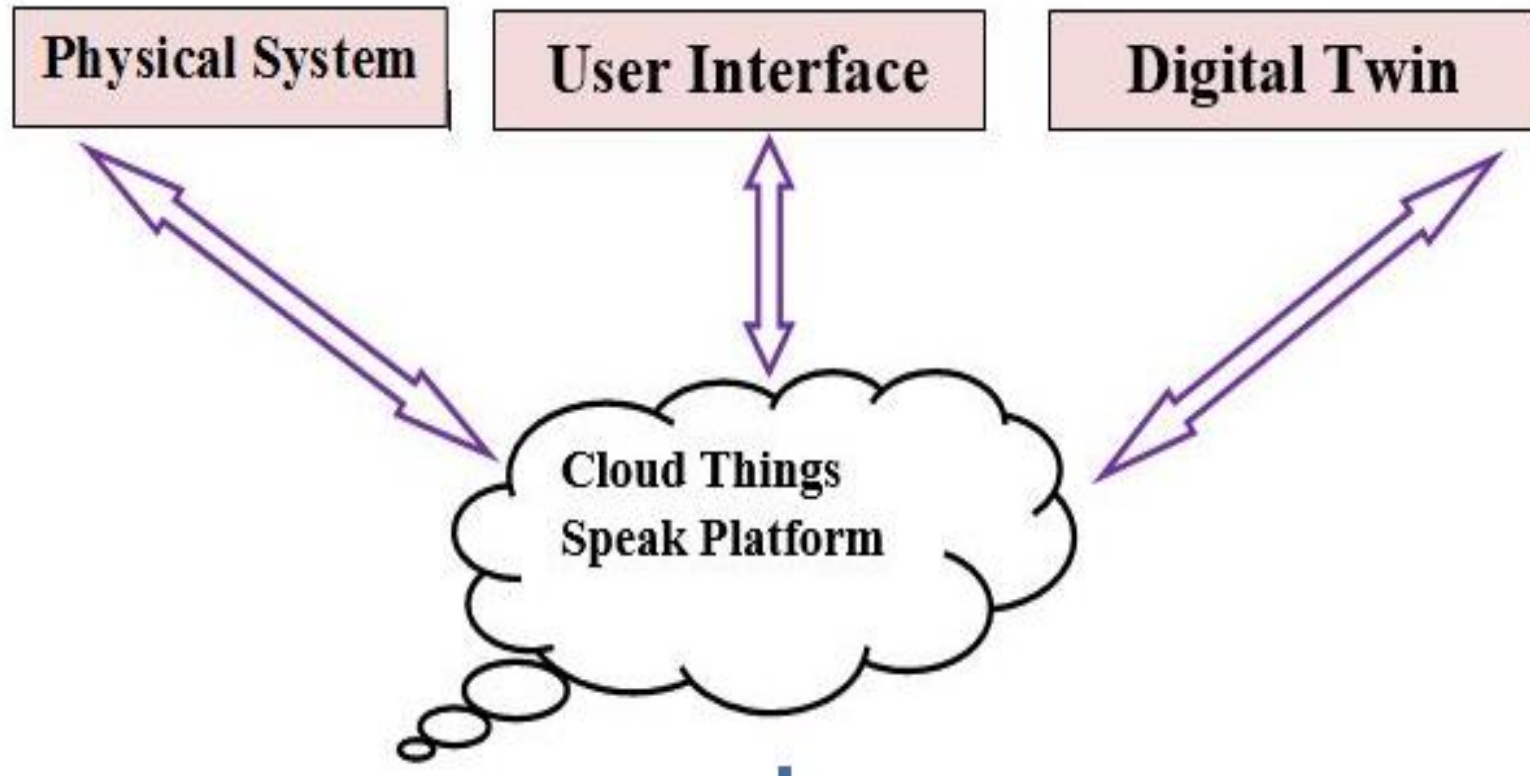
Participant Name	CT /DT Number	Role (Team Leader / Member)	Bachelors Discipline	Expected Year of Passing	Gender
Shravan S	CT20182402848	Team Leader	Electrical and Electronics Engineering	2020	M
Raagul A S	CT20182408578	Team Member	Mechanical Engineering	2020	M
Anirudh P S	CT20182401507	Team Member	Electronics and Communication Engineering	2020	M
Ramprakash V	CT20182408971	Team Member	Computer Science Engineering	2020	M



- Pumping of drainage, effluent and sewage is different from pumping of water due to the polluted nature of the materials.
- The pollutants are in the form of suspended solids and floating solids. Hence, the **load on the pump varies widely.**
- The proposed system will monitor single phasing fault, stator temperature, vibration, and ambient temperature, and will **dynamically control the speed of the pump and ensure sustained reliable performance of the pump.**



- Intelligence, by way of machine learning, is built into the system using **mathematical expressions that characterize the physical system, and data analytics.**
- The users can **query the digital twin** to know the historical performance, and current operating conditions.
- The proposed system can **trigger alarms as early warnings, and also make predictions** about possible system anomalies, if and when they occur

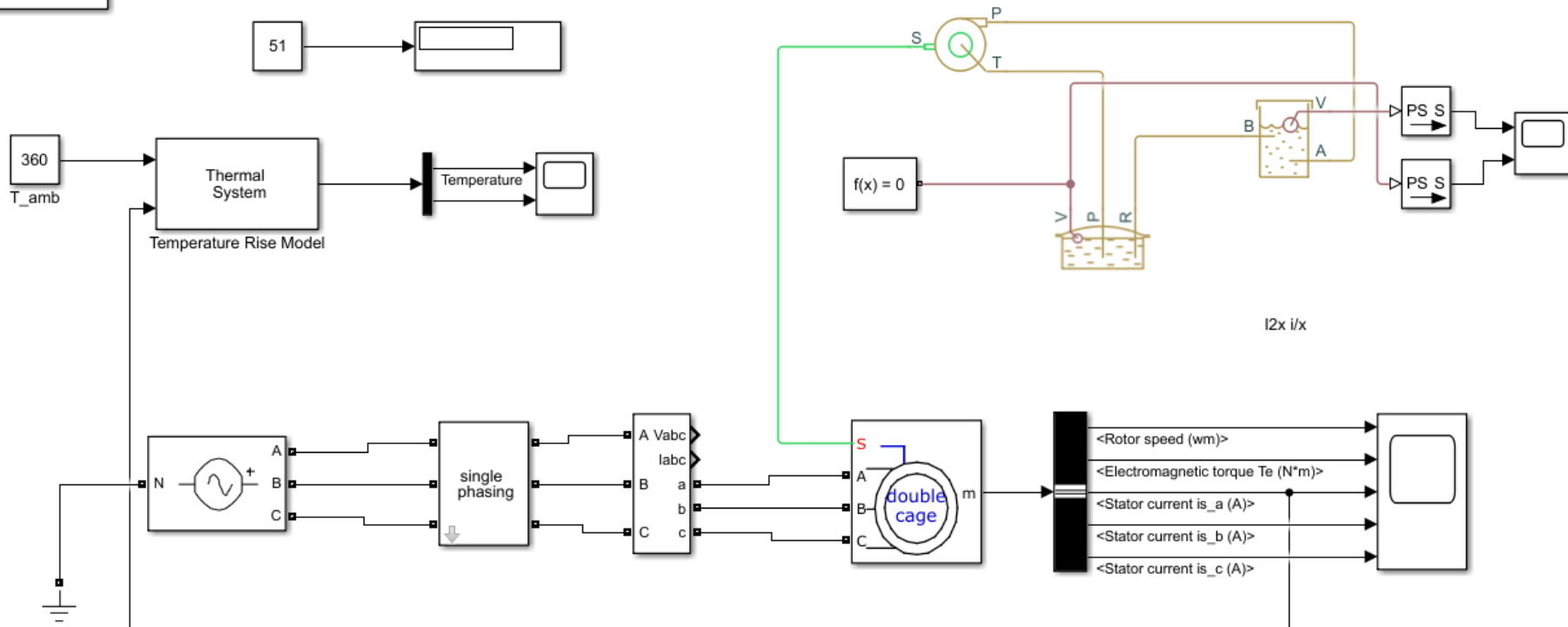




# Digital Twin



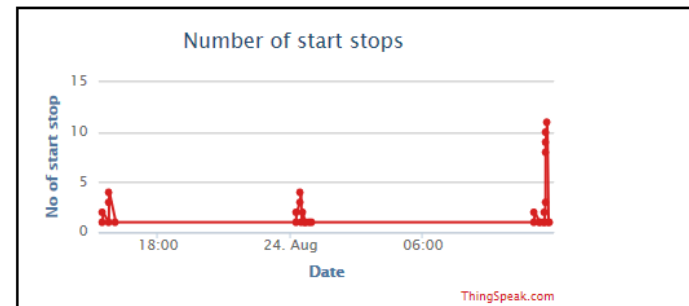
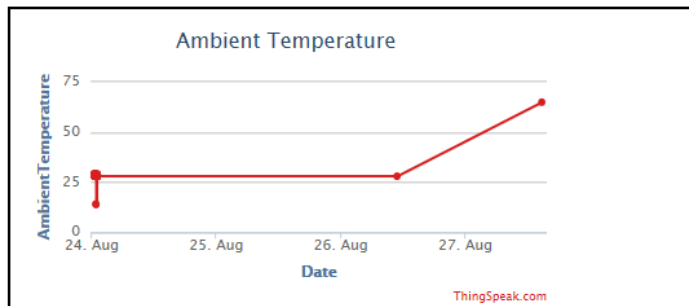
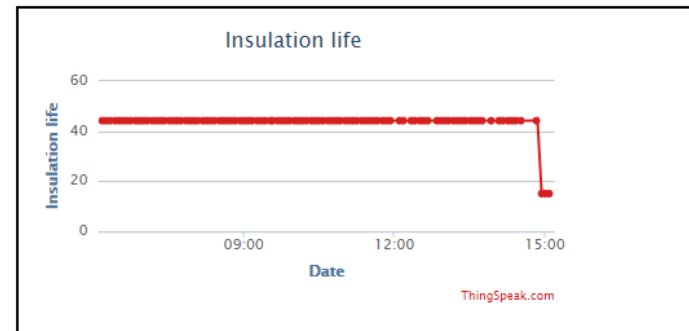
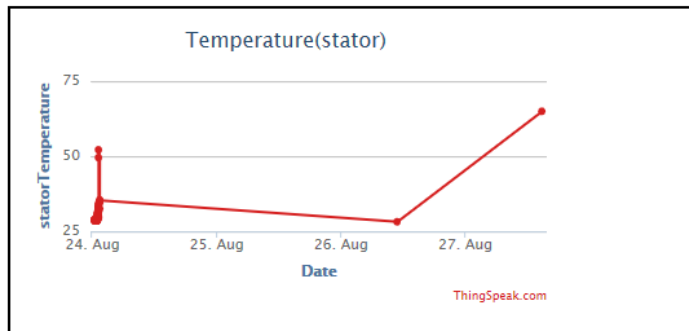
Continuous







## SMART DRAINAGE EFFLUENT AND SEWAGE PUMPING SYSTEM





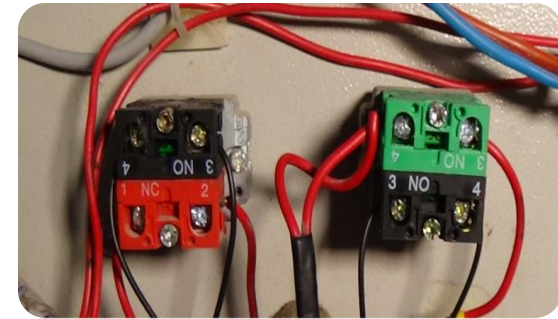


# SYSTEM OPERATION

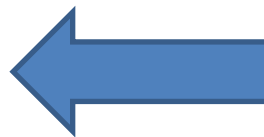
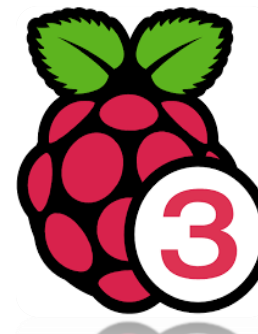
# Starting The Digital Twin



Start push button is pressed



Auxiliary contactors gets actuated



# Starting The Digital Twin

```
pi@raspberrypi: ~/enginx/adx1345-python
File Edit View Help
pi@raspberrypi: ~/enginx/adx1345-python $ python start2.py
start2.py:16: SyntaxWarning: name 'flag' is assigned to before global declaration
    global flag
start2.py:24: SyntaxWarning: name 'flag' is assigned to before global declaration
    global flag
start2.py:26: SyntaxWarning: name 'counts' is assigned to before global declaration
    global counts
start
```

```
Command Window
>> cloud
Warning: Cannot query I2C bus speed.
> In raspi/getAvailablePeripherals (line 929)
   In raspi (line 247)
   In cloud (line 6)
motor started0.1417180.1507130.1505130.141550
3.767813
```

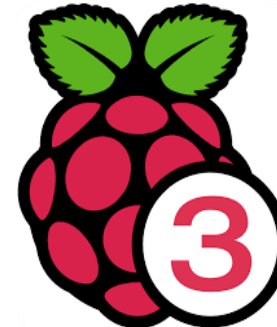
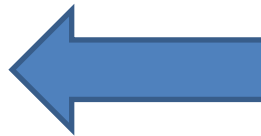
# Stopping The Digital Twin



Stop push button is pressed



Auxiliary contactors get actuated



# Stopping The Digital Twin

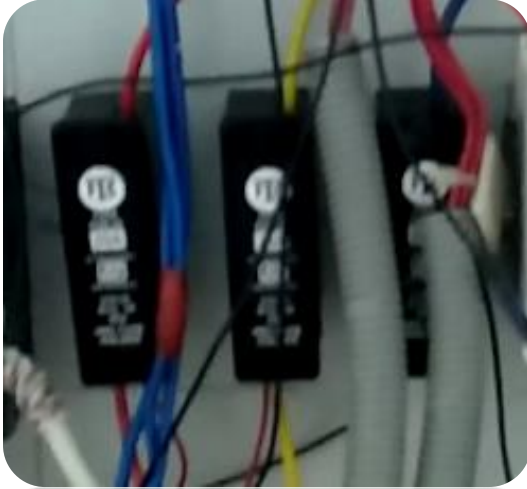
```
raspberrypi: /engine/bin/100 python start2.py
start2.py:16: SyntaxWarning: name 'flag' is assigned to before global declaration
    global flag
start2.py:24: SyntaxWarning: name 'flag' is assigned to before global declaration
    global flag
start2.py:26: SyntaxWarning: name 'counts' is assigned to before global declaration
    global counts
start2.py:8: RuntimeWarning: This channel is already in use, continuing anyway.
Use GPIO.setwarnings(False) to disable warnings.
    GPIO.setup(17,GPIO.OUT)
start
stop
1
start
stop
2
```

Command Window

```
>>
>>
>>
>>
/x >> Motor stopped
```

# SINGLE PHASING FAULT

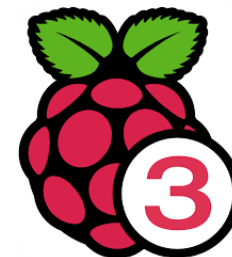
# Single Phasing Fault



Electrical Fuse

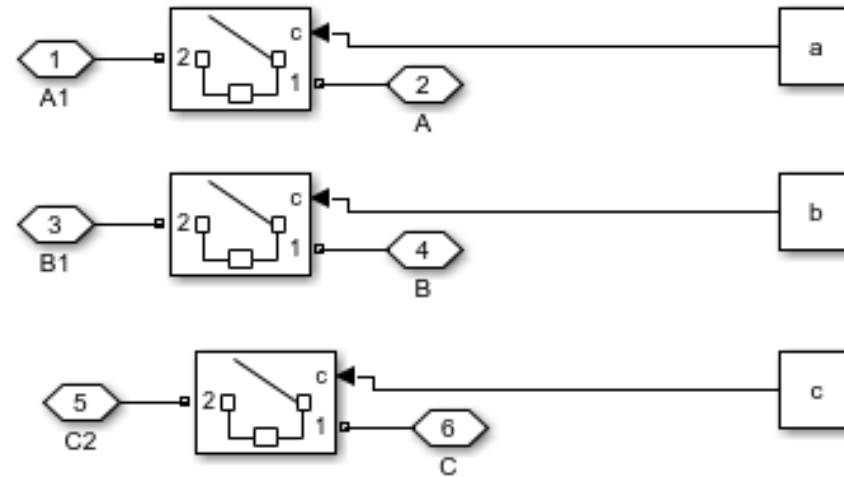


Energy Meter

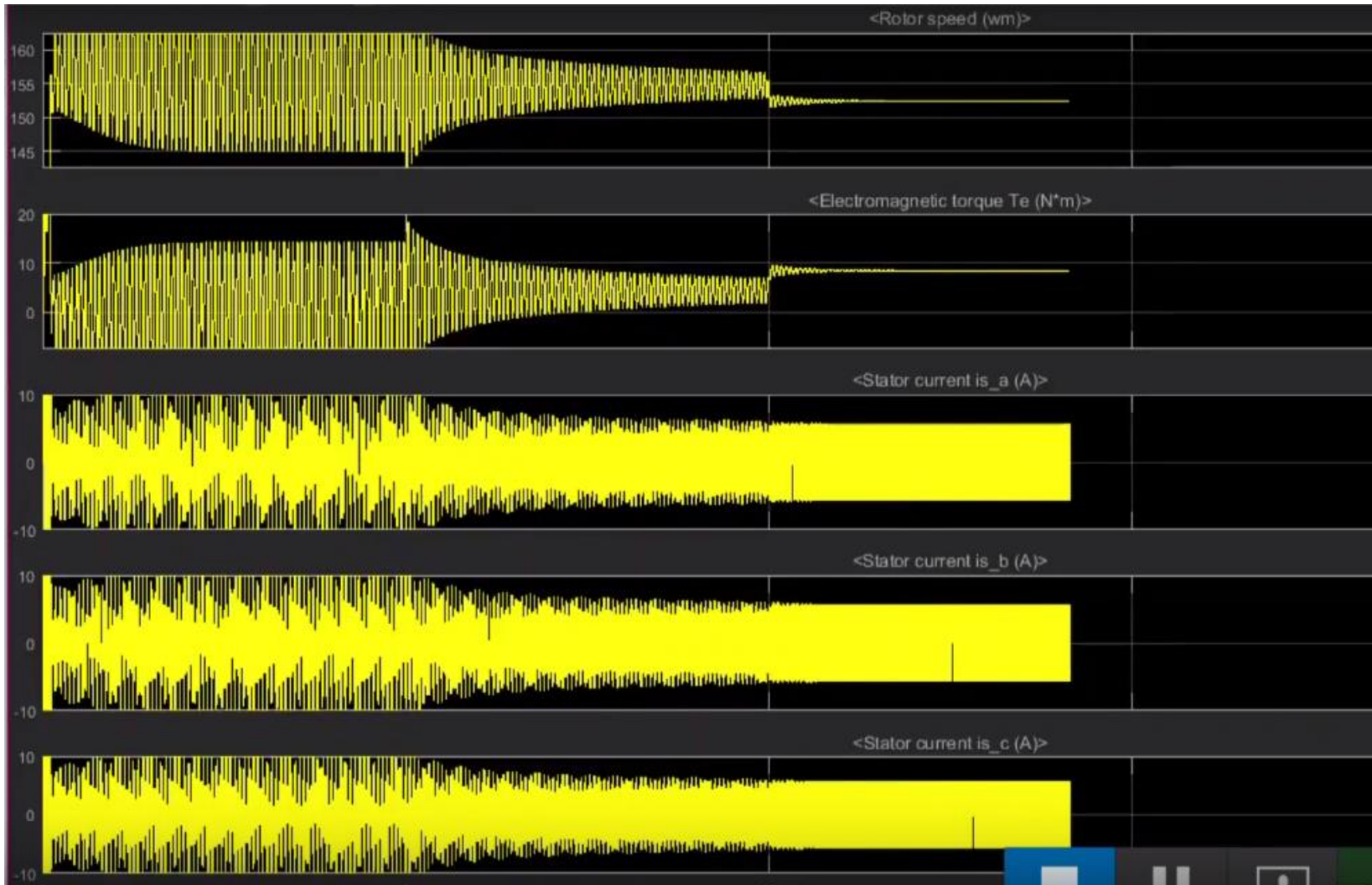




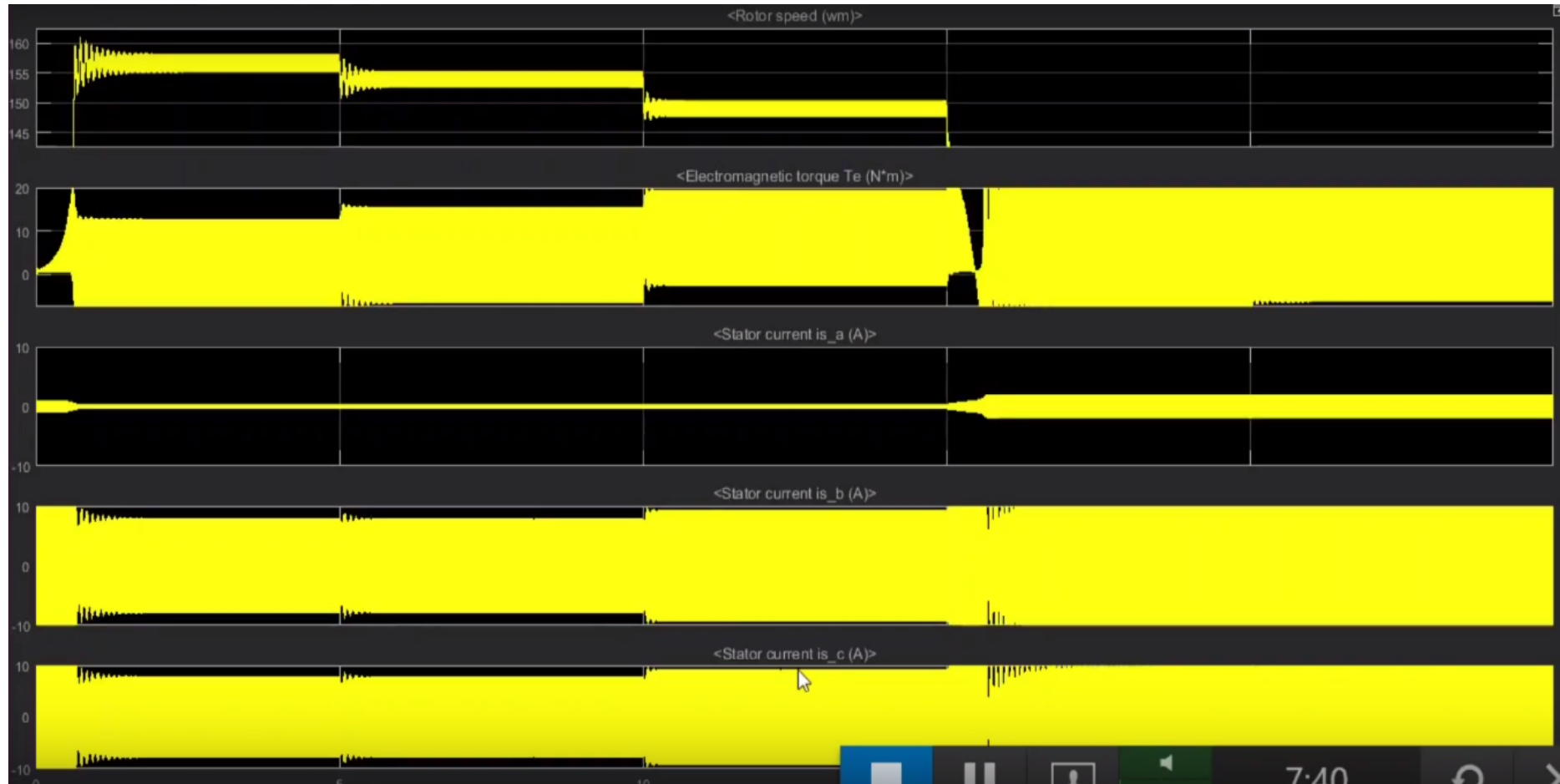
# Single Phasing Fault-Twin



# Electrical Parameters Waveform



# Single Phasing Fault-Waveform

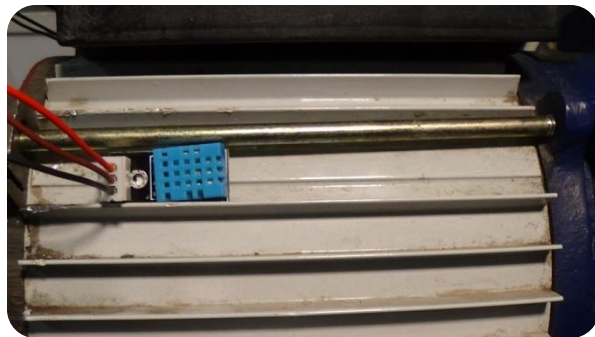


# THERMAL SUBSYSTEM

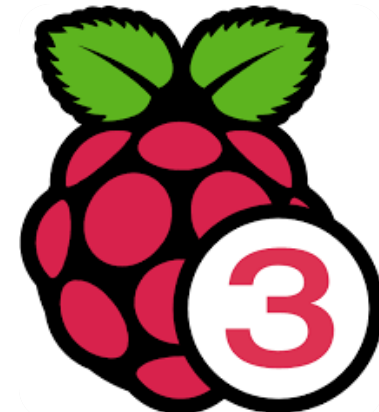
# Insulation Life Estimation



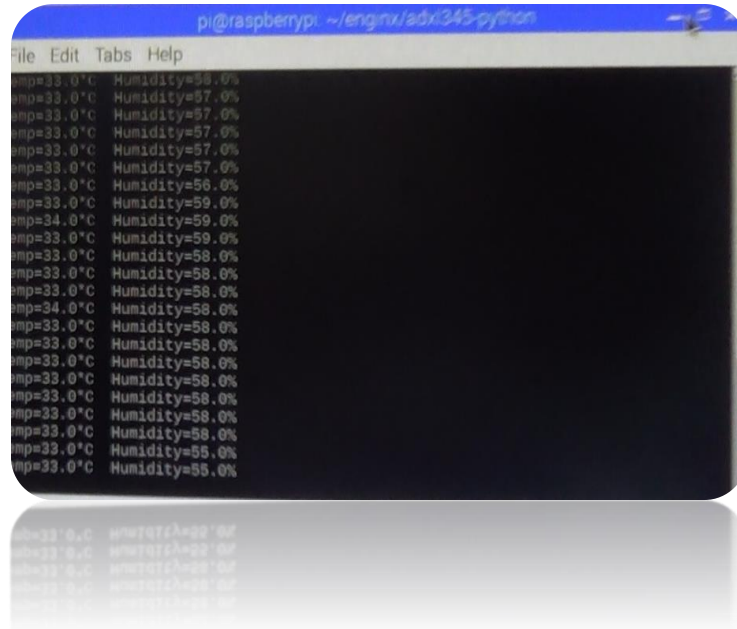
Load current data from  
Energy Meter



Ambient Temperature  
from DHT 11



# Insulation Life Estimation



```
pi@raspberrypi: ~/enginx/adx345-python
File Edit Tabs Help
temp=33.0°C Humidity=58.0%
temp=33.0°C Humidity=57.0%
temp=33.0°C Humidity=57.0%
temp=33.0°C Humidity=57.0%
temp=33.0°C Humidity=57.0%
temp=33.0°C Humidity=57.0%
temp=33.0°C Humidity=56.0%
temp=33.0°C Humidity=59.0%
temp=34.0°C Humidity=59.0%
temp=33.0°C Humidity=59.0%
temp=33.0°C Humidity=58.0%
temp=33.0°C Humidity=58.0%
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temp=33.0°C Humidity=58.0%
temp=33.0°C Humidity=58.0%
temp=33.0°C Humidity=55.0%
temp=33.0°C Humidity=55.0%
```



Ambient Temperature readings in Raspberry Pi Screen

Ambient Temperature Readings Being Plotted In ThingSpeak

# Insulation Life Estimation



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ThingSpeak Channels Apps Blog Support Sign In Sign Up

## Apps

ThingSpeak channels store data. Upload data from the web or send data from devices to a ThingSpeak channel. Use these apps to transform and visualize data or trigger an action. See [Tutorial: ThingSpeak and MATLAB](#) to create a channel.

### Analytics

**MATLAB Analysis**  
Explore and transform data.

**MATLAB Visualizations**  
Visualize data in MATLAB plots.

**Plugins**  
Display data in gauges, charts, or custom plugins.

Ambient Temperature Readings Being Plotted In ThingSpeak



Load Current Data From MATLAB

Insulation Lifetime Estimation Using MATLAB Analytics in ThingSpeak platform





## Algorithms Behind Insulation Life Estimation:

- Montsinger Rule
- Arrhenius Equation
- Final consolidated equations:

1. Hotspot Temperature =[Loss Factor\* Allowable temperature rise]+Ambient Temperature

$$2. L_x = L_{100} * 2 \exp \left[ \frac{T_c - T_x}{HIC} \right]$$



## Code Snippets:

```
1 clc;
2 % Parameter Declarations
3 Class=[105 130 155 180];
4 HIC=[14 11 9.3 8];
5 Load_Factor=50:1:150;
6 [r c]=size(Load_Factor);
7 Loss_Factor=zeros(1,c);
8 for i=1:1:c
9     Loss_Factor(i)=0.012*Load_Factor(i)*Load_Factor(i)-(0.83*Load_Factor(i))+65;
10 end
11 Loss_Factor;
12 %%
13 Channel_ID=559226;
14 Channel_ID2=562478;
15 [Torque,timestamp]=thingSpeakRead(Channel_ID,'Fields',1,'NumPoints',10)
16 [rr cc]=size(Torque);
17 Torque_Tot=0;
18 count=0;
19 for i=1:1:rr
20     temp=Torque(i);
21     if isnan(temp)==0
22         Torque_Tot=Torque_Tot+temp;
23         count=count+1;
24     end
25 end
26 Torque_Tot=Torque_Tot;
27 Temp_amb=thingSpeakRead(Channel_ID2,'ReadKey','PBLRW9SSRWYQI66Z','Fields',1);
28 Time_Period=20; % Current Load Duty time
```

```
28 Time_Period=20; % Current Load Duty time
29 Base_Load_Time=24;
30 LV_PU = Time_Period/Base_Load_Time;
31 Load_Tot= ceil(sum(Torque_Tot)/(LV_PU+count-1))
32 % Getting Class Insulation
33 %Ins_Class_Cell=thingSpeakRead(Channel_ID,'Fields',3,'OutputFormat','table');
34 %Ins_Class=Ins_Class_Cell.InsulationClass{1};
35 Ins_Class='B';
36
37 if Ins_Class=='A'
38     c=Class(1);
39     h=HIC(1);
40 end
41 if Ins_Class=='B'
42     c=Class(2);
43     h=HIC(2);
44 end
45 if Ins_Class=='F'
46     c=Class(3);
47     h=HIC(3);
48 end
49 if Ins_Class=='H'
50     c=Class(4);
51     h=HIC(4);
52 end
53 Inter_Load=Load_Factor-Load_Tot;
54 z=find(Inter_Load==0)
55 F=Loss_Factor(z)/100;
56 fprintf('Loss Factor F: %f\n',F);
57 T_hot=F*(c-Temp_amb)+Temp_amb;
58 fprintf('Hot spot Tx: %f\n',T_hot);
59 Lx=100*(2^((c-T_hot)/h));
60 E_year=((Lx*20000)/(100*24*365));
```

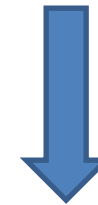
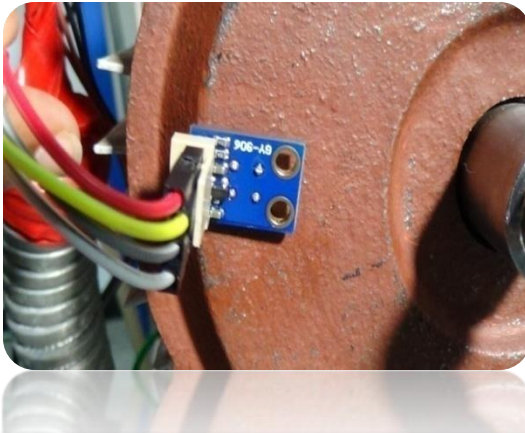
```
61 E_hours=E_year*24*365;
62 Elapsed=(LV_PU+count-1)*24;
63 E_year_rem=(E_hours-Elapsed)/(365*24);
64 fprintf('Expected life in year: %f \nDays: %f \nHours: %f\n',E_year_rem,E_year_rem*365,E_y
65 thingSpeakWrite(559243,E_year,'Fields',1,'WriteKey','SRHS7G3QM6EE7A0H');
66 if E_year*365 <100
67     fprintf('Please Check for Insulation Replacement \n');
68     thingSpeakWrite(563794,1,'WriteKey','T07N0YH6P6LL9WDB');
69 else
70     thingSpeakWrite(563794,0,'WriteKey','T07N0YH6P6LL9WDB');
71 end
```



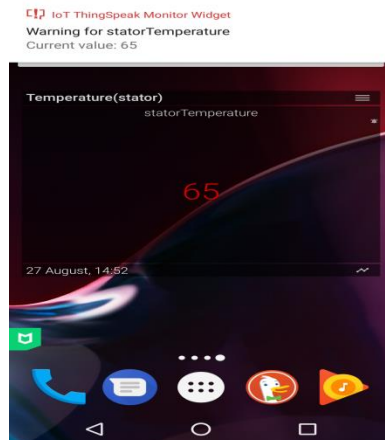
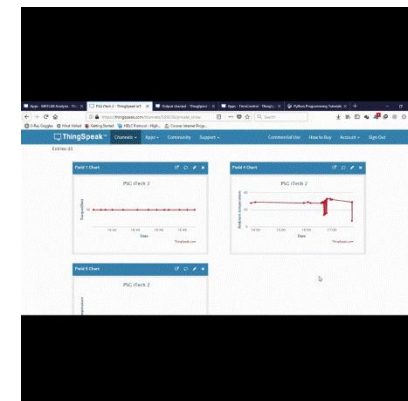
```
Loss Factor F: 0.542880  
Hot spot Tx: 88.859200  
Expected life in year: 30.493791  
Days: 11130.233571  
Hours: 267125.605715
```

Output Window Displaying The Estimated Insulation Life

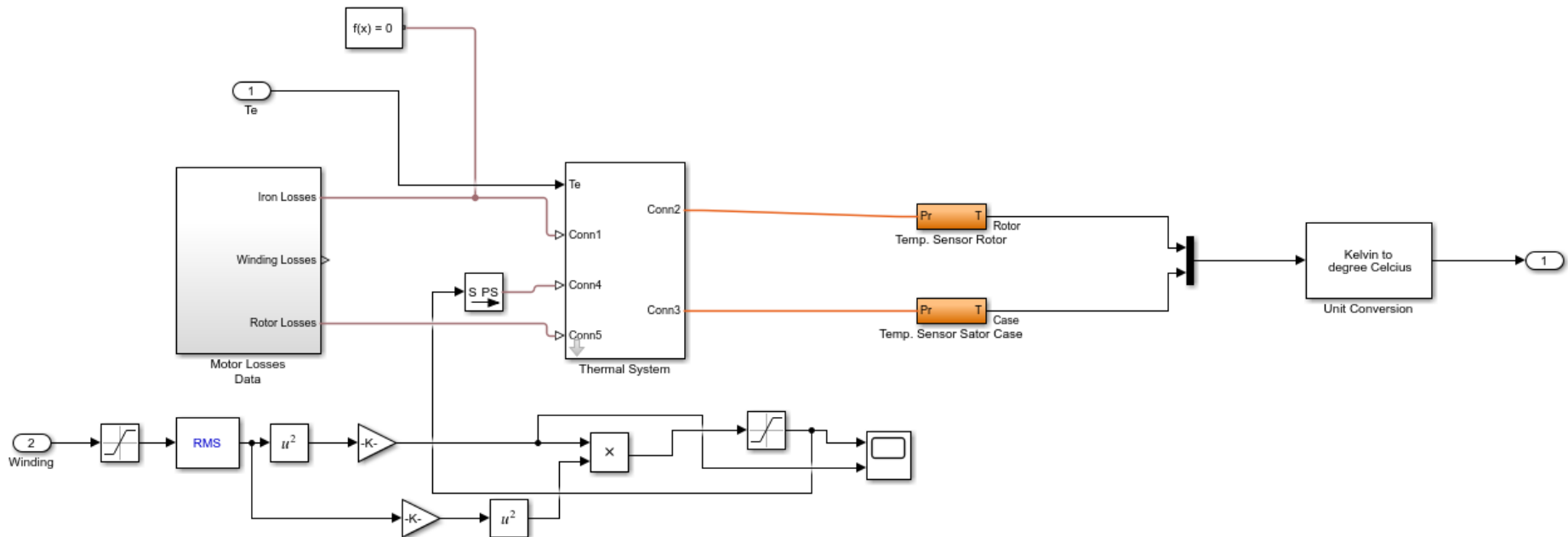
# Monitoring Temperature Rise In Stator Winding



Stator Winding Temperature  
being detected by MLX90614



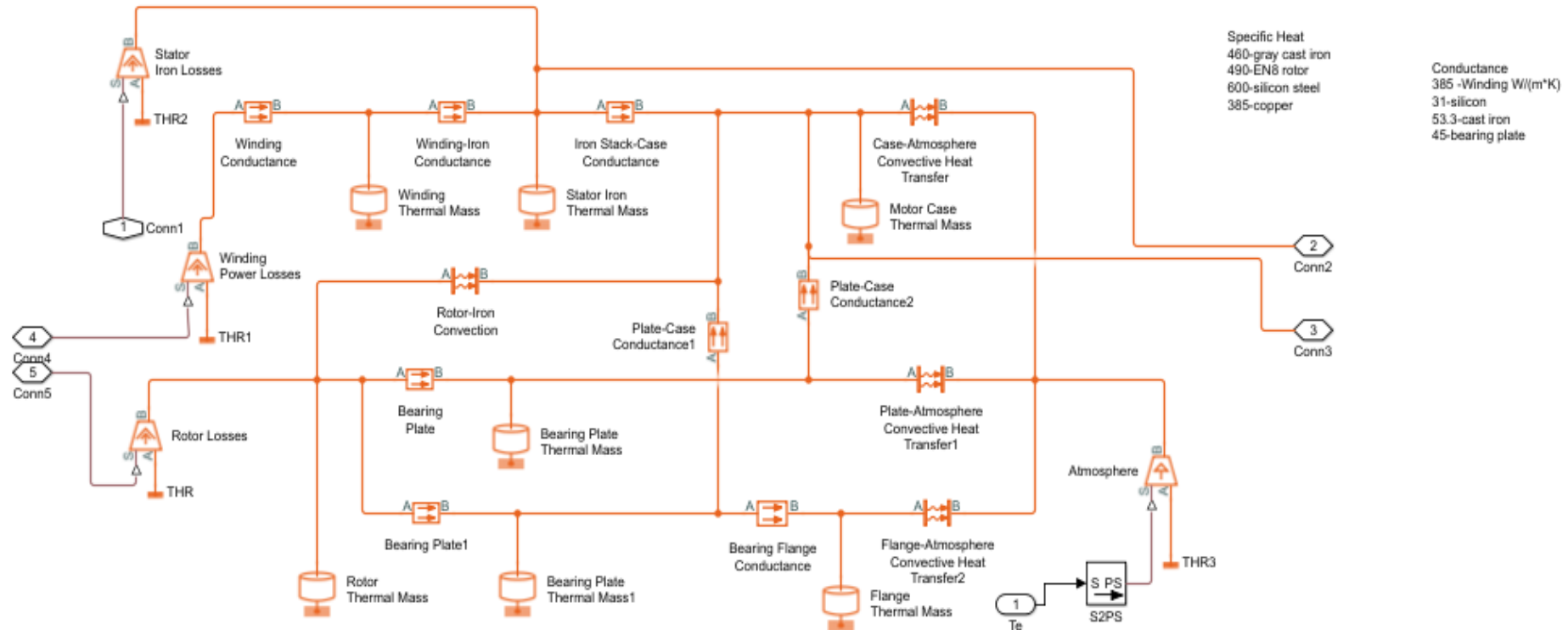
# Thermal Sub System-Digital Twin



# Heat Transfer-Digital Twin



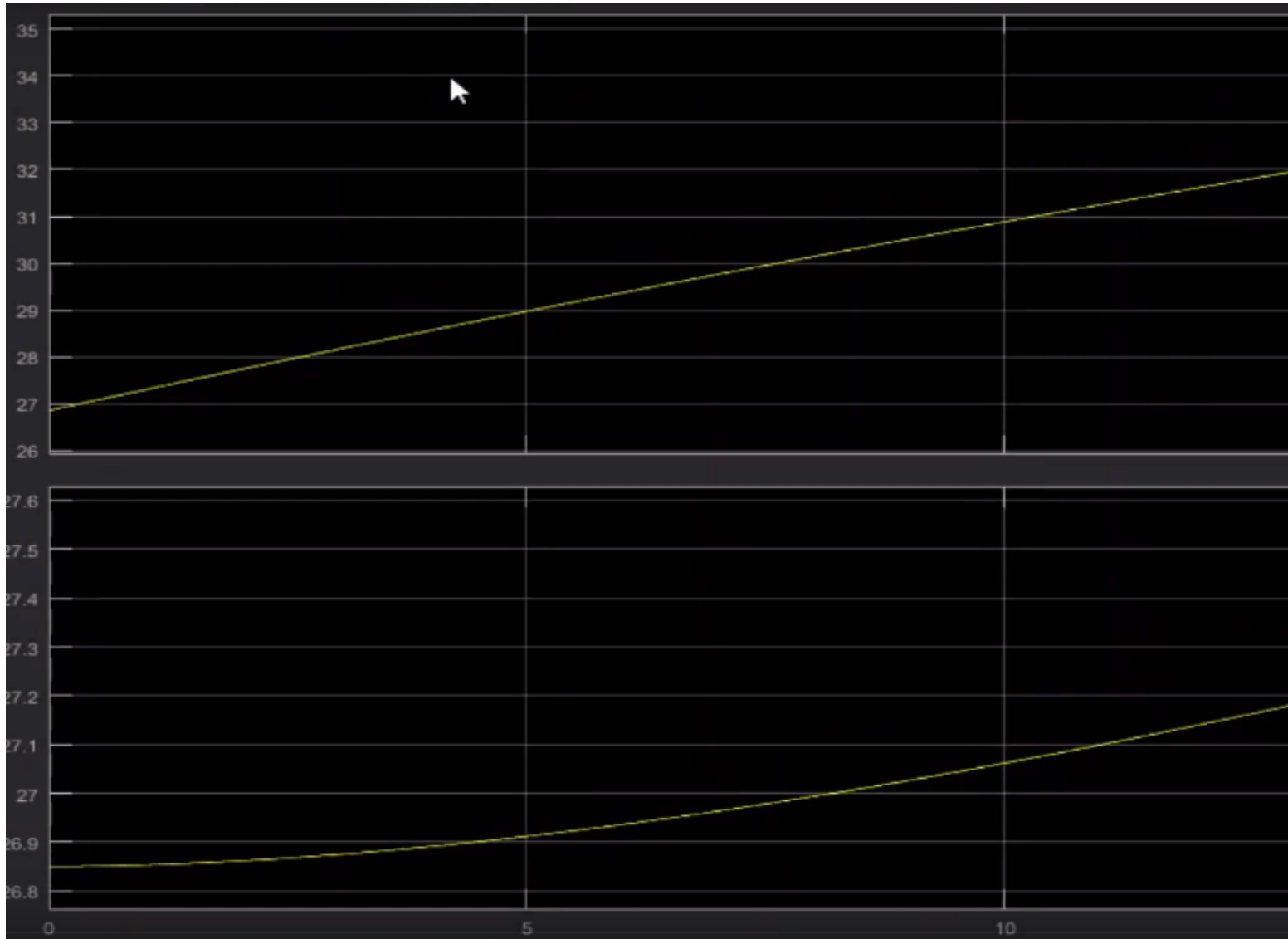
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# Temperature Rise Waveform



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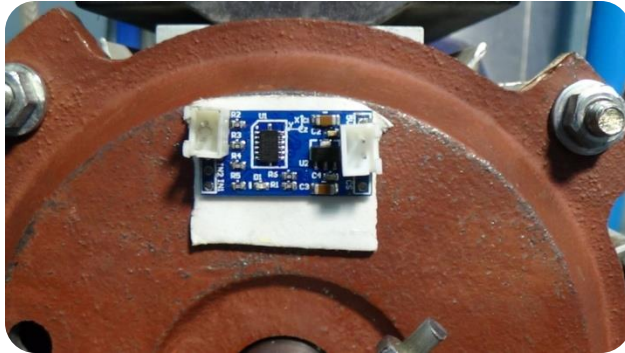




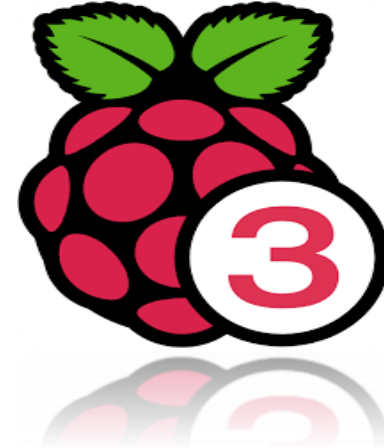


# VIBRATION ANALYSIS

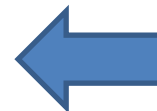
# Vibration Analysis

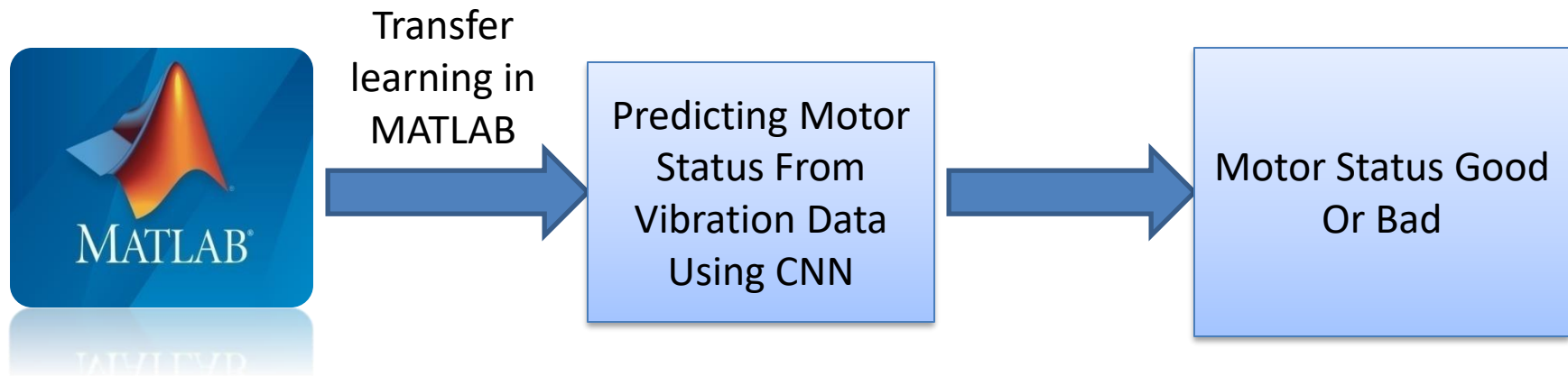


ADXL345  
mounted on the  
casing

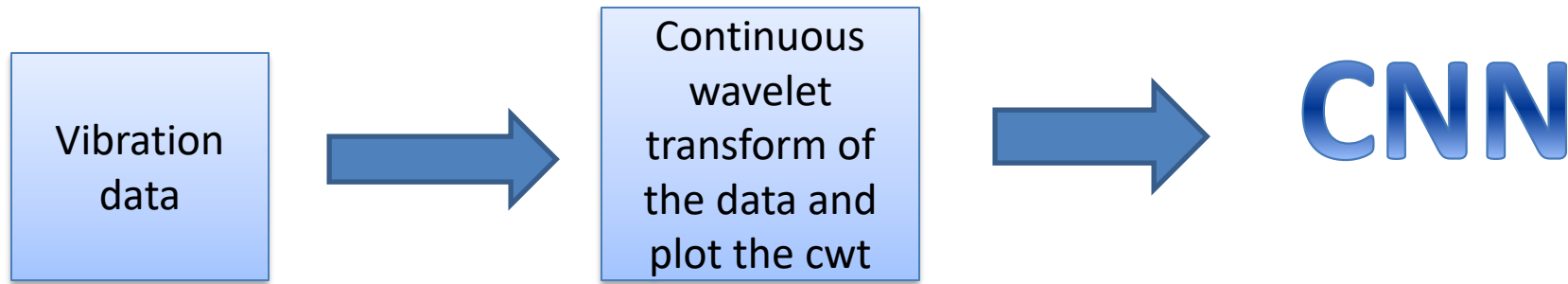


```
pi@raspberrypi: ~/enginx/adxl345-python
y = -1.0486
z = -0.2086
ADXL345 on address 0x53:
x = 2.0446
y = -1.7086
z = 2.0446
mount the sensor properly
ADXL345 on address 0x53:
x = 1.7926
y = -1.1286
z = 2.0446
ADXL345 on address 0x53:
x = 1.2806
y = 0.7126
z = 2.0446
ADXL345 on address 0x53:
x = 2.0446
y = -2.0486
z = 1.9446
ADXL345 on address 0x53:
x = 2.0446
y = -2.0486
z = 1.3706
```





# VIBRATION ANALYSIS:



# VIBRATION ANALYSIS:

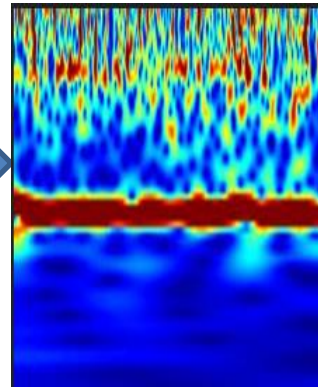


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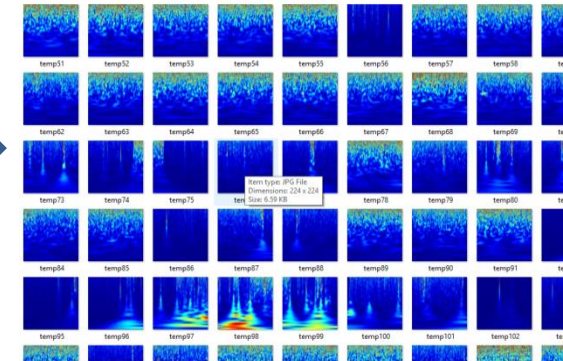


IMPORT		VIEW					
<input checked="" type="radio"/> Delimited	Column delimiters: <input type="text" value="Comma"/>	Range: <input type="text" value="A2:1000000"/>					
<input type="radio"/> Fixed Width	<input checked="" type="radio"/> Delimiter Options	Variable Names Row: <input type="text" value="1"/>					
DELIMITERS		SELECTION					
vibration.csv							
A		B		C		D	
time		axis		axis		axis	
datetime	Number	Number	Number	Number	Number	Number	Number
1 22:16:43	-0.048		-0.24		2.044		
2 22:16:43	2.044		-2.048		2.044		
3 22:16:43	2.044		-1.48		1.376		
4 22:16:43	2.044		-1.696		2.044		
5 22:16:43	2.044		-2.048		2.044		
6 22:16:43	0.6		-0.352		2.044		
7 22:16:43	2.044		-2.048		0.016		
8 22:16:43	2.044		-1.536		2.044		
9 22:16:43	1.456		0.032		2.044		
10 22:16:43	1.848		-1.408		2.044		
11 22:16:43	2.044		-2.048		2.044		
12 22:16:43	1.544		0.408		2.044		
13 22:16:43	2.044		-2.048		2.044		
14 22:16:43	1.568		-2.048		2.044		
15 22:16:43	2.044		0.152		2.044		
16 22:16:43	2.044		-1.992		2.044		
17 22:16:43	2.044		-1.688		0.516		
18 22:16:43	1.792		-0.872		1.928		
19 22:16:43	2.044		-1.128		2.044		
20 22:16:43	2.016		-0.856		2.044		
21 22:16:43	2.044		-1.872		-2.048		
22 22:16:43	-0.24		0.16		-0.004		

Vibration data



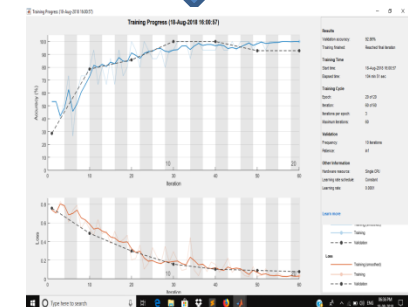
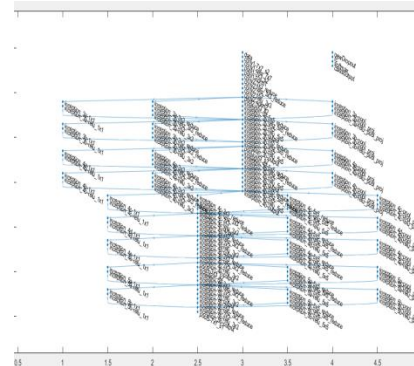
CWT of the data and plot the CWT



Create image data store

GOOD  
Or  
BAD

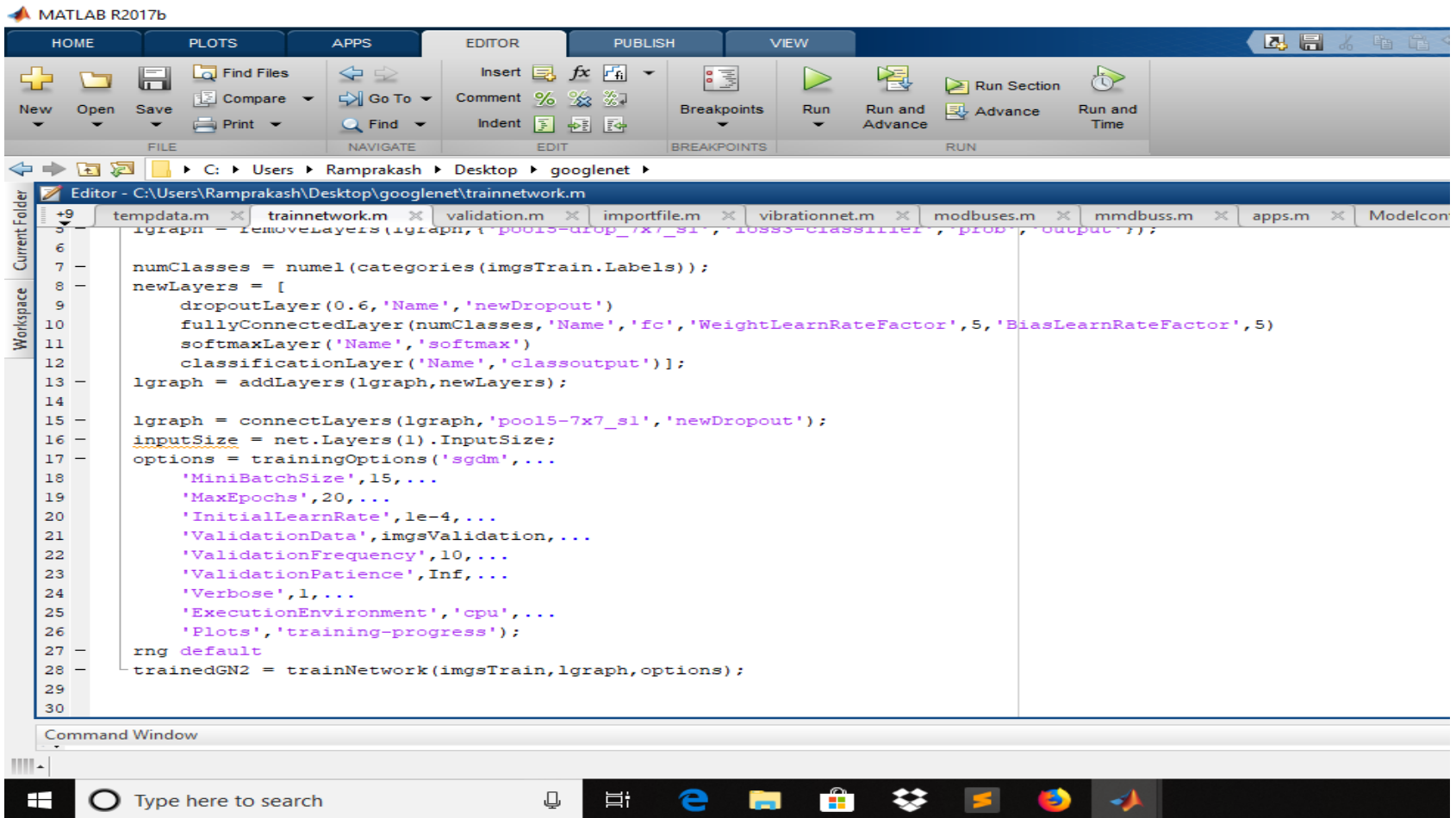
Motor status



Training the CNN



## Code snippets for training



The image shows the MATLAB R2017b interface with the Editor window open. The code snippet is for training a neural network, specifically a convolutional neural network (CNN) for vibration analysis. The code is as follows:

```
1 lgraph = removeLayers(lgraph,{'pool5=drop_7x7_s1','loss=classifier','prob','output'});
2
3 numClasses = numel(categories(imgsTrain.Labels));
4 newLayers = [
5     dropoutLayer(0.6,'Name','newDropout')
6     fullyConnectedLayer(numClasses,'Name','fc','WeightLearnRateFactor',5,'BiasLearnRateFactor',5)
7     softmaxLayer('Name','softmax')
8     classificationLayer('Name','classoutput')];
9 lgraph = addLayers(lgraph,newLayers);
10
11 lgraph = connectLayers(lgraph,'pool5-7x7_s1','newDropout');
12 inputSize = net.Layers(1).InputSize;
13 options = trainingOptions('sgdm',...
14     'MiniBatchSize',15,...
15     'MaxEpochs',20,...
16     'InitialLearnRate',1e-4,...
17     'ValidationData',imgsValidation,...
18     'ValidationFrequency',10,...
19     'ValidationPatience',Inf,...
20     'Verbose',1,...
21     'ExecutionEnvironment','cpu',...
22     'Plots','training-progress');
23
24 rng default
25 trainedGN2 = trainNetwork(imgsTrain,lgraph,options);
```

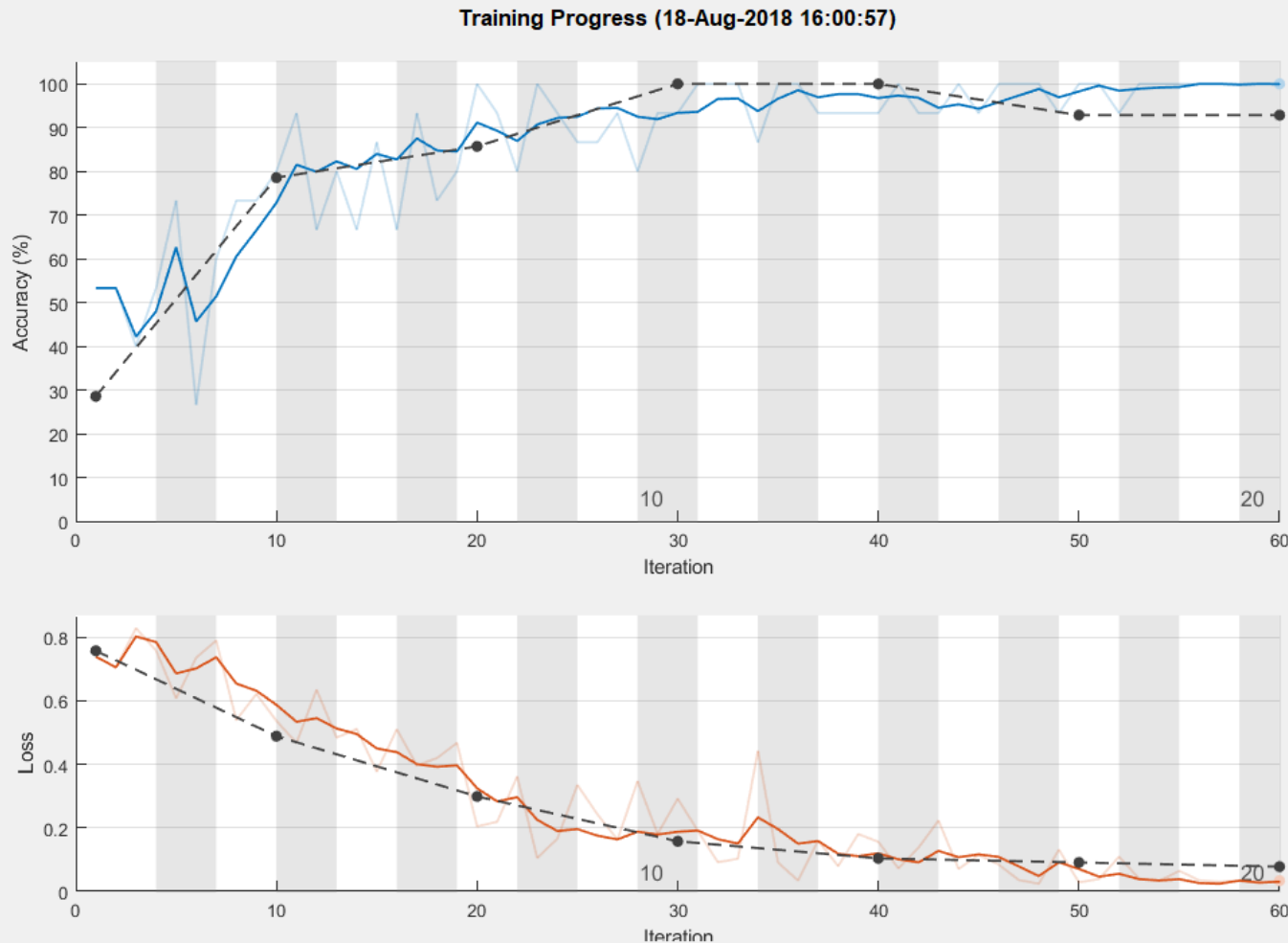
# VIBRATION ANALYSIS:



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Training Progress (18-Aug-2018 16:00:57)



## Results

Validation accuracy: 92.86%  
Training finished: Reached final iteration

## Training Time

Start time: 18-Aug-2018 16:00:57  
Elapsed time: 104 min 51 sec

## Training Cycle

Epoch: 20 of 20  
Iteration: 60 of 60  
Iterations per epoch: 3  
Maximum iterations: 60

## Validation

Frequency: 10 iterations  
Patience: Inf

## Other Information

Hardware resource: Single CPU  
Learning rate schedule: Constant  
Learning rate: 0.0001

## Learn more



Type here to search

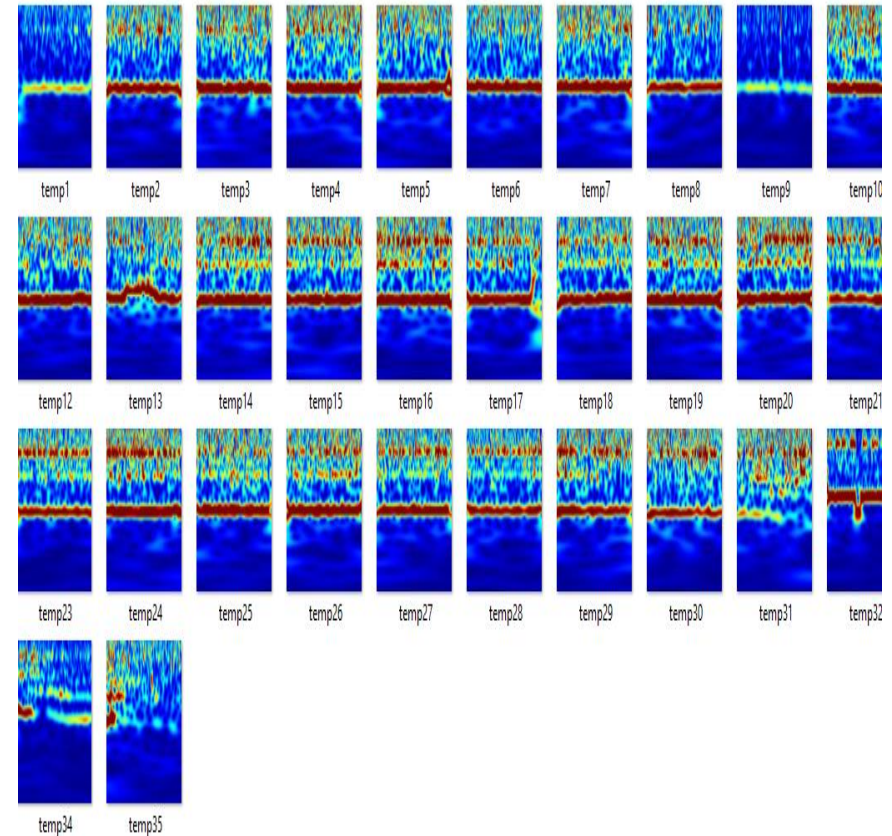
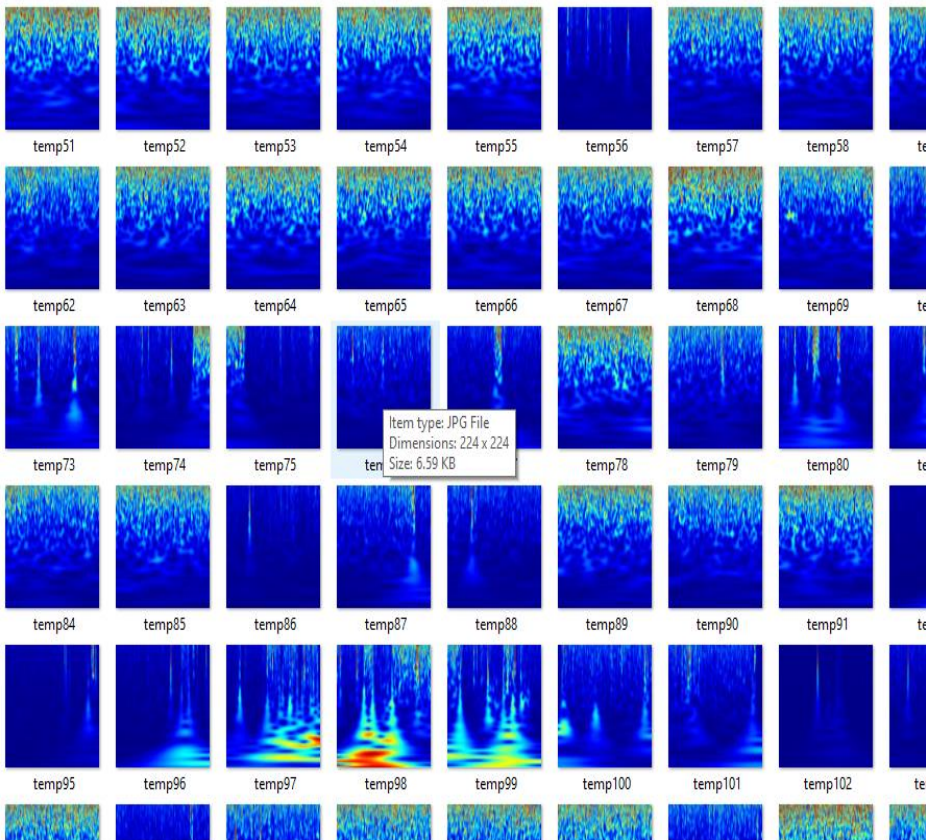


06:36 PM 18-08-2018





## DATA SET





- **First few layers of the CNN**
  - Input Layer       $224*224*3$
  - conv1-7x7\_s2     $64 \ 7*7*3$
  - conv1-relu\_7x7 ReLu
  - pool1-3x3\_s2     $3*3$  max pooling
- **Last four layers of the CNN**
  - Dropout layer
  - Fully connected layer
  - Softmax layer
  - Class output layer
  - [Vibration.pptx](#) (detailed explanation of ML)



# BUSINESS PLAN

# Why Smart Slurry Pumping System?



- **Slurry is one of the most challenging materials for a pump to handle.**
- **This is due to uncertainty in the nature of suspended matter.**
- **Hence the motor would be subjected to fluctuating loading conditions.**



## Customer Requirement Identification



- Survey is taken from our local industries.
- Continuous monitoring of vibration is required.
- Continuous monitoring of insulation temperature is required.
- Monitoring of standard electrical parameters such as voltage, current, frequency and harmonic distortions are also required
- Retrofitting is most important







# **Different types of fault in rotating machines in percentage**

## **Causes of different faults**

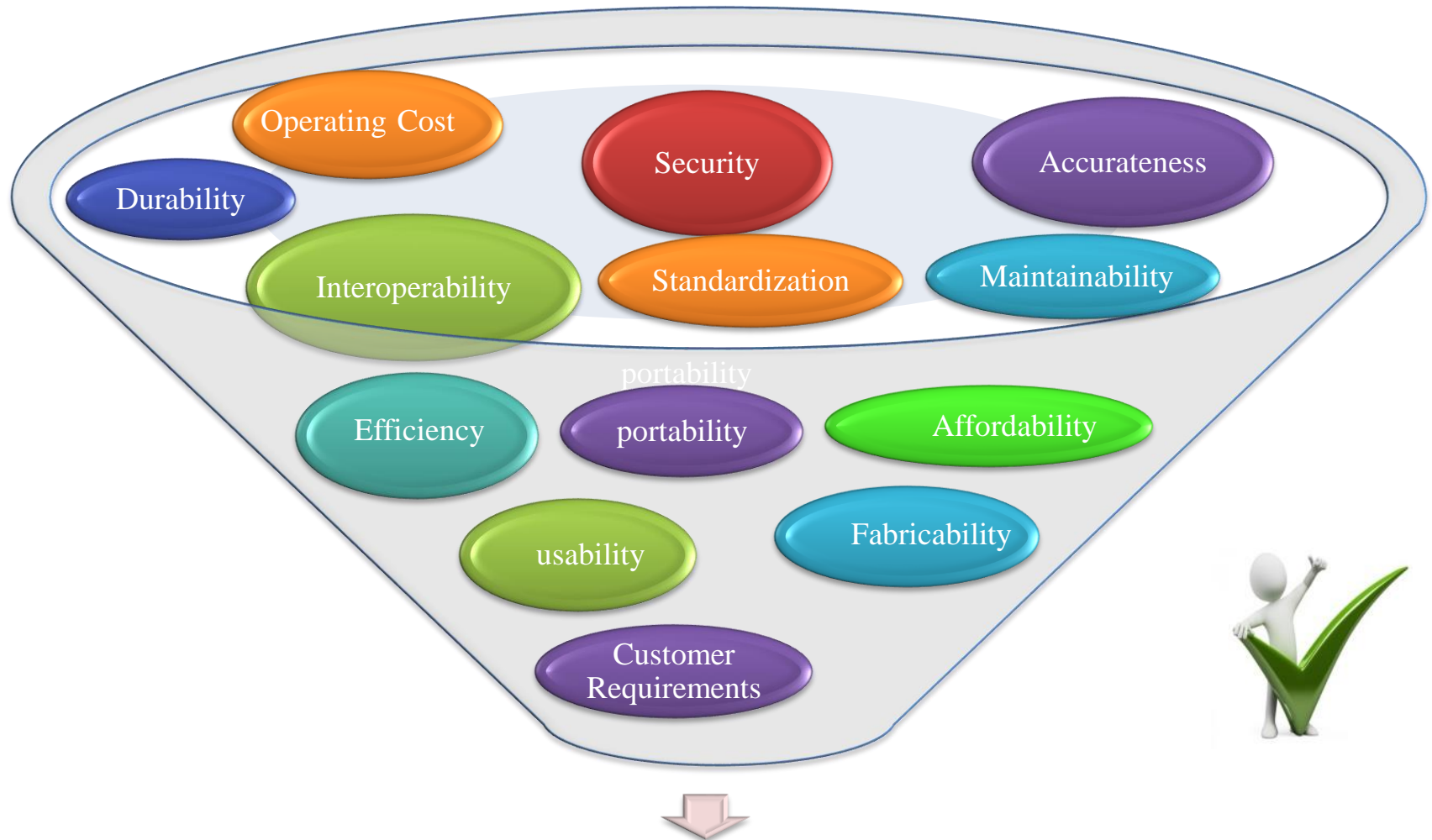
**Reason routinely highlighted for all faults are:**

**Thermal stress – Temperature rise**

**Mechanical stress – Vibration**

**Electrical stress – Faults and harmonics**

# Benefits Of The System



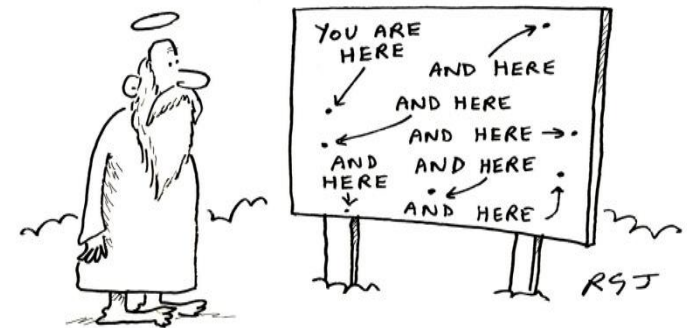
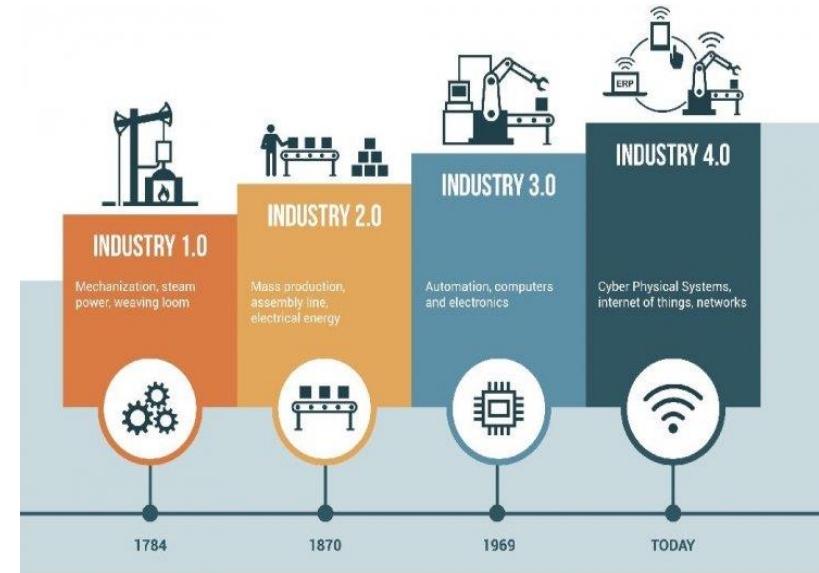
## Smart Slurry Pumping System

EngiNx 2018

# Uniqueness of the system



- **Industry 4.0**
- **Centralized operation**
- **Less maintenance**
- **Digital Twin**
- **Cloud computing technology**
- **Parallel processing of all parameters.**
- **Generalized Solution to problem of industrial omnipresent i.e. motor.**



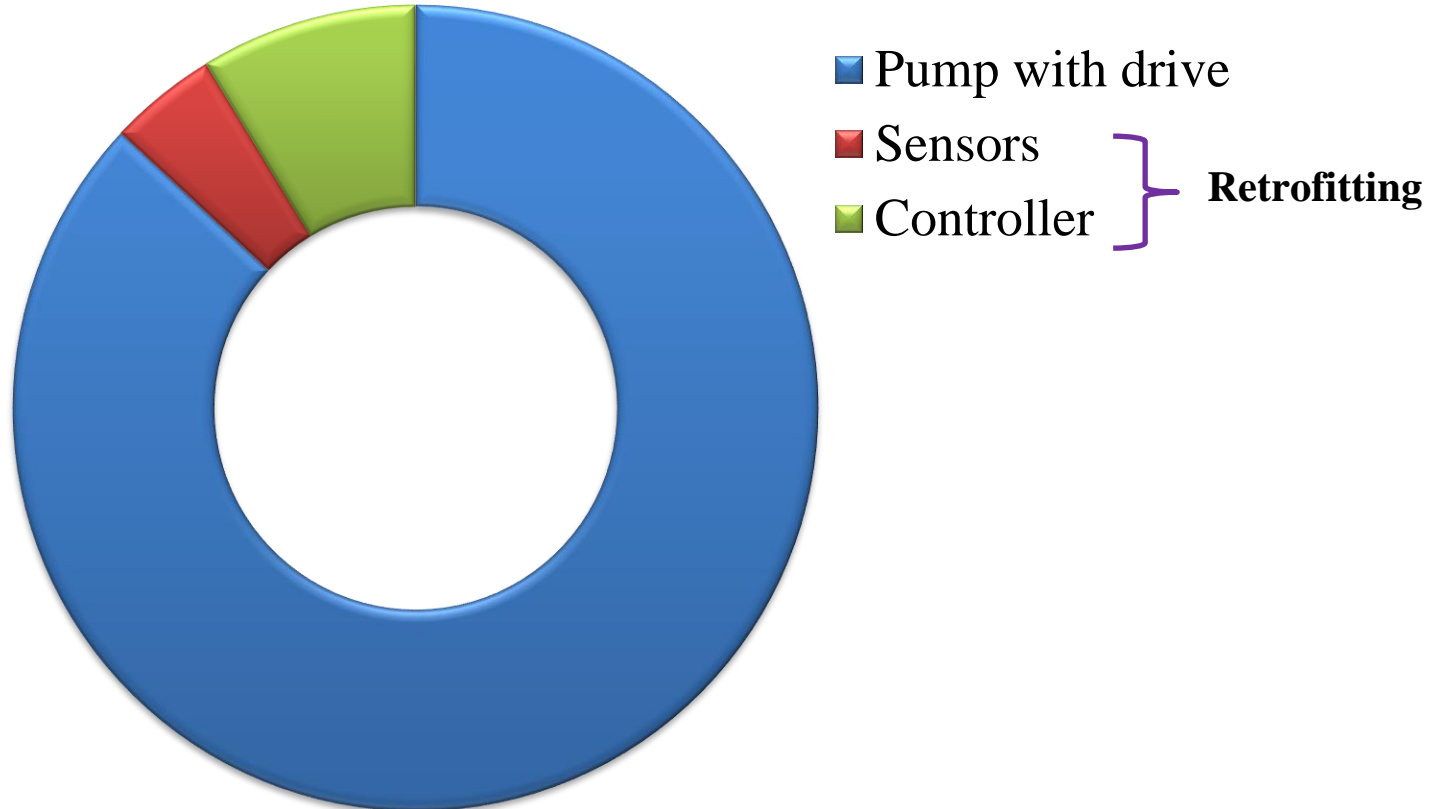


# Bill of Materials



S. No.	Components	Amount (Rs.)
1	Tri-axial accelerometer	500
2	Non-contact type temperature sensor	500
3	Contact type temperature sensor	200
4	Raspberry Pi 3 Model B	3000
5	Energy management unit	10000
6	Software development	20000
	<b>TOTAL</b>	<b>34200</b>

# Implementation Cost Distribution



# Improving Reliability and Uptime



**The Digital Twin  
can act before  
product failure  
begins.**



**Better product  
reliability,  
increased uptime  
and lower  
maintenance cost  
creates customer  
value.**



- **Tease**

**What is the problem?**

- **Please**

**What is the solution?**

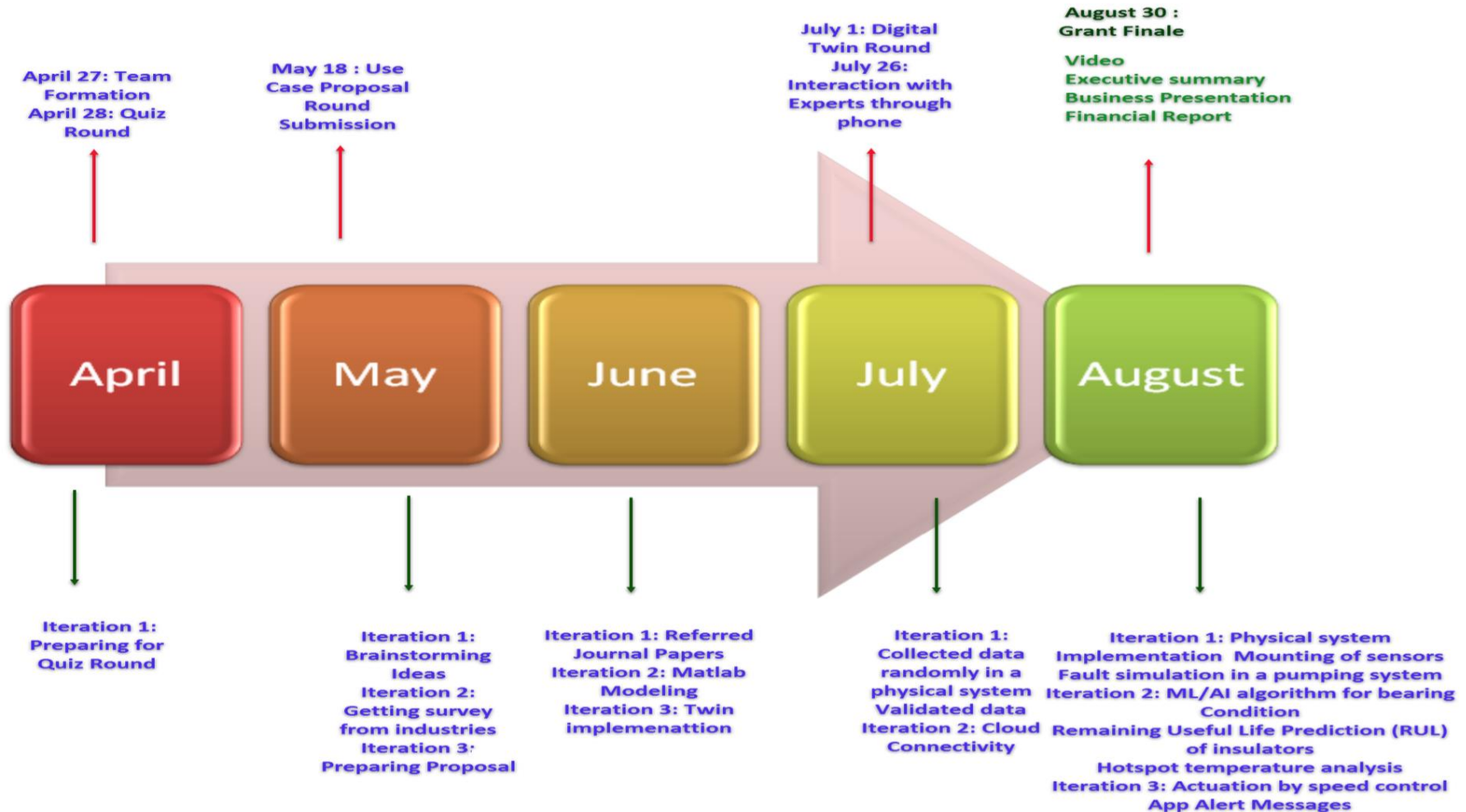
- **Seize**

**What are the  
opportunities?**

**Demonstrating  
Prototype.**



# Milestones





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- **Schoen RR, Habetler TG, Kamran F, Bartheld RG (1995) Motor bearing damage detection using stator current monitoring. IEEE Trans Ind Appl 31(6):1274–1279**
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**THANK YOU**