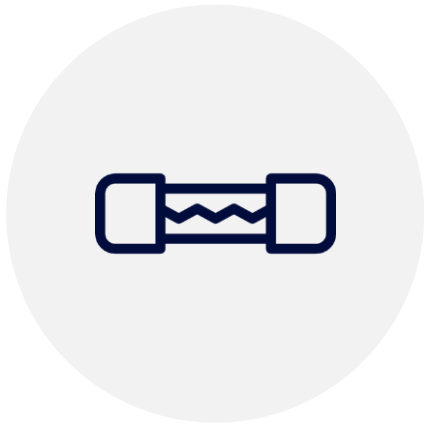




# FAULT DETECTION

SAMPLE METHODOLOGIES EXPLAINED

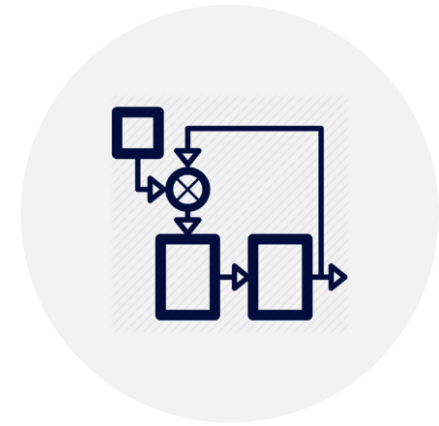
# PROGNOSTICS HEALTH MANAGEMENT (PHM) - PROCESS OF PREDICTING THE FUTURE RELIABILITY OF A PRODUCT BY ASSESSING THE EXTENT OF DEVIATION OR DEGRADATION OF A PRODUCT FROM ITS EXPECTED NORMAL OPERATING CONDITIONS



EXPENDABLE DEVICES

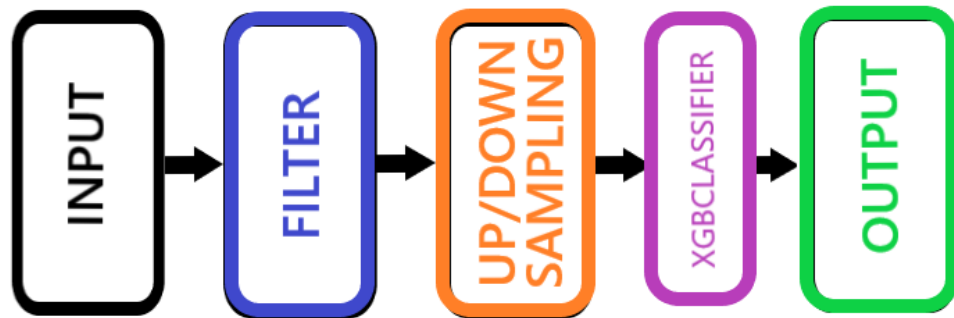


PERFORMANCE  
PARAMETERS

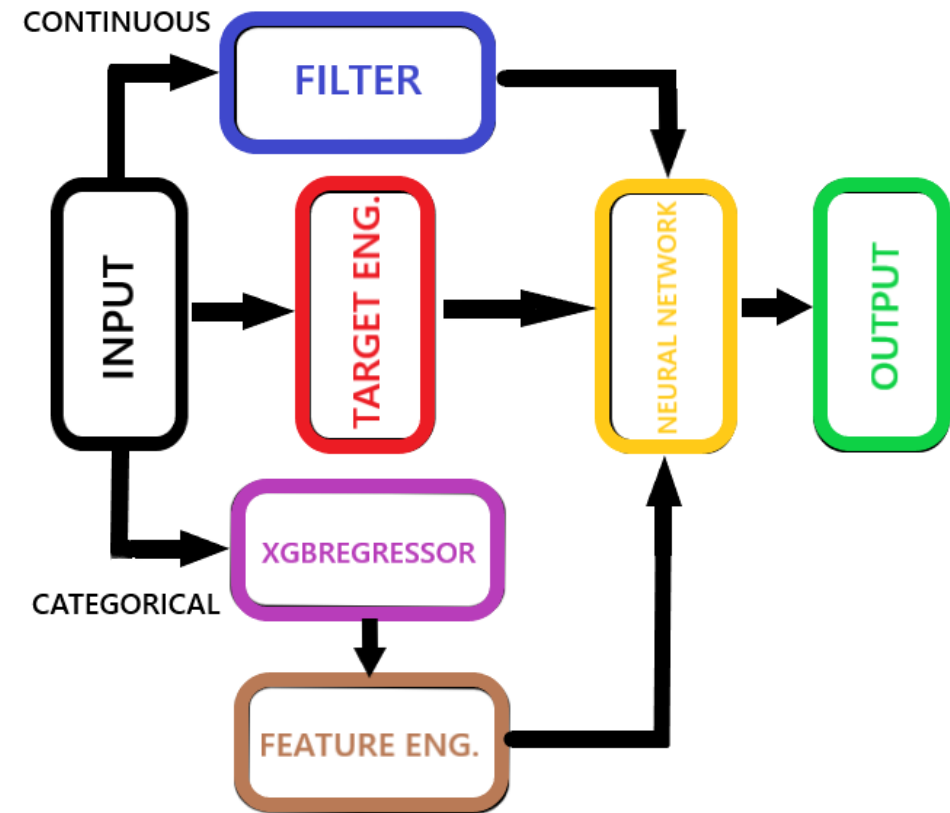


DAMAGE MODELING

# DAMAGE MODELING

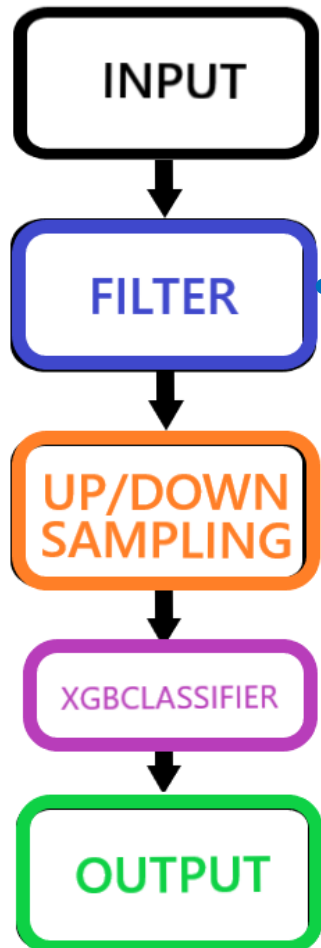


FAULT CLASSIFICATION

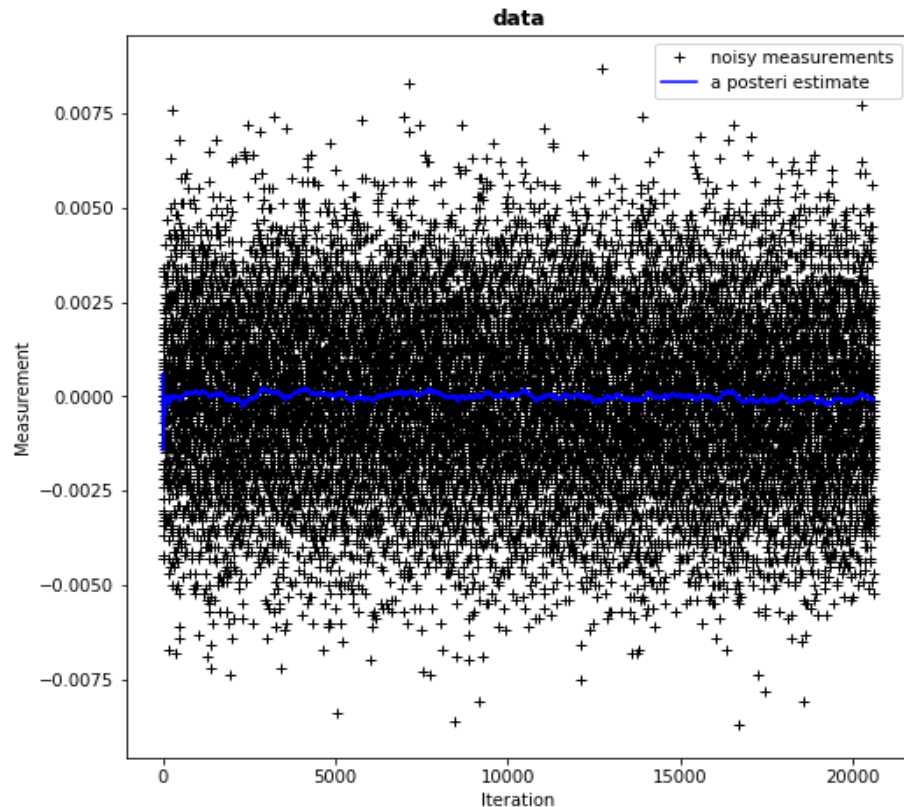


REMAINING USEFUL LIFE

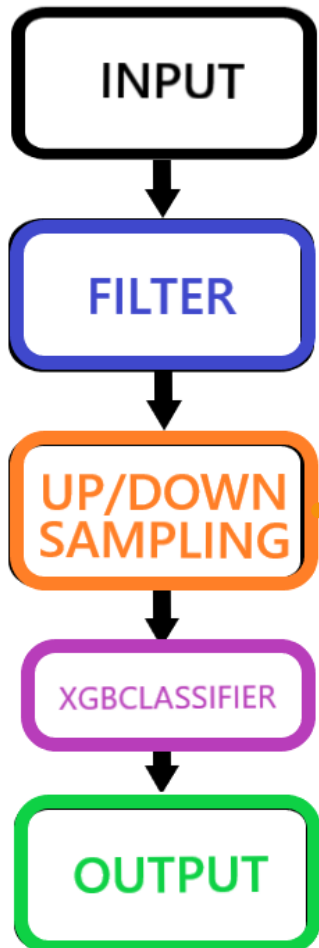
# KALMAN FILTER



AKA: linear quadratic estimation ( LQE ) uses a series of measurements observed over time, containing statistical noise and other inaccuracies, and produces estimates of unknown variables that tend to be more precise than those based on a single measurement alone.

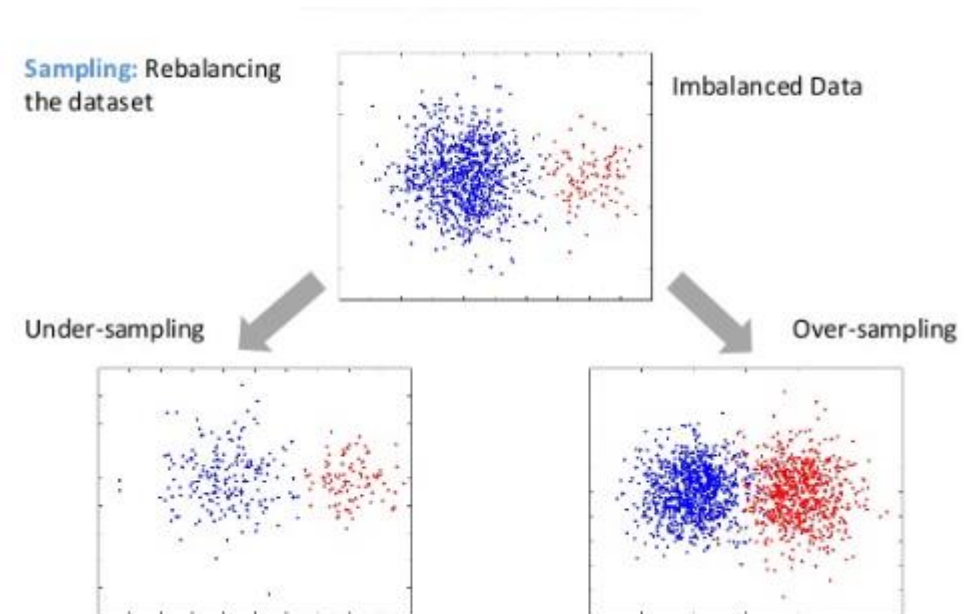


# UP / DOWN SAMPLING

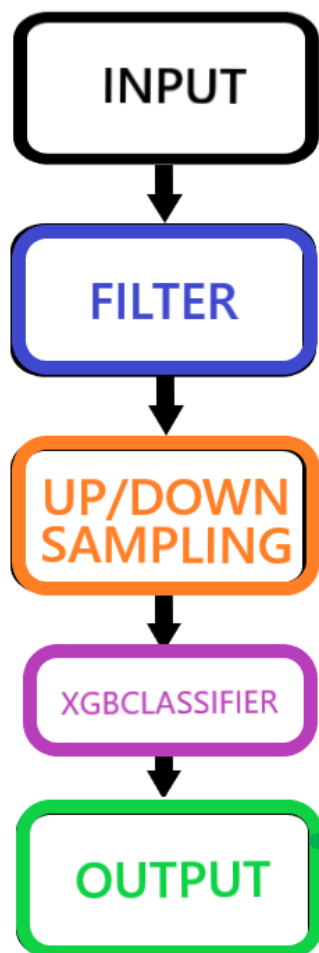


## **SMOTE** stands for **Synthetic Minority / Majority Oversampling Technique**.

This is a statistical technique for increasing or decreasing the number of cases in your dataset in a balanced way. The module works by generating new or removing instances from existing cases that you supply as input. This implementation of SMOTE does not change the number of non-target cases.

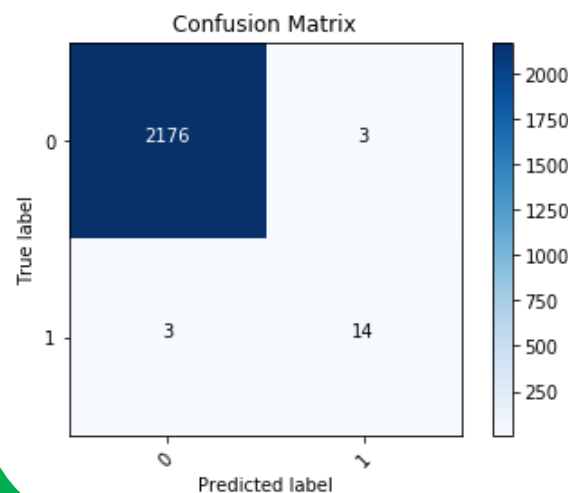


# RESULTS



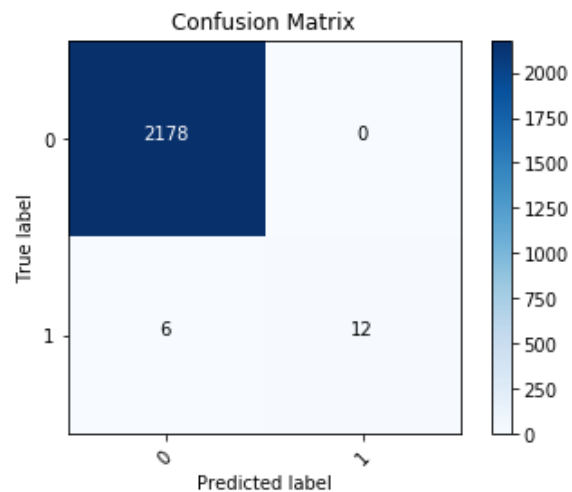
## Out-Of-Box

PRECISION: 0.91  
RECALL: 0.91  
FI: 0.91  
CROSS-VAL: 0.997



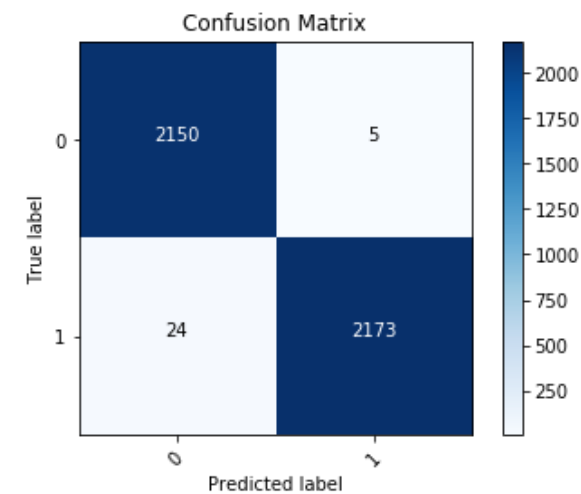
## After Filter

PRECISION: 1.00  
RECALL: 0.83  
FI: 0.90  
CROSS-VAL: 0.997

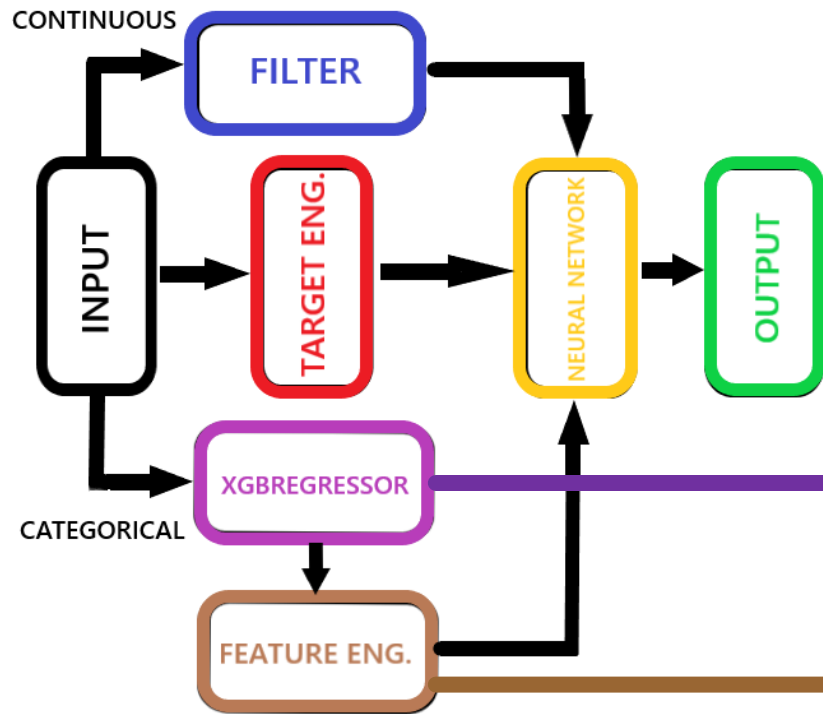


## After Filter + Upsample

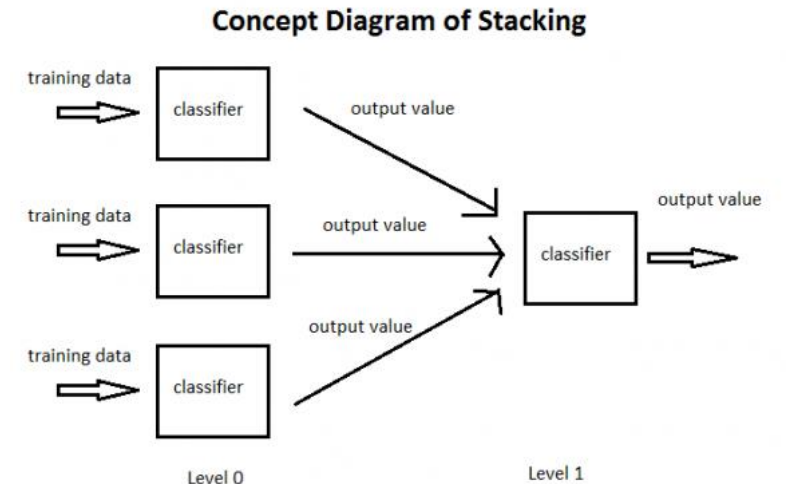
PRECISION: 0.99  
RECALL: 0.99  
FI: 0.99  
CROSS-VAL: 0.956



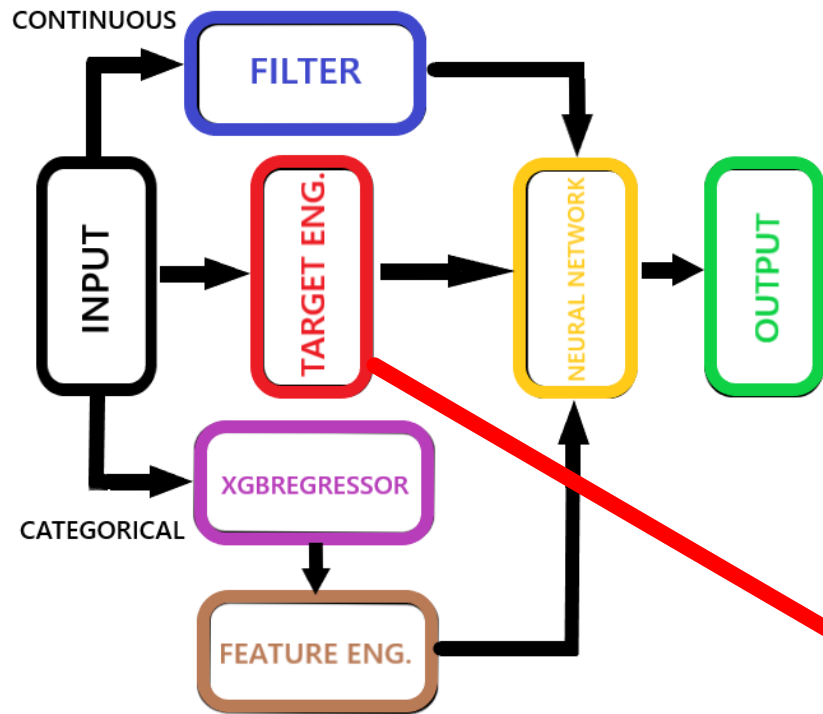
# MODEL STACKING



**Stacking, called meta ensembling,** is a model ensembling technique used to combine information from multiple predictive models to produce a new one. In stacking, the combining mechanism is that the output of the initial model will be used as training data for another model.



# TARGET ENGINEERING

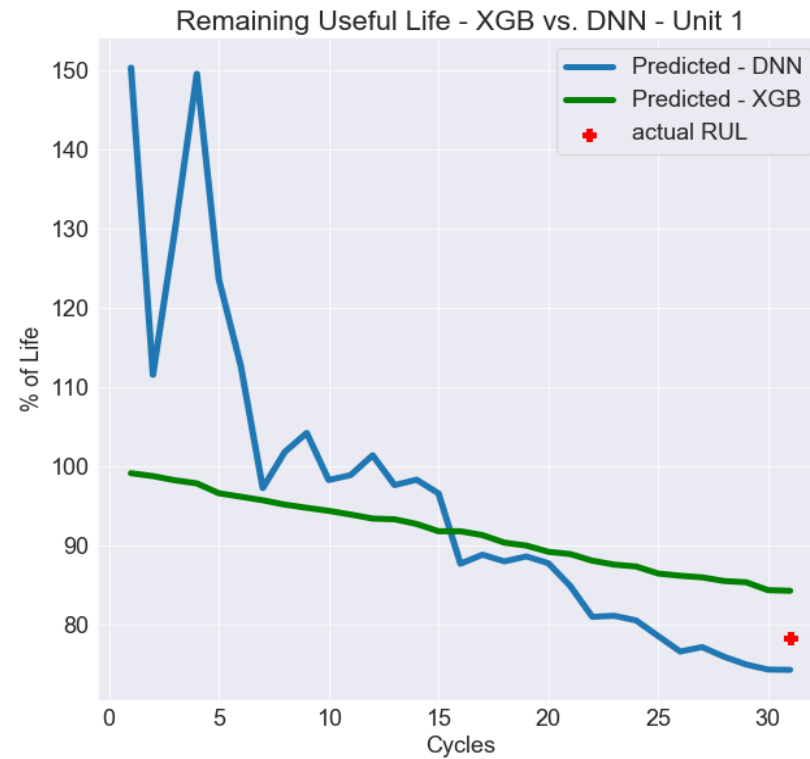
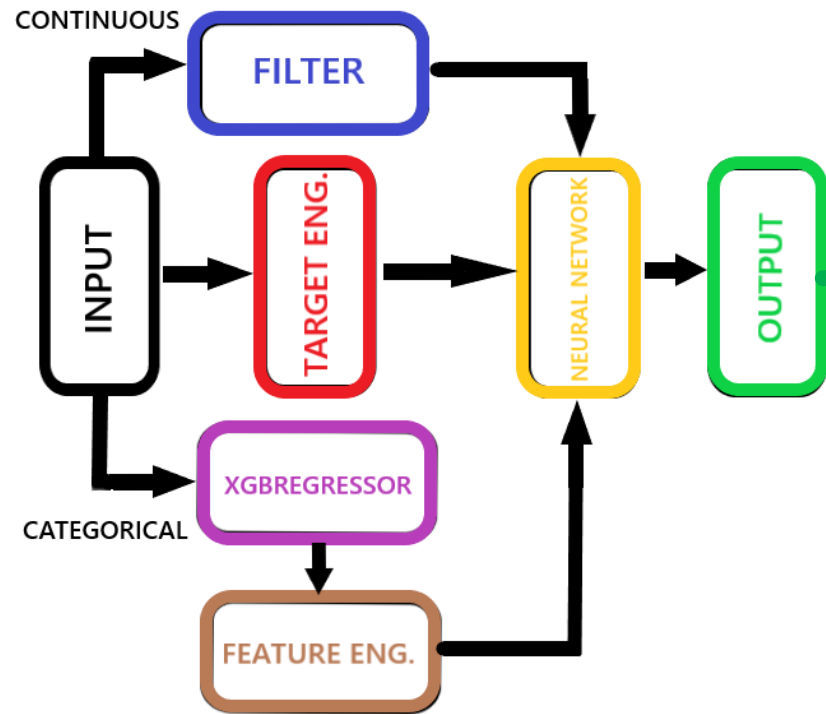


**Feature engineering** is the process of using **domain knowledge** of the data to create features that make machine learning algorithms work. If engineering is done correctly, it increases the predictive power of machine learning algorithms and help facilitate the machine learning process. **Feature engineering is an art.**

```
unit = 1
rul_perc = []
while unit <= 100:
    cycles = train['cycle'][train['unit']==unit].values
    val = sorted(cycles, reverse=True)
    percent = [i / max(val) for i in val]
    rul_perc.append(percent)
    unit = unit + 1
```

- ITERATES THROUGH EACH UNIT
- PULLS / REVERSE SORTS VALUES
- CONVERTS TO % OF REMAINING LIFE

# RESULTS



ROOT MEAN SQUARED  
ERROR (RMSE)

XGB – 48.40

DNN – 34 - 40



# THANK YOU

## JASON RICHARDS

'THE DATA DETECTIVE'

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