

## Mathematical Calculations for IBD Model (GPR Model)

1.  $X_{\text{scaled}} = (X - \mu) / \sigma$

Where:

- $X$  is the original value of the feature.
- $\mu$  is the mean of the feature.
- $\sigma$  is the standard deviation of the feature.

2. 
$$k(x, x') = \exp \left( -\frac{\|x - x'\|^2}{2\ell^2} \right)$$

- `kernel = RBF(1.0, (1e-4, 1e3))`
- $\ell$  is **1.0** initially.
- $k(x, x')$  is the similarity (or `kernel`) between two input points  $x$  and  $x'$ .
- $\|x - x'\|^2$  is the squared Euclidean distance between the two points  $x$  and  $x'$ :  
 $\|x - x'\|^2 = \sum_{i=1}^n (x_i - x'_i)^2$  where  $x_i$  and  $x'_i$  are the individual components of the vectors  $x$  and  $x'$ .

- $$\|x - x'\|^2 = \sum_{i=1}^n (x_i - x'_i)^2$$

- $\ell$  is the **length scale** parameter, which controls the width of the kernel. This is determined during the model training process, with bounds defined by (1e-4, 1e3).

### 3. Train-Test Split (`train_test_split`)

The `train_test_split` function splits your dataset into training and testing datasets.

The formula for splitting the data into training and testing sets:

`Xtrain, Xtest = train_test_split(Xencoded, yencoded, test_size=0.5, random_state=100)`

Where:

- `Xencoded` is the feature matrix.
- `yencoded` is the target variable.
- The **test size** is 50% of the data (`test_size=0.5`), meaning you use 50% for training and 50% for testing.
- **`random_state=100`** is used to control the randomness of the data splitting.

### 4. Prediction and Thresolding

`ypred, sigma_pred = GPR.predict(Xtest, return_std=True)`

Where:

- `ypred` are the predicted continuous values.

- $\sigma_{\text{pred}}$  is the uncertainty (standard deviation) of the predictions.

Since we're performing **classification** (CKD vs. non-CKD), we apply a threshold of **0.7** to convert the continuous predictions into binary values (0 or 1):

$$y_{\text{pred\_class}} = \begin{cases} 1 & \text{if } y_{\text{pred}} > 0.7 \\ 0 & \text{if } y_{\text{pred}} \leq 0.7 \end{cases}$$

## 5. Confusion Matrix

The **confusion matrix** is computed using:

```
MatrixConfusion Matrix=confusion_matrix(ytest,ypred_class)
```

This matrix gives:

- True Positives (TP): Predicted 1, Actual 1
- True Negatives (TN): Predicted 0, Actual 0
- False Positives (FP): Predicted 1, Actual 0
- False Negatives (FN): Predicted 0, Actual 1

## 6. Accuracy Calculation

The **accuracy** is computed using:

$$\text{Accuracy} = \frac{\text{TP} + \text{TN}}{\text{FP} + \text{FN} + \text{TP} + \text{TN}}$$

Where:

- TP = True Positives
- TN = True Negatives
- FP = False Positives
- FN = False Negatives

## 7. Accuracy per Point

The **accuracy per point** is calculated by measuring how close each predicted value is to the actual value:

$$\text{Accuracy per point} = 1 - |y_{\text{test}} - y_{\text{pred}}|$$