

Using same example in lab 03 (about KNN) and modify the code

- Define which K-value will be the best to split the data when using cross validation (**determine the number of K folds**)
 - Assign value for the K from 5 to 25 for the same problem and report the **accuracy for each k-value**
 - Define **which k value** will outperform other values.
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- Before fed the training data directly to the classifier, apply pre-processing for your data
 - To do that, apply zero-Score which illustrated earlier in this lab
 - Calculate the **mean** value for the feature
 - ❖ You can use built in method
 - Calculate the **Standard deviation** value for the feature (first, calculate the variance)
 - ❖ You **cannot use built in method**
 - For each x, subtract from mean value, and divide over standard deviation
 - Then apply again the assignment 2 solution to report which k fold will be the best after previous steps.

$$z = \frac{x - \mu}{\sigma}$$

Select your own-dataset and do the following (Binary classification Task)

- According to your understanding of the concepts of **Recall** and **Precision** and the trade-off between them

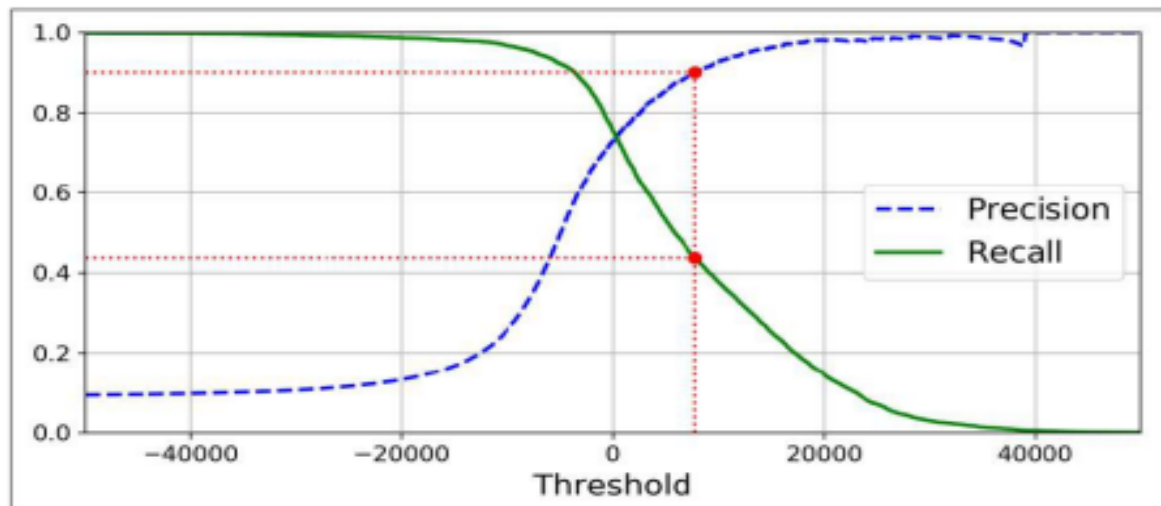


Figure 3-4. Precision and recall versus the decision threshold

- 1- Apply **stochastic gradient classifiers**(SGD), and report the
 - Accuracy
 - Recall
 - Precision
 - Specificity
- 2- Plot precision and recall as functions of the threshold value
- 3- After that, from the curve, define the best threshold value (θ_1) that makes Recall twice Precision
- 4- Using this **threshold value (θ_1)**, and perform the classification process using the new threshold-value **θ_1**
- 5- again, from the curve, define the best threshold value (θ_2) that makes Precision twice Recall
- 6- Using this **threshold value (θ_2)**, and perform the classification process using the new threshold-value **θ_2**