

# Medical Image Analysis

## 4. Medical image classification(3)

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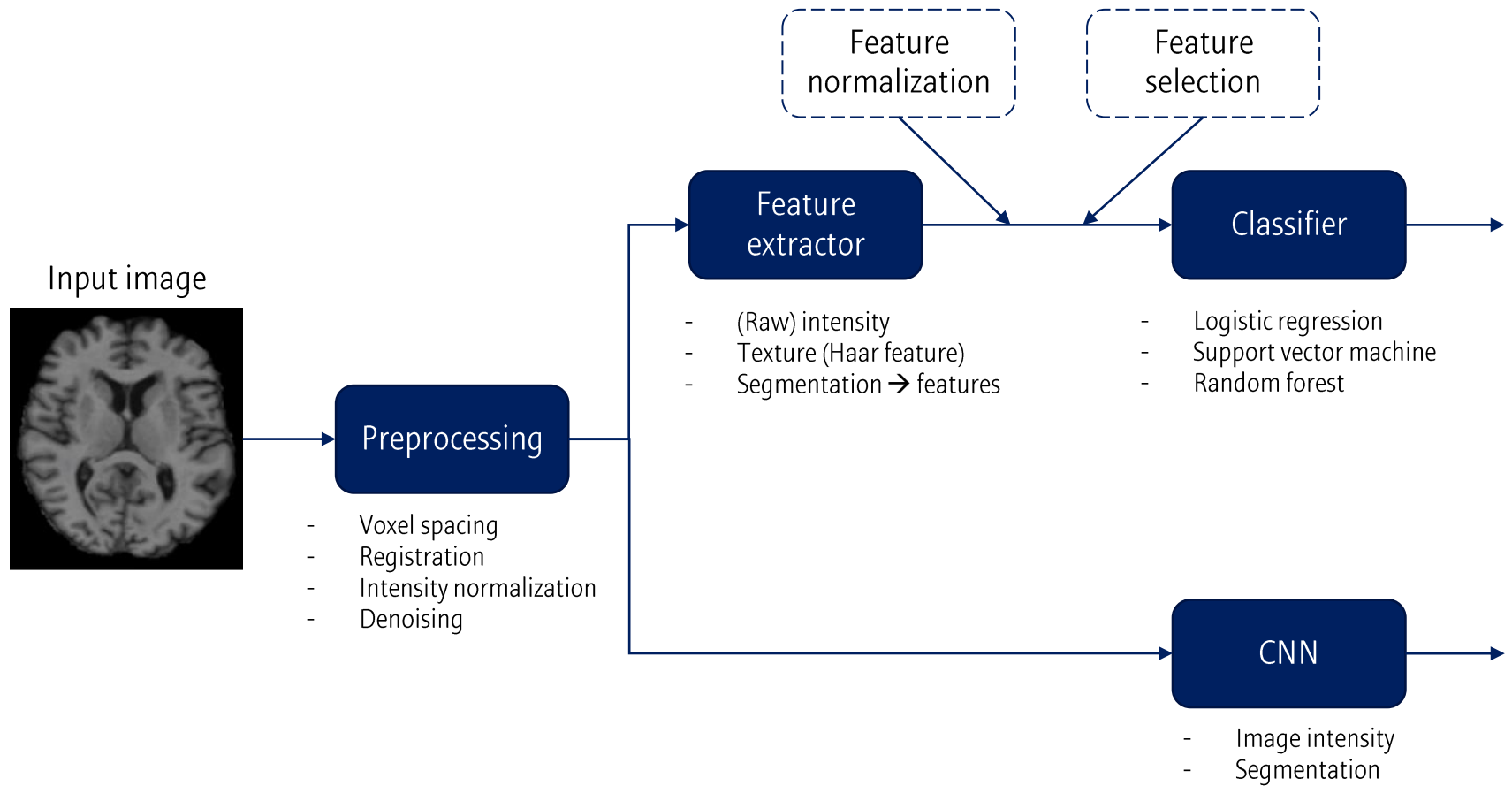
Taeyang Yang

Oct. 2020

<https://tyami.github.io/>

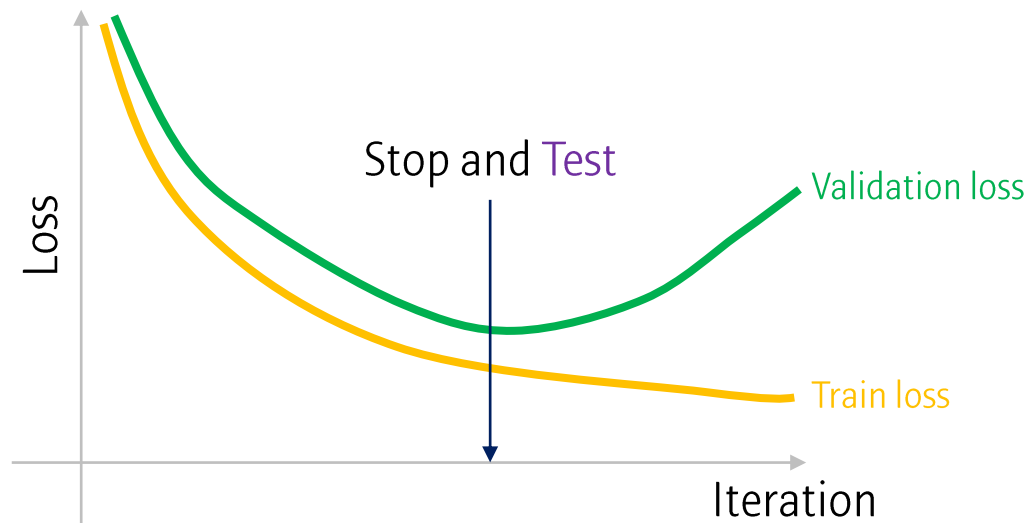
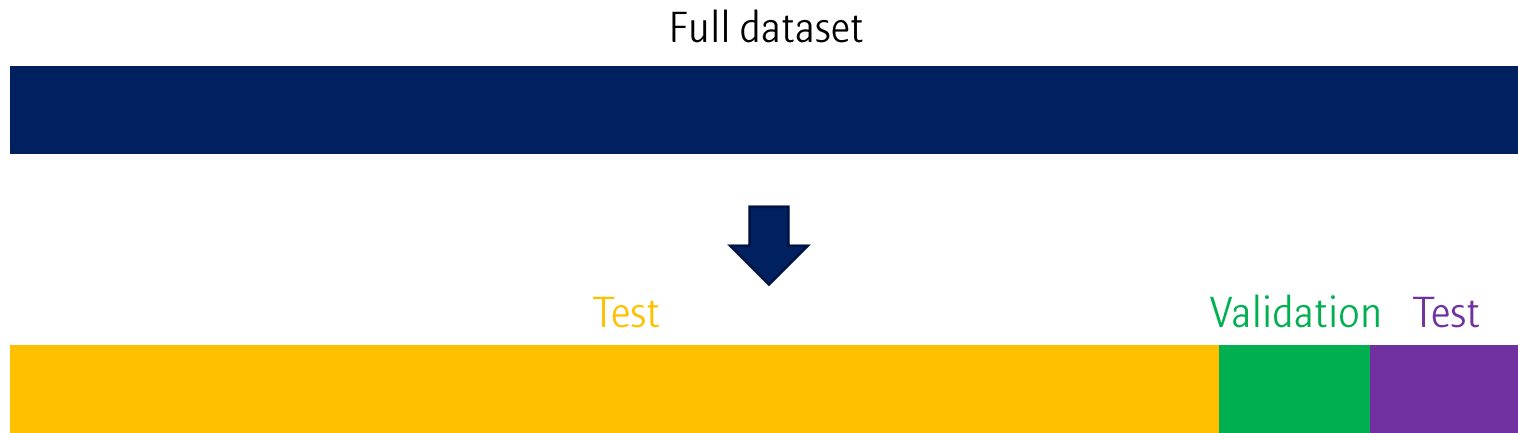
# 1. Overall procedure

## Overall procedure for classification



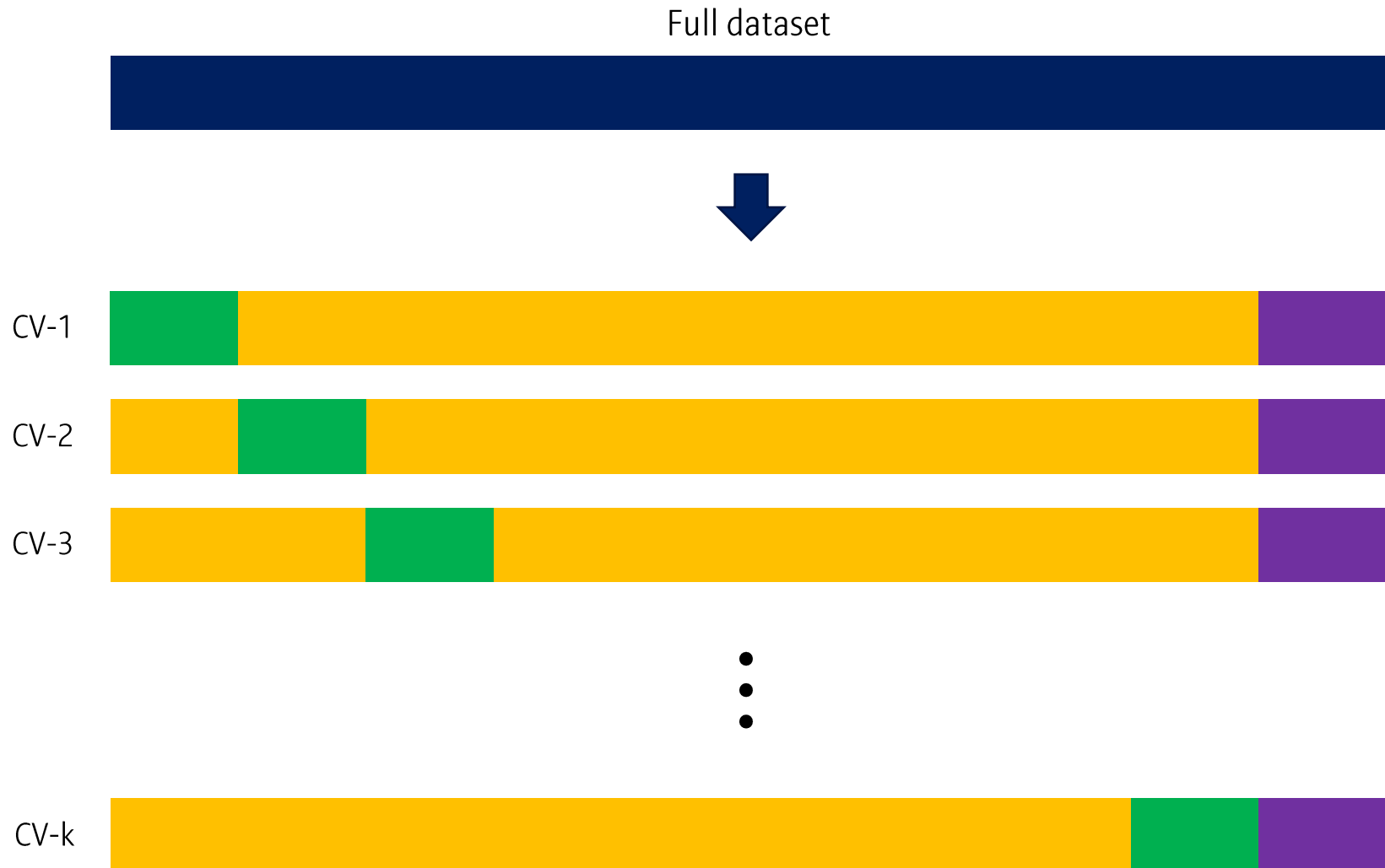
# 2. Validation

## Validation



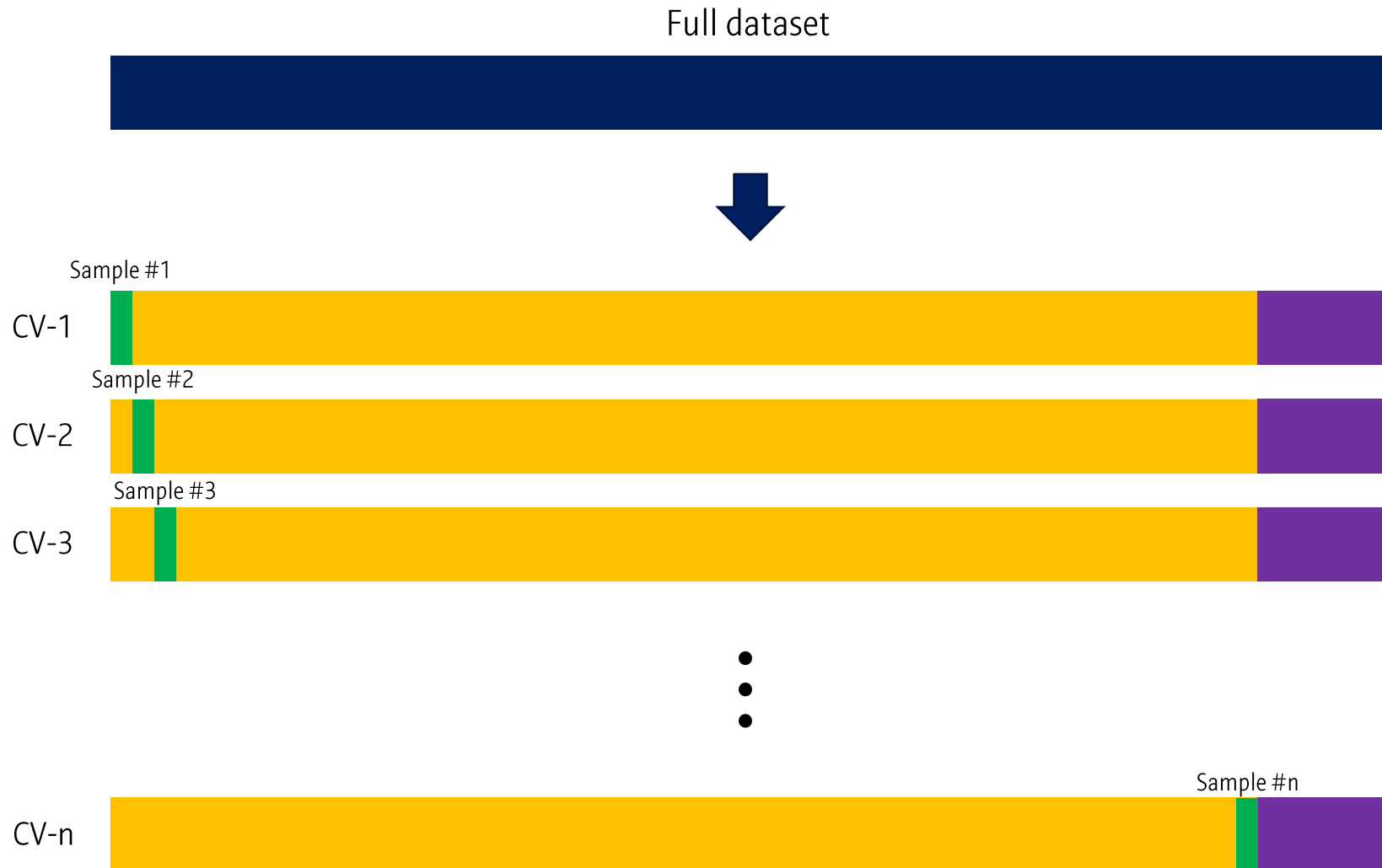
## 2. Validation

### K-fold Cross validation



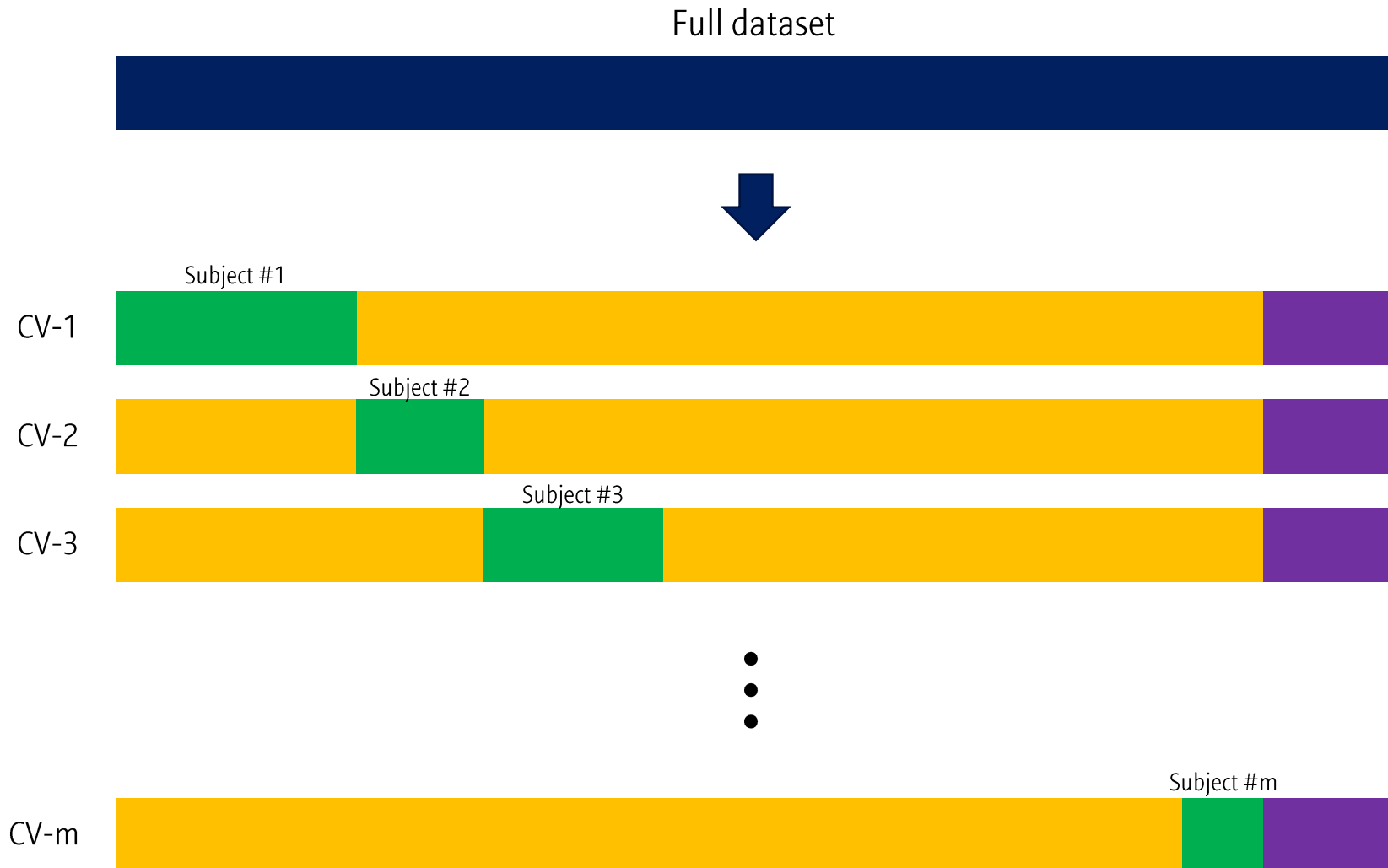
## 2. Validation

### Leave-one-out Cross validation



## 2. Validation

### Leave-one-subject-out Cross validation



# 3. Overfitting / Regularization

- Regularization: Ridge (L2), Lasso (L1), and Elastic Net regression
  - 이전 포스팅 참고: <https://tyami.github.io/machine%20learning/regularization-Ridge-Lasso-ElasticNet/>

# 4. Transfer Learning

## Large dataset in normal object classification

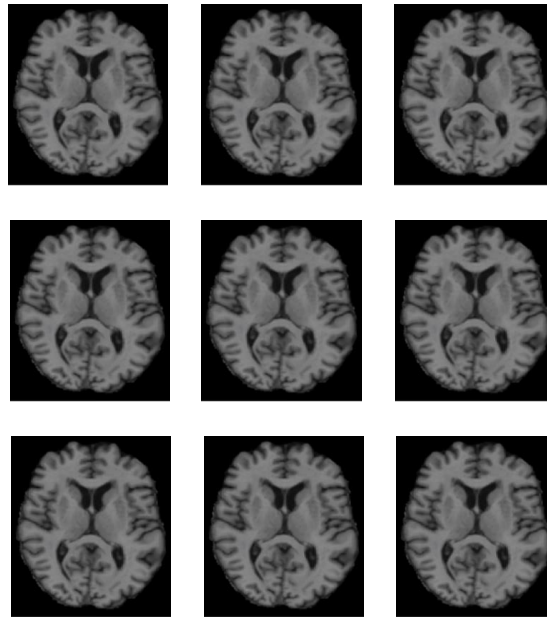
- Large dataset
  - ImageNet: 14 millions





# 4. Transfer Learning

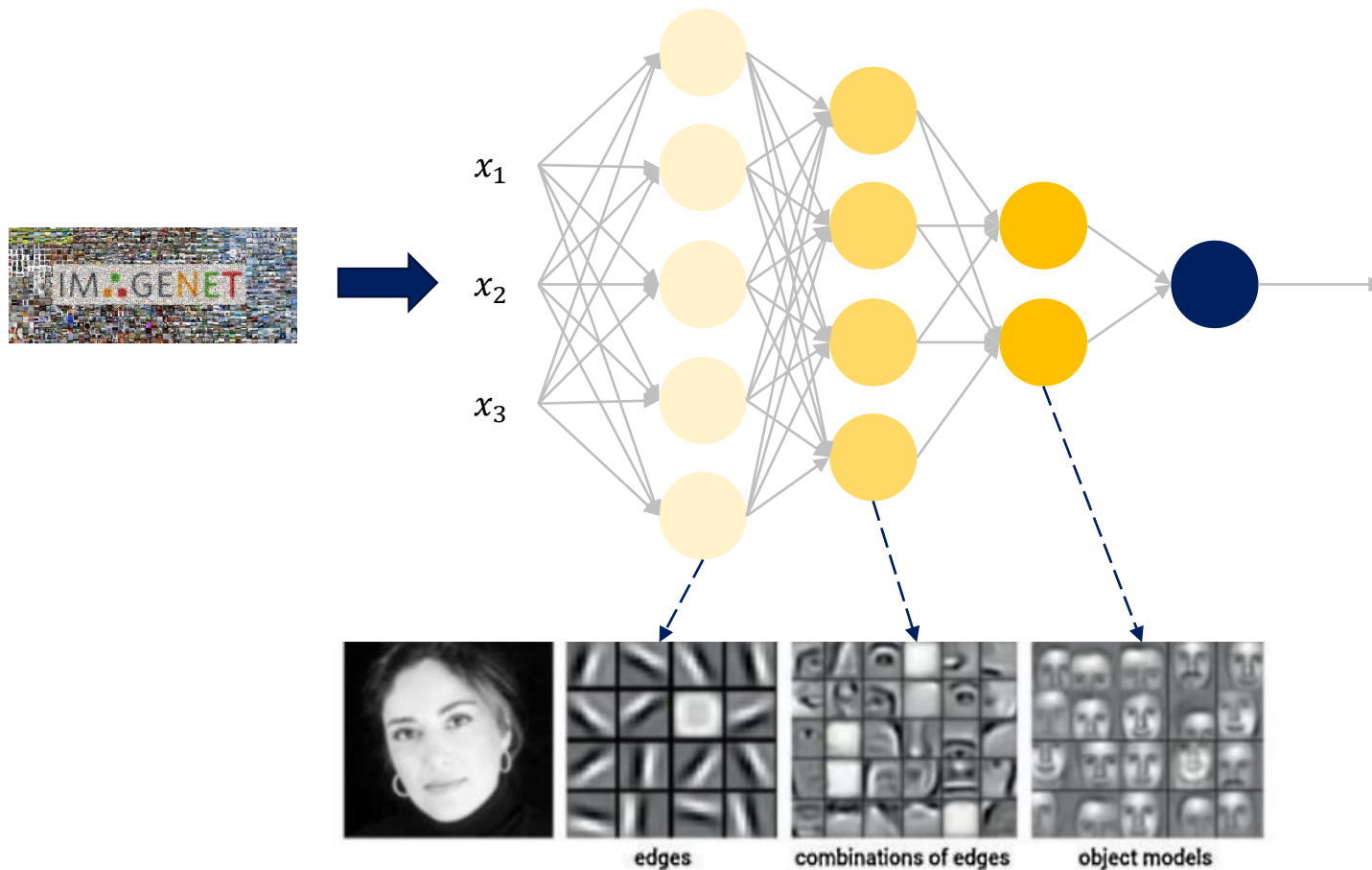
Small dataset in domain problem



# 4. Transfer Learning

## Original training

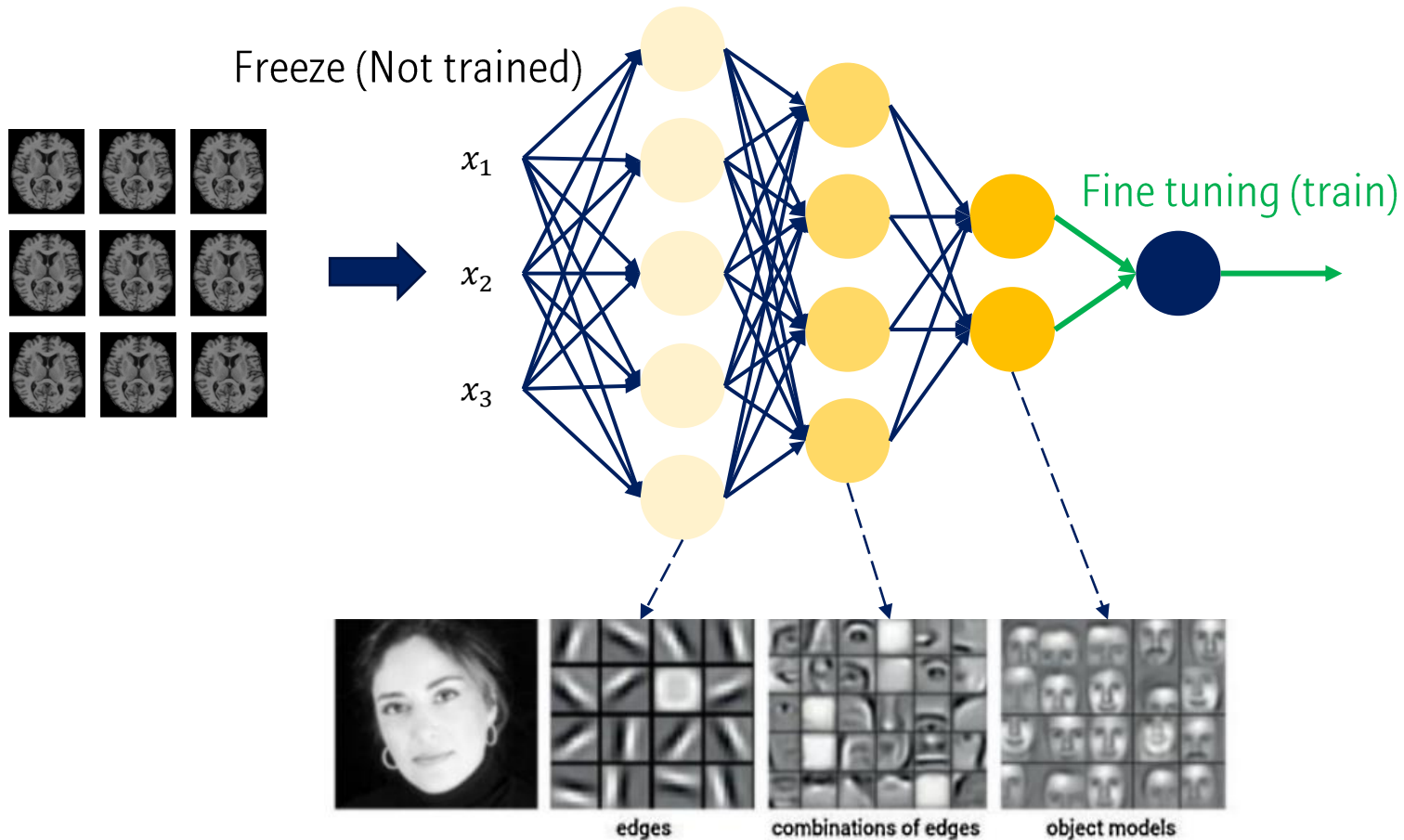
- Assumption: Low-level feature is similar between medical image and normal image



# 4. Transfer Learning

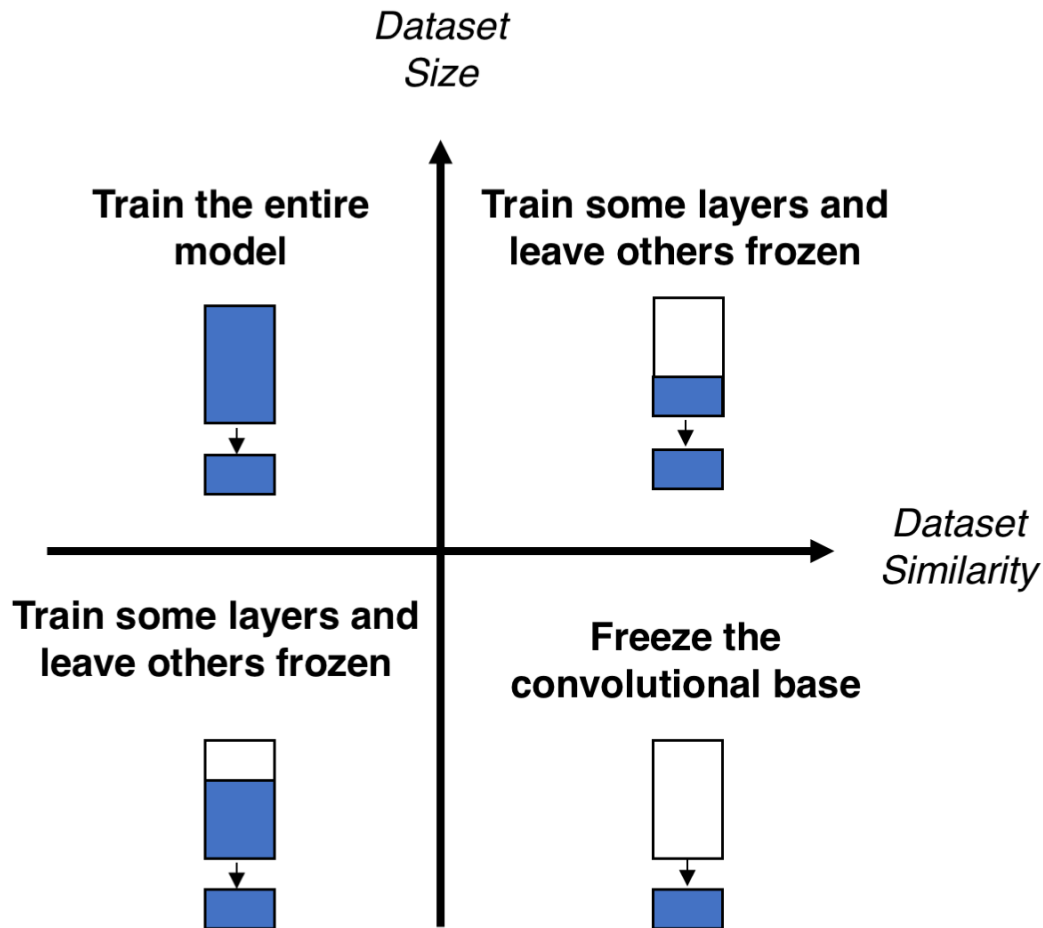
## Fine tuning

- Assumption: Low-level feature is similar between medical image and normal image



# 4. Transfer Learning

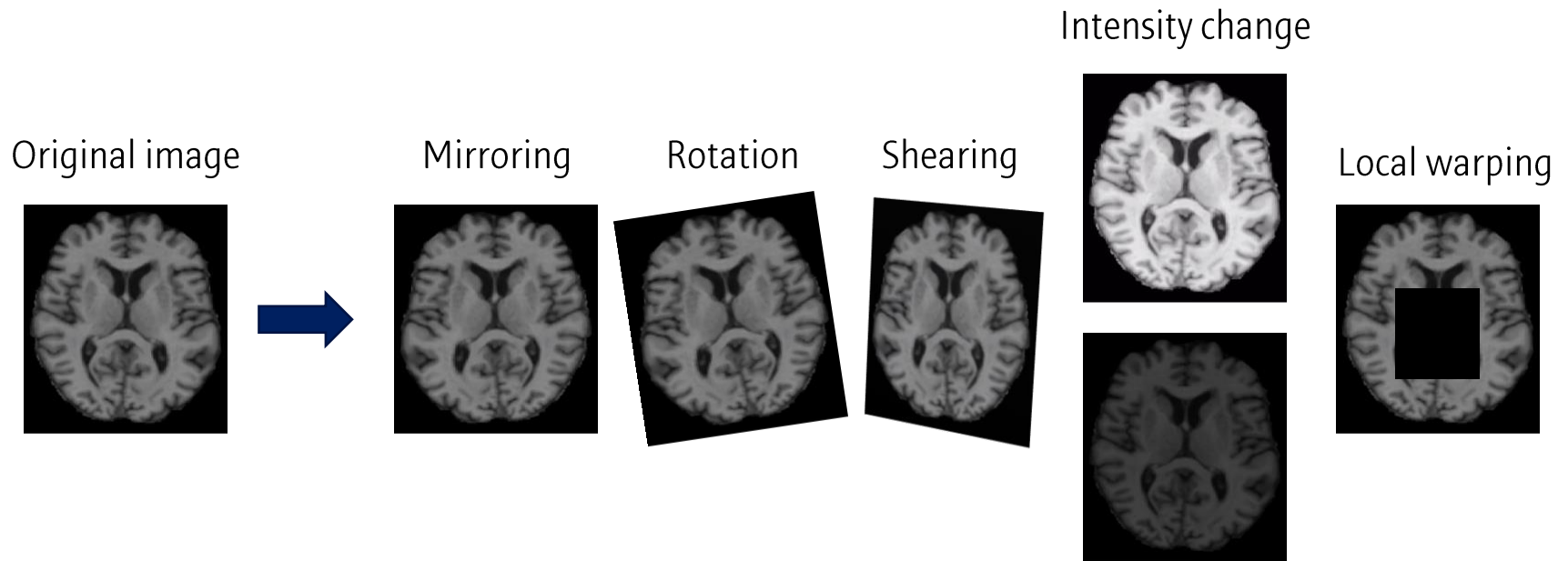
## Four types of fine tuning



# 5. Data Augmentation

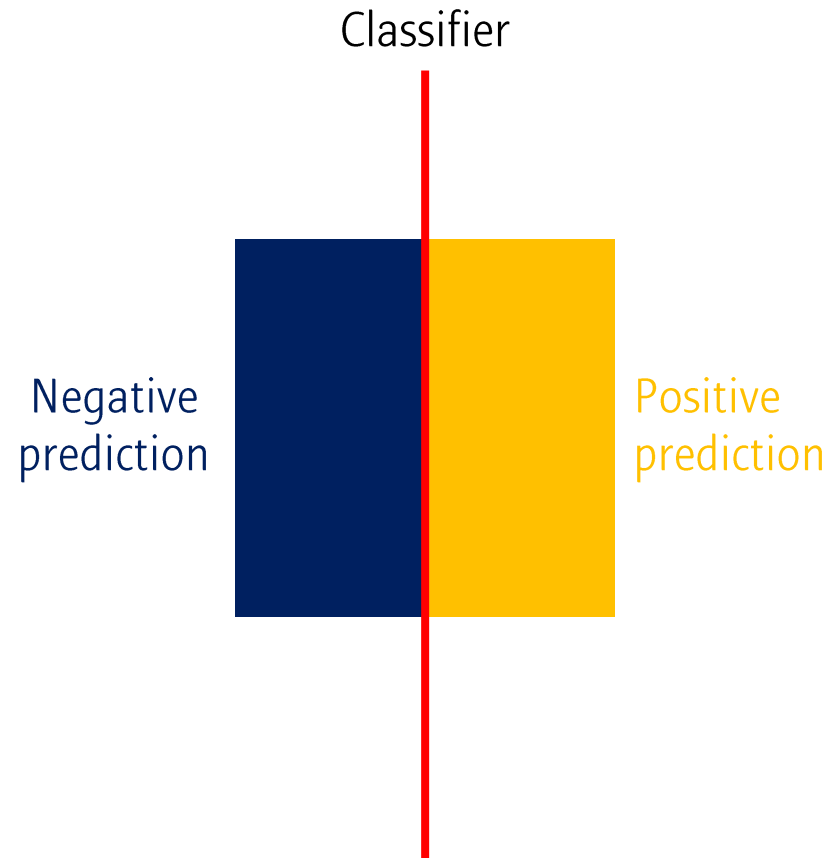
## Diverse data augmentation

- Small sample → Large dataset



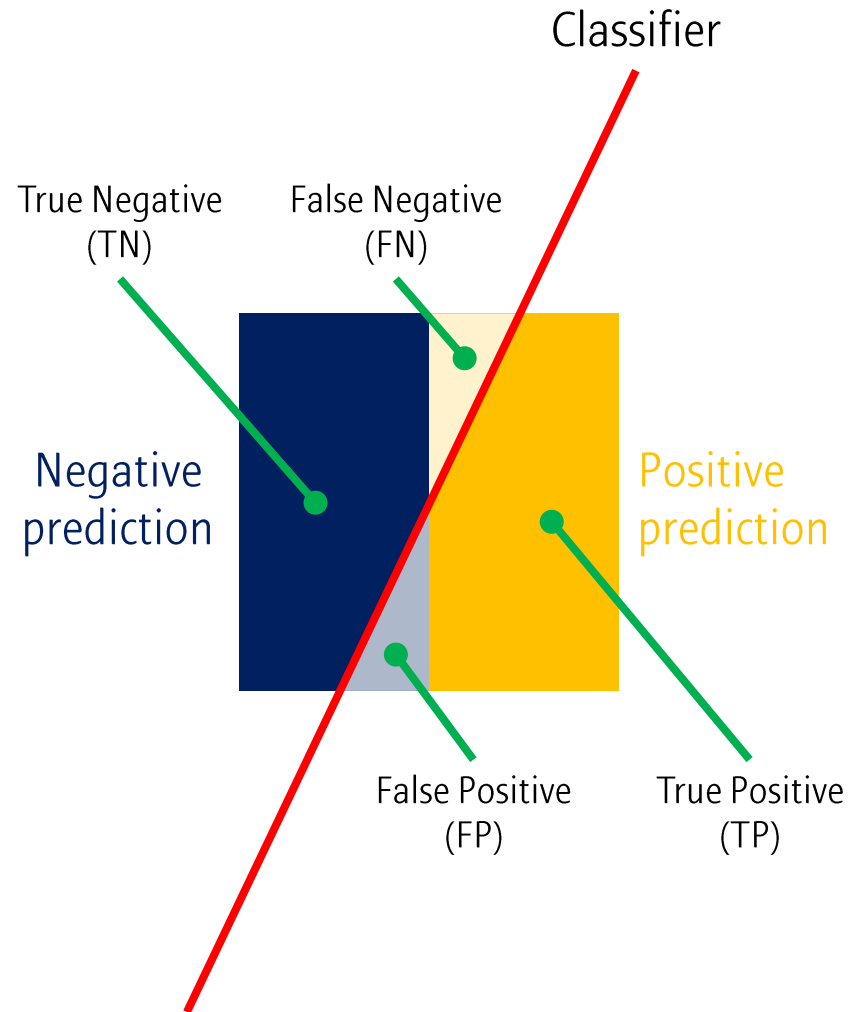
# 6. Evaluation of classification model

Classification model (Ideal)



# 6. Evaluation of classification model

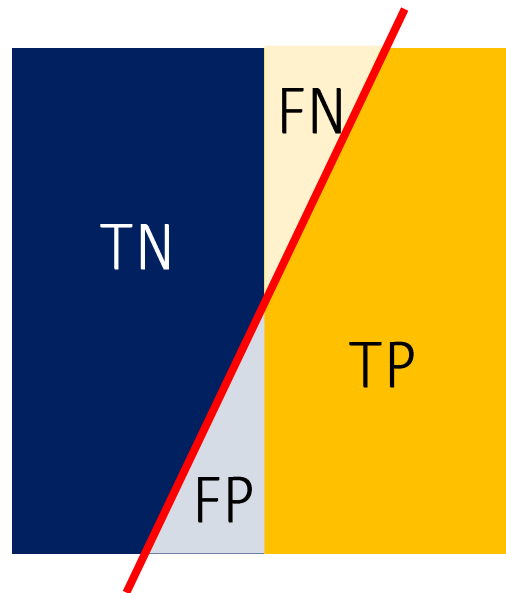
Classification model (Real)



# 6. Evaluation of classification model

Measures: Accuracy

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

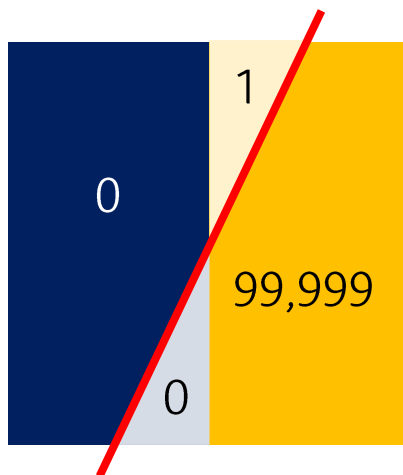




# 6. Evaluation of classification model

## Problem of accuracy

10,000명 중 1명 암 (Positive) 환자일 때, 모델이 전부 Normal (Negative) 이라고 판정하면,



$$Accuracy = 99.999\%$$

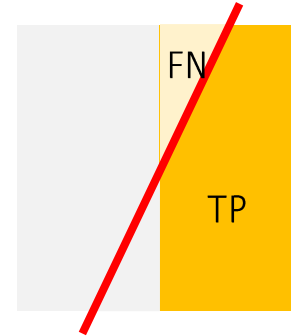
Accuracy는 높지만, 좋은 모델이라고 할 수 없다!

# 6. Evaluation of classification model

Other Measures: Sensitivity, Specificity, Precision

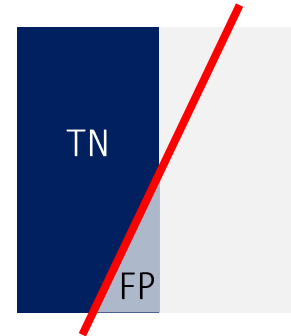
Sensitivity (Recall)

$$\text{Sensitivity} = \frac{TP}{TP + FN}$$



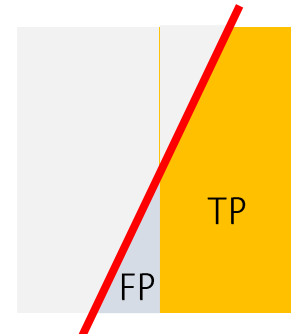
Specificity

$$\text{Specificity} = \frac{TN}{TN + FP}$$



Precision

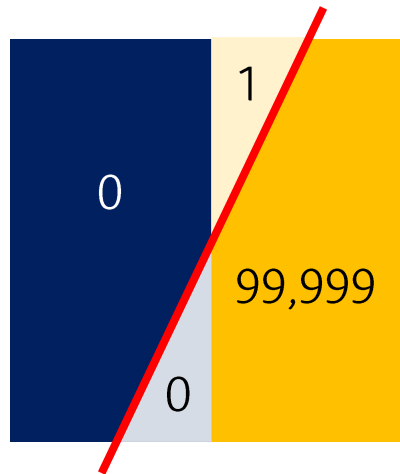
$$\text{Precision} = \frac{TP}{TP + FP}$$



# 6. Evaluation of classification model

## Problem of accuracy solved

10,000명 중 1명 암 (Positive) 환자일 때, 모델이 전부 Normal (Negative) 이라고 판정하면,



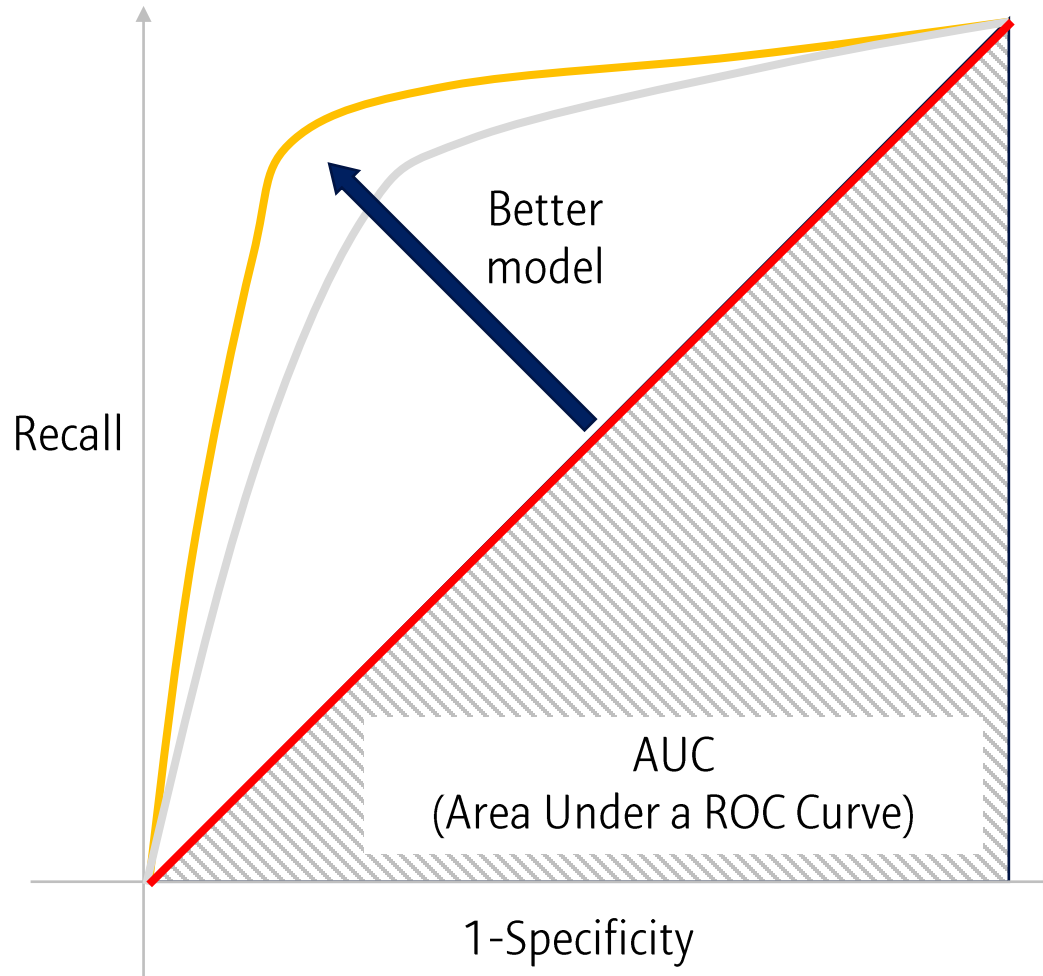
$$Accuracy = 99.999\%$$

$$Sensitivity = \frac{1}{99999}$$

$$Precision = \frac{0}{99999}$$

# 6. Evaluation of classification model

ROC (Receiver Operating Characteristics) curve

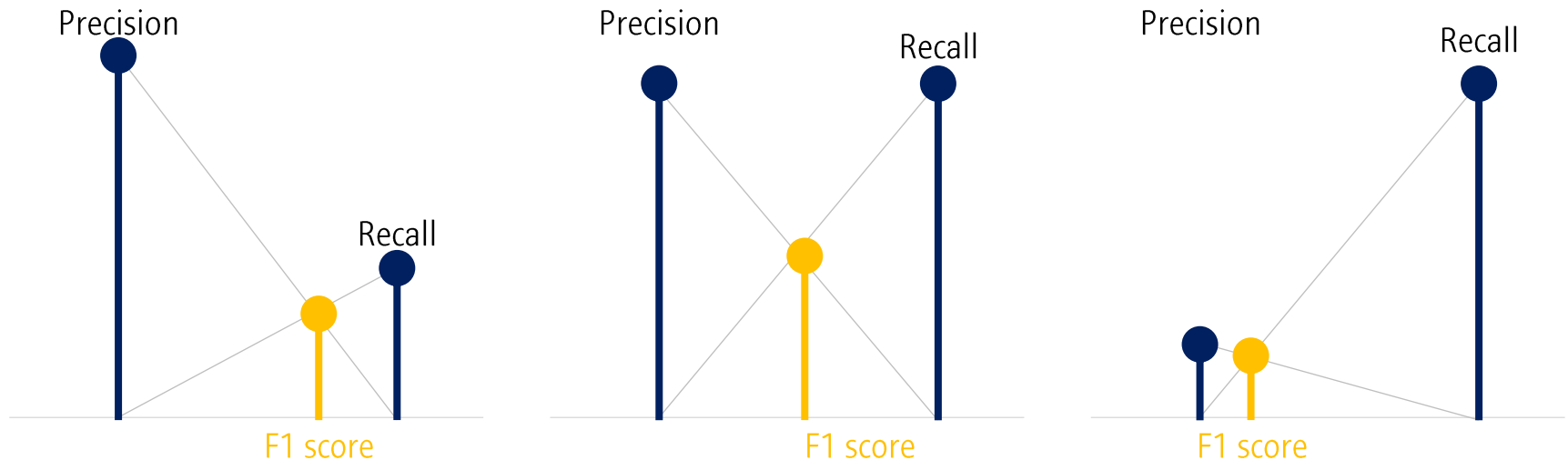
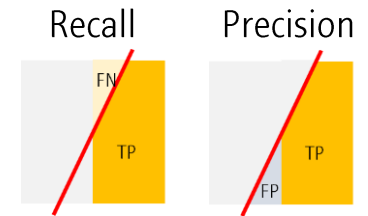


# 6. Evaluation of classification model

## F1 score

F1 score: Harmonic mean of recall and precision

$$F1\ score = 2 \times \frac{Precision \times Recall}{Precision + Recall}$$



Precision과 Recall이 모두 높아야 F1 score가 높음  
Frequently used for imbalanced dataset

## 7. Evaluation of classification model (Multi-class)

Confusion matrix

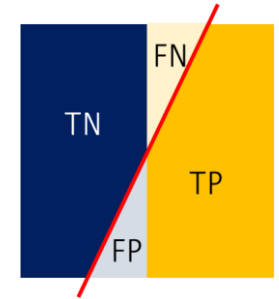
		Prediction			
True		A	B	C	D
	A	10	1	0	0
	B	1	20	10	1
	C	4	0	24	8
	D	0	4	1	16

$$Accuracy = \frac{10 + 20 + 24 + 16}{100} = 70\%$$

# 7. Evaluation of classification model (Multi-class)

F1 score for multi-label

A인 것을 A라고 했느냐,  
A가 아닌 것을 A가 아니라고 했느냐



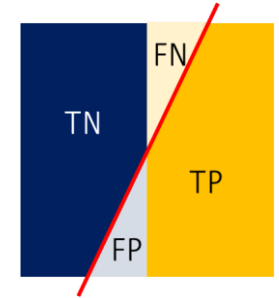
Prediction

True		A	B	C	D
	A	10	1	0	0
	B	1	20	10	1
	C	4	0	24	8
	D	0	4	1	16

# 7. Evaluation of classification model (Multi-class)

F1 score for multi-label

B인 것을 B라고 했느냐,  
B가 아닌 것을 B가 아니라고 했느냐



Prediction

True		A	B	C	D
	A	10	1	0	0
	B	1	20	10	1
	C	4	0	24	8
	D	0	4	1	16