

SC201

Lecture 5

data

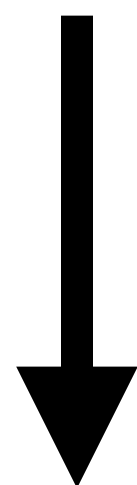


data

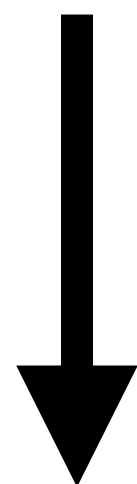


Train

data

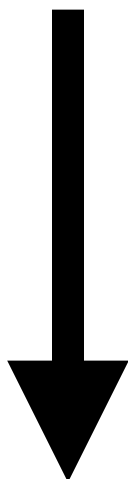


Train

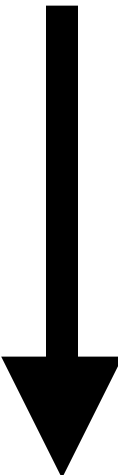


Test

data

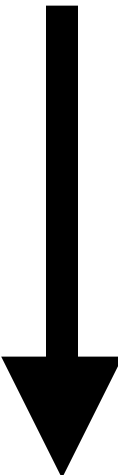


data



Training Data

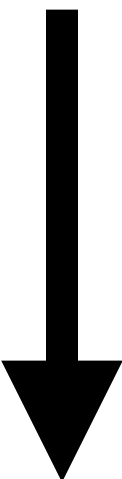
data



Training Data

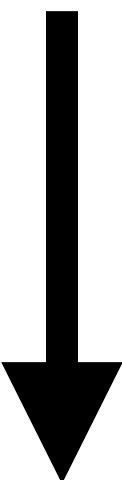
Validation Data

data



Training Data

Validation Data



Test

Cross Validation

data



data



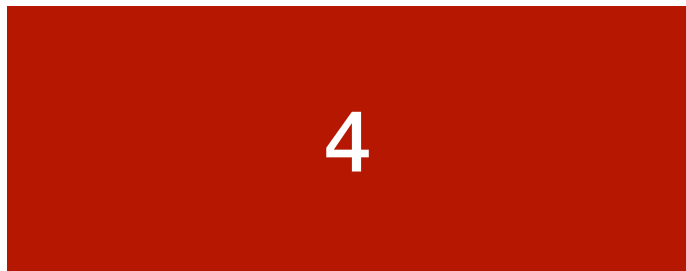
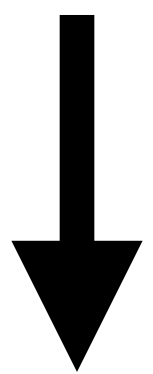
k folds



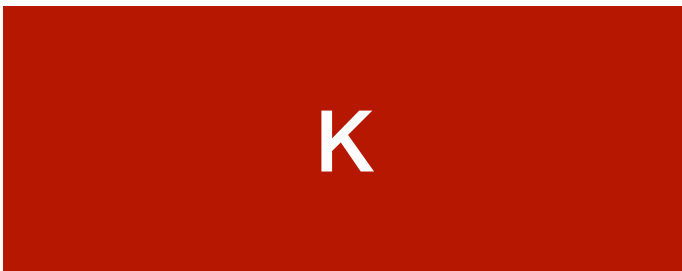
data



k folds



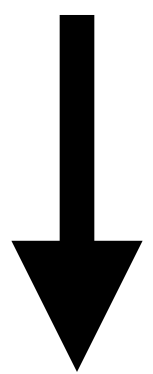
... ..



data



k folds



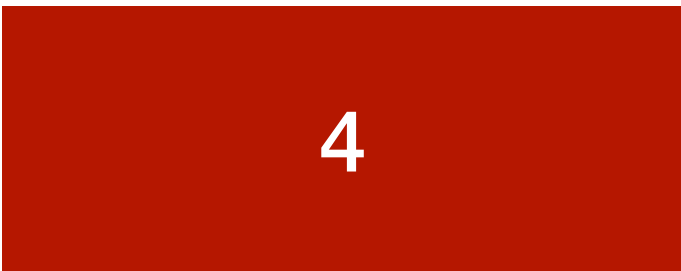
D_{val}



D_{train}

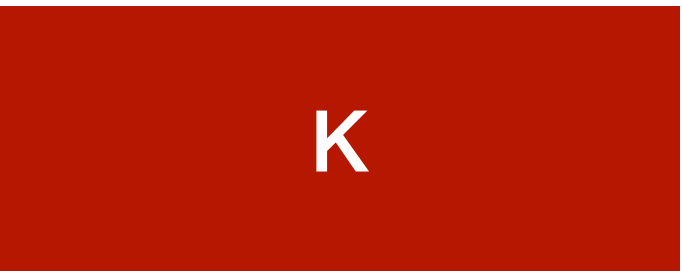


D_{train}



D_{train}

...

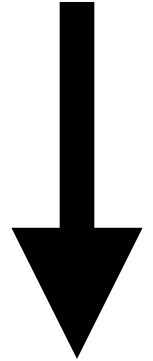


D_{train}

data



k folds



D_{train}



D_{val}

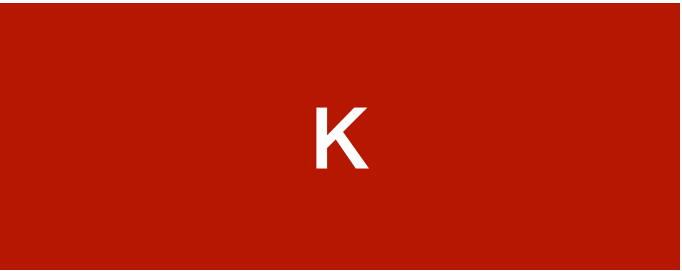


D_{train}



D_{train}

... ..

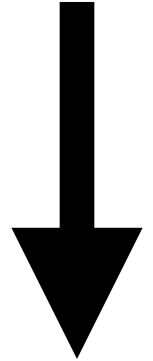


D_{train}

data



k folds



D_{train}



D_{train}

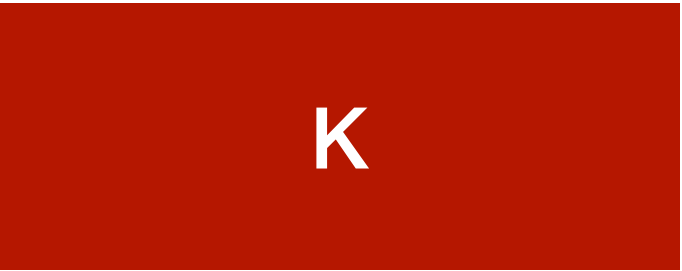


D_{val}



D_{train}

...

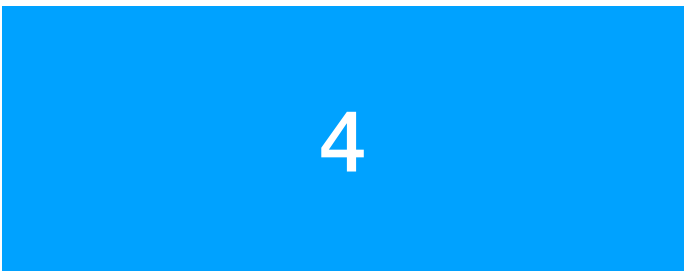
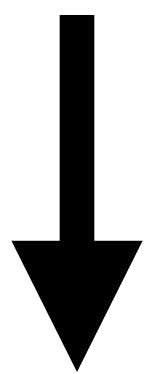


D_{train}

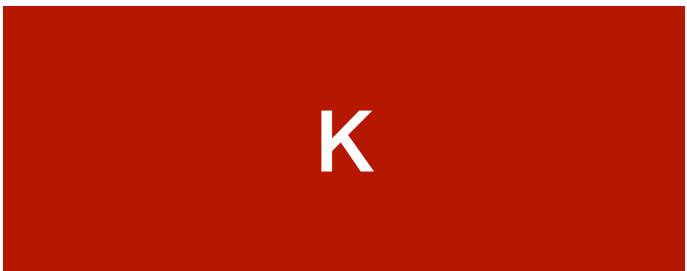
data



k folds



...



D_{train}

D_{train}

D_{train}

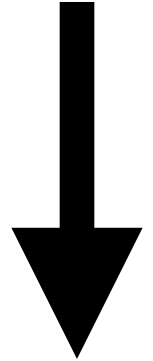
D_{val}

D_{train}

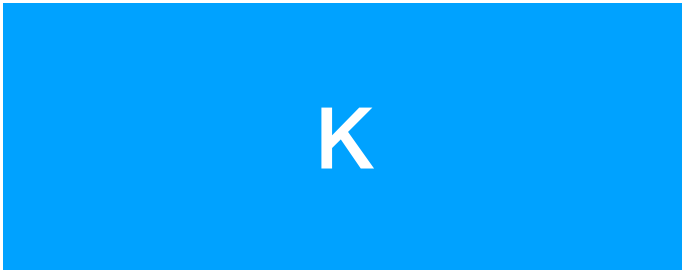
data



k folds



...



D_{train}

D_{train}

D_{train}

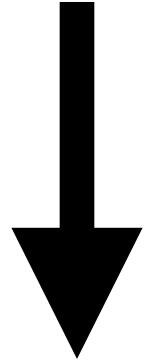
D_{train}

D_{val}

data



k folds



D_{val}



D_{train}

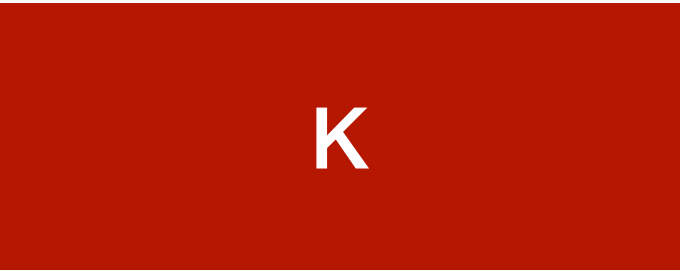


D_{train}



D_{train}

...

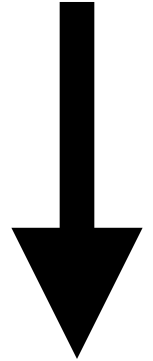


D_{train}

data



k folds



D_{train}



D_{val}

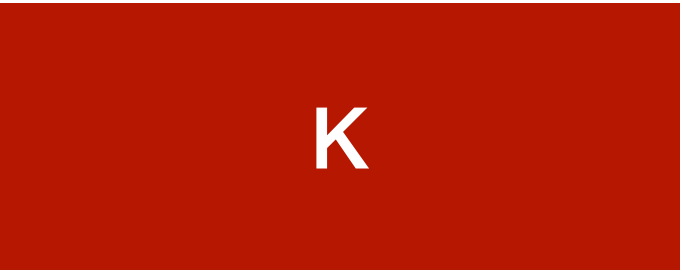


D_{train}



D_{train}

... ..

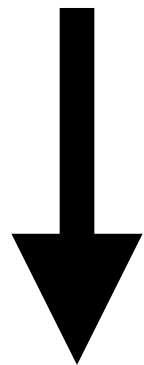


D_{train}

data



k folds



D_{train}



D_{train}

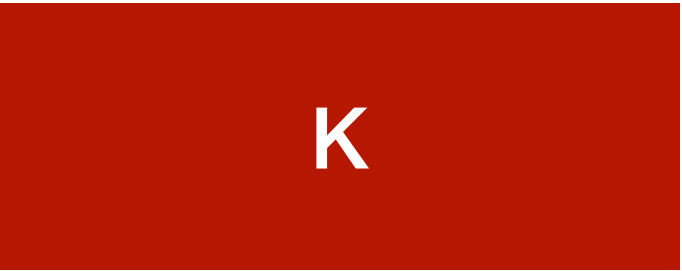


D_{val}



D_{train}

... ..

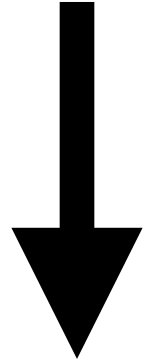


D_{train}

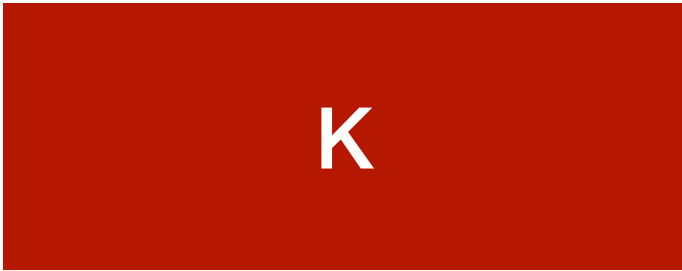
data



k folds



...



D_{train}

D_{train}

D_{train}

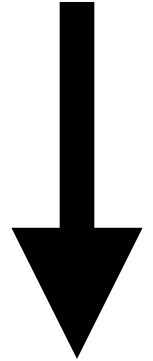
D_{val}

D_{train}

data



k folds



D_{train}



D_{train}

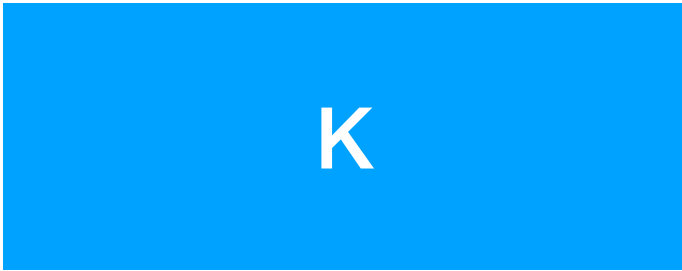


D_{train}



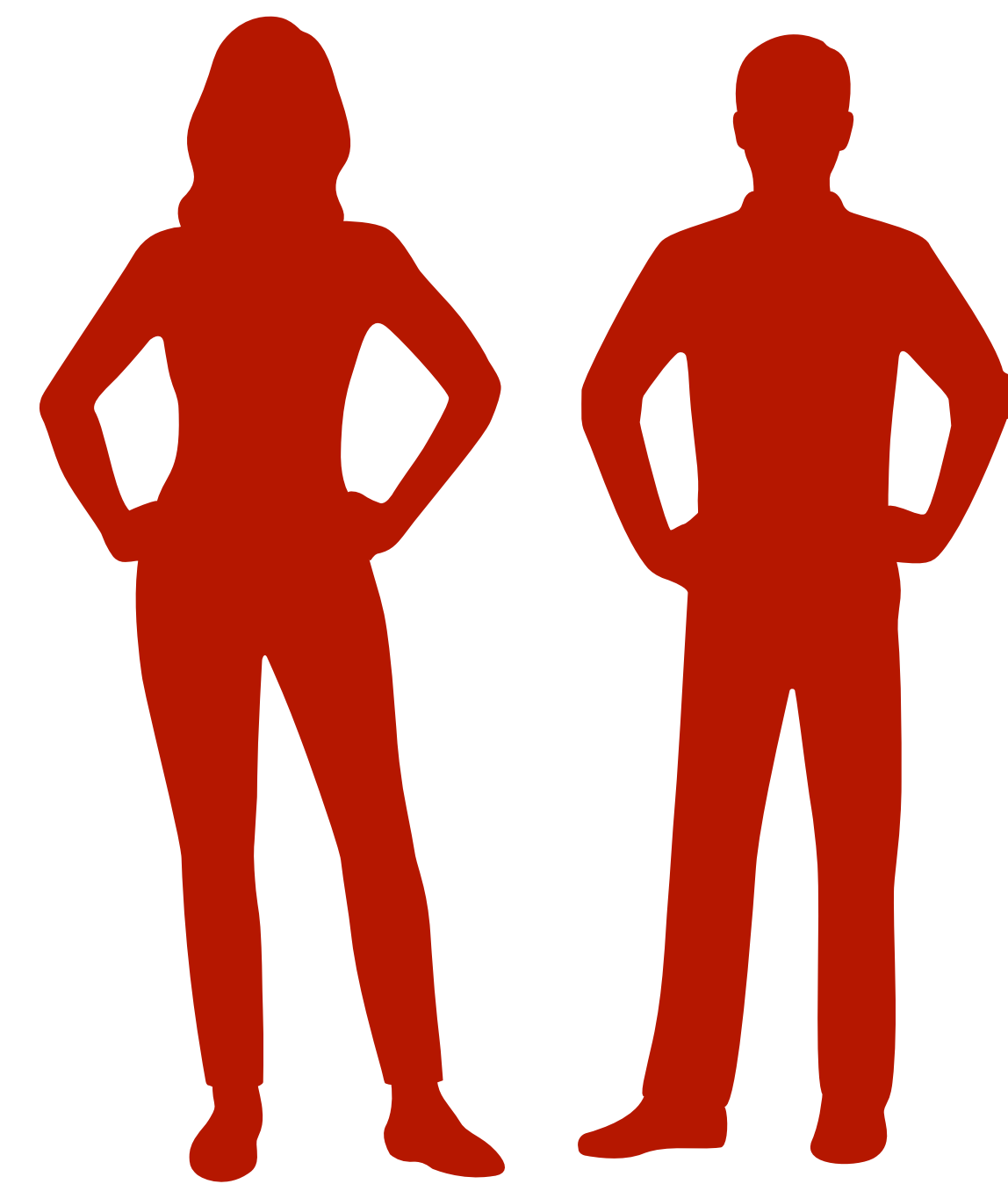
D_{train}

...



D_{val}

房地產公司要您建造一個「房價預測模型」！
如果您的模型估價比市價高，沒有顧客會想要跟你們家買房
但如果估價比市價低，公司利潤又會大受影響
因此最好的辦法就是讓模型預測出來的結果「比正確數值高一點」
這樣還可以有降價的空間（如果比市價低，買走了，就沒有了...）
假設 y_i 是 True Label, h_i 是模型預測的結果
請問您要怎麼調整 Loss Function (L)
(對 $L = (h_i - y_i)^2$ 做調整)
讓模型 train 完之後產出的結果就會比正確數值高？



Suppose you train a logistic regression classifier and your hypothesis function h is

$$h = \sigma(\theta_0 + \theta_1 x_1 + \theta_2 x_2)$$

where $\theta_0 = 6$; $\theta_1 = 0$; $\theta_2 = -1$

Which of the following figure will represent the decision boundary as given by above classifier?

