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MELBOURNE

Workshop 10

COMP20008

Elements of Data Processing
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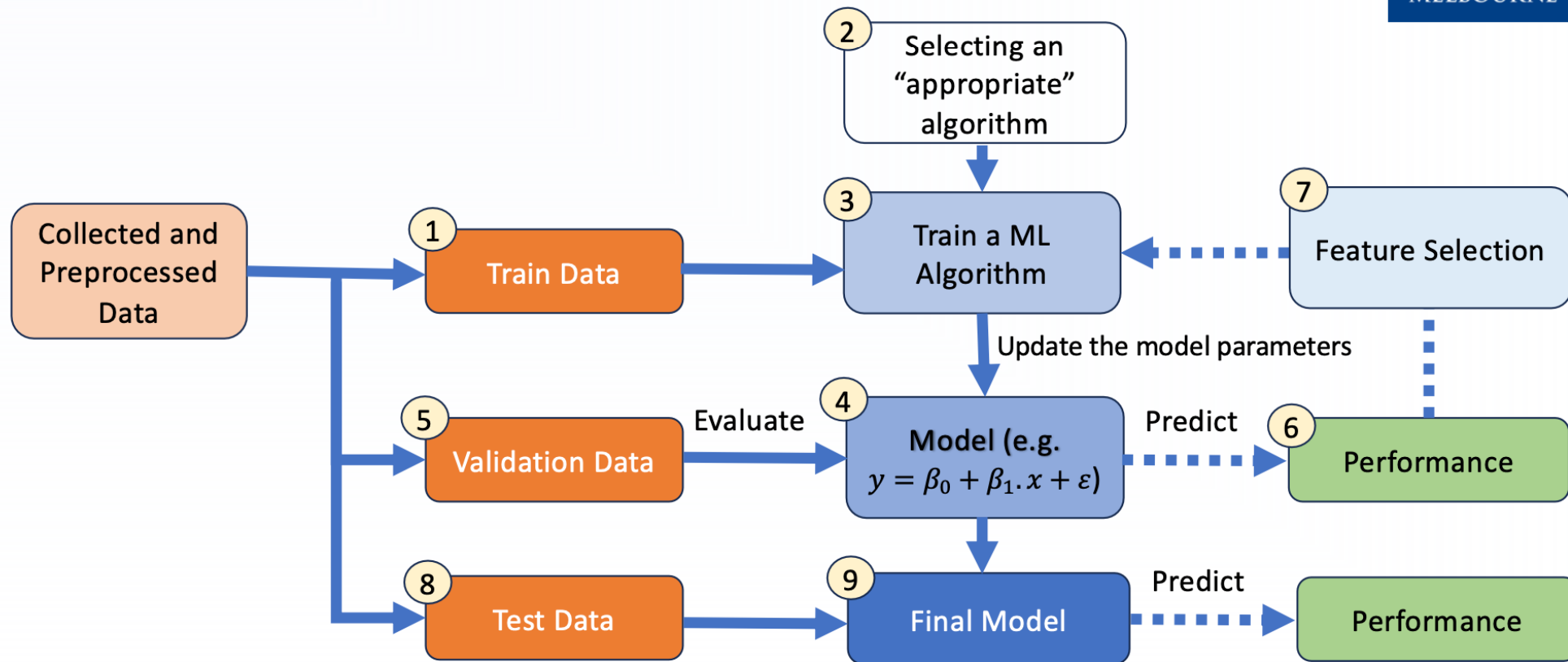


Agenda

- Supervised ML Evaluation
 - Why train/validation/test split
 - Holdout, Cross-validation & Stratification
 - Bootstrapping
- Performance metrics

Why train/validation/test split

Supervised Models Pipeline

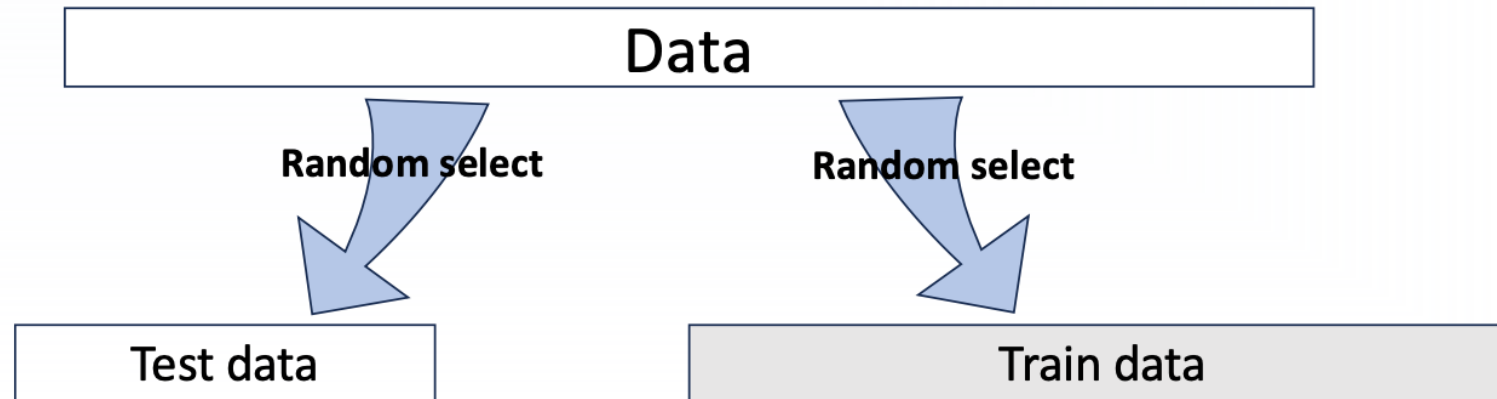


Evaluation methods

- We want to quantitatively determine how well our model performs on unseen data
- We split data into three parts to prevent data leakage
 - Model learns from training data
 - We tune hyperparameters on validation data
 - Fit tuned model to combined train & validation data, test its performance on testing data

Evaluation methods

- Three methods
 - Holdout: Hold out a fixed proportion of data as testing set



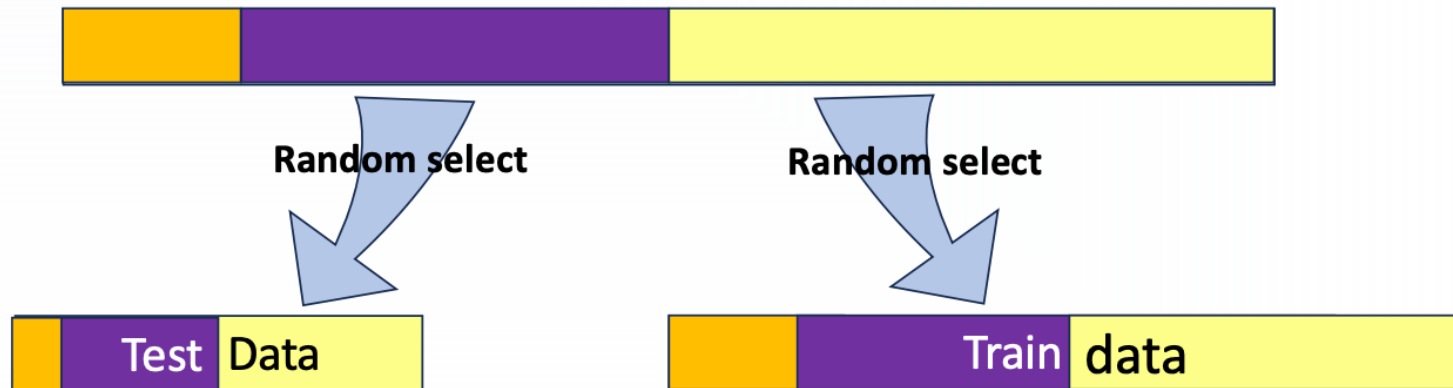
Evaluation methods

- Three methods
 - m -fold Cross-validation: Split the data into m folds, train on $m-1$, test on 1, repeat for all folds to and evaluate average performance



Evaluation methods

- Three methods
 - Stratification: preserves class distribution in training and testing set



Bootstrapping

- An additional method to evaluate model
 - Randomly draw sample from data set with replacement to obtained bootstrapped sample
 - Train model on bootstrapped sample, evaluate performance on corresponding test sample (data not seen in bootstrapped set)
 - Repeat many times to get average performance results

Performance metrics

- Some metrics for classification:

- Accuracy $\frac{TP+TN}{P+N}$
- Precision $\frac{TP}{TP+FP}$
- Recall $\frac{TP}{TP+FN}$
- F1-score $\frac{2*precision*recall}{precision+recall}$

Total =
P+N

Predicted (\hat{Y})

Actual (Y)

		Predicted (\hat{Y})	
		$\hat{Y} = +1$	$\hat{Y} = -1$
Actual (Y)	Y = P	TP	FN
	Y = N	FP	TN



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Thank you

More Resources: Canvas