

S6-ESE-AI

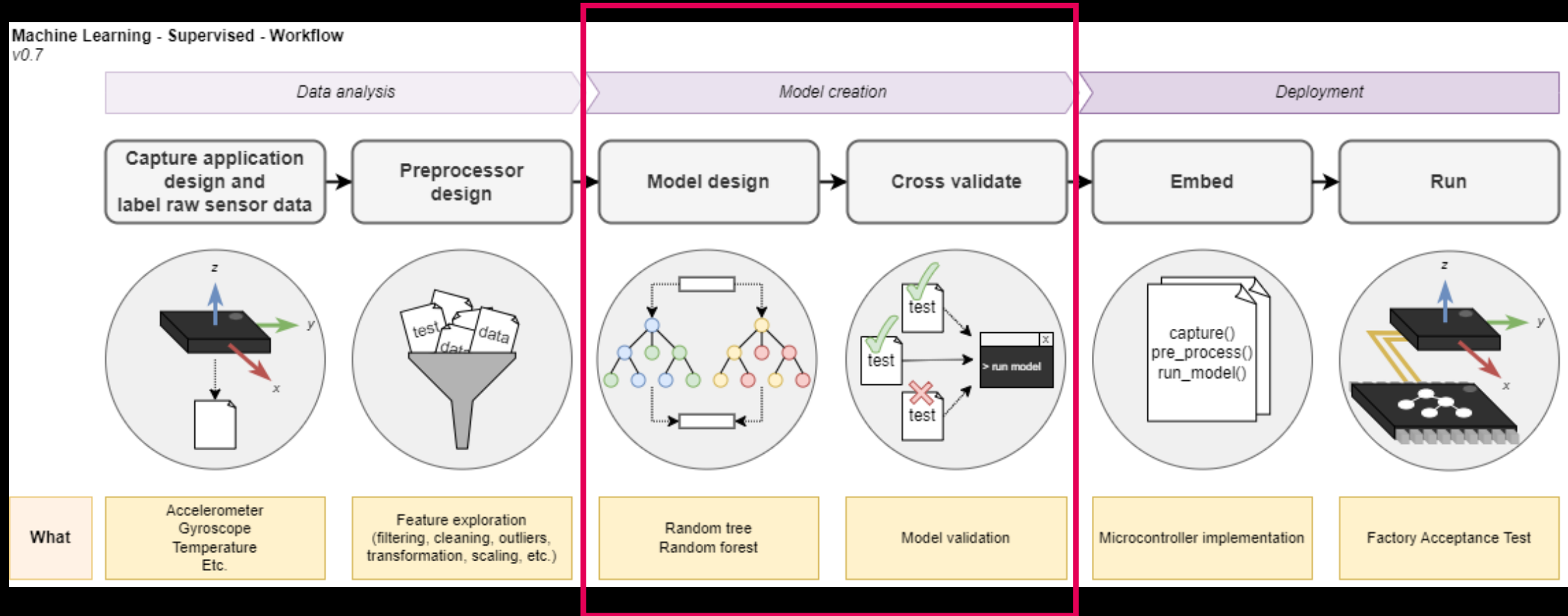
PERFORMANCE

JEROEN VEEN
HUGO ARENDS

WORKFLOW

WORKFLOW

Machine Learning - Supervised - Workflow
v0.7

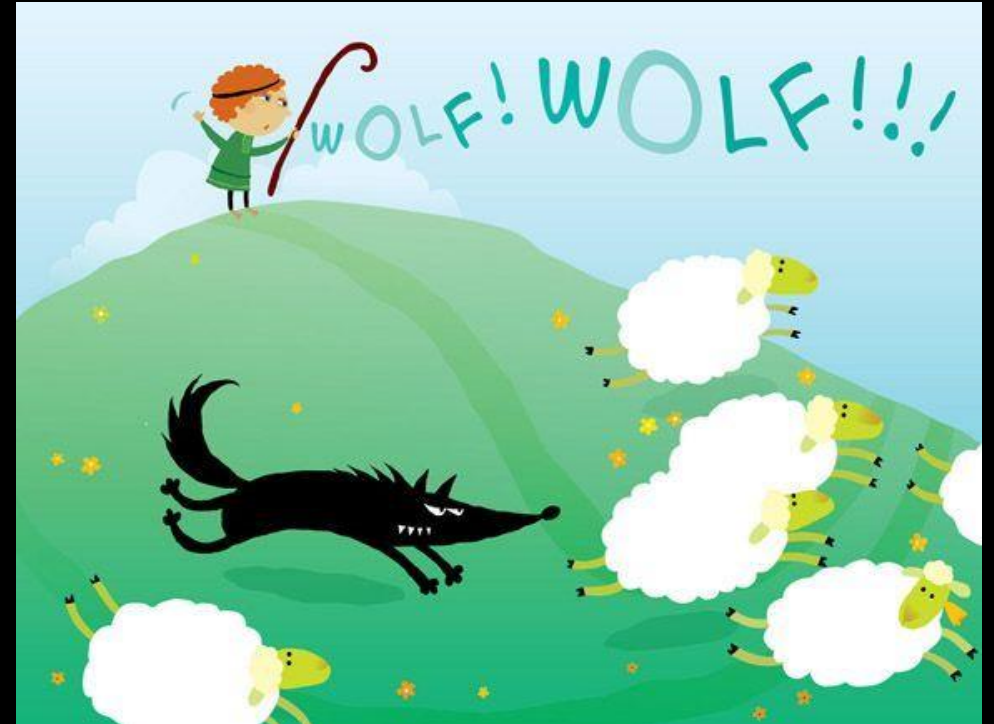


AGENDA

- Confusion matrix
- Accuracy
- Cross-validation
- Classification report
- Generalization

THE BOY WHO CRIED WOLF

- "Wolf" is a positive class.
- "No wolf" is a negative class
- An Aesop's Fable ~620 BCE









Source: Sam Taplin

CONFUSION MATRIX

		ACTUAL	
PREDICTED		True Positive (TP) Reality: A wolf threatened. Shepherd said: "Wolf." Outcome: Shepherd is a hero.	False Positive (FP) Reality: No wolf threatened. Shepherd said: "Wolf." Outcome: Villagers are angry at shepherd for waking them up (Type I error)
		False Negative (FN) Reality: A wolf threatened. Shepherd said: "No wolf." Outcome: The wolf ate all the sheep (Type II error)	True Negative (TN) Reality: No wolf threatened. Shepherd said: "No wolf." Outcome: Everyone is fine.

CONFUSION MATRIX

Actual (what the data says)	Predicted (what our model says))				
	CLASSES	 A	 B	 C	Row totals
	 A	5	2	3	10
	 B	2	6	0	8
	 C	3	2	2	7
	Column Totals	10	10	5	25

Diagonal numbers are rightly classified observations

Total number of observations/ records

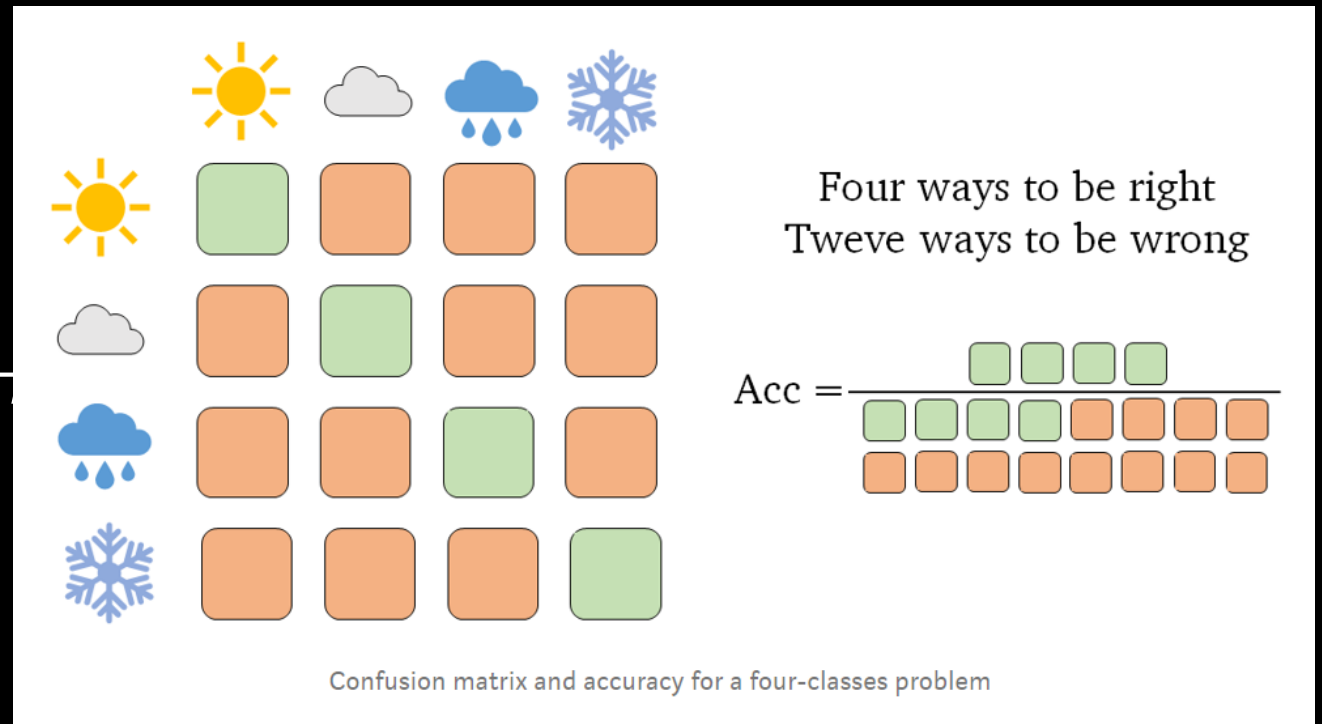
#MLmuse
CLAIRVOYANT

MEASURING PERFORMANCE

- Accuracy

$$= \frac{\text{Number of correct predictions}}{\text{Total number of predictions}}$$

$$= \frac{TP + \dots}{\dots}$$



Source: <https://towardsdatascience.com/the-illustrated-guide-to-classification-metrics-the-basics-cf3c2e9b89b2>

SPLITTING DATA

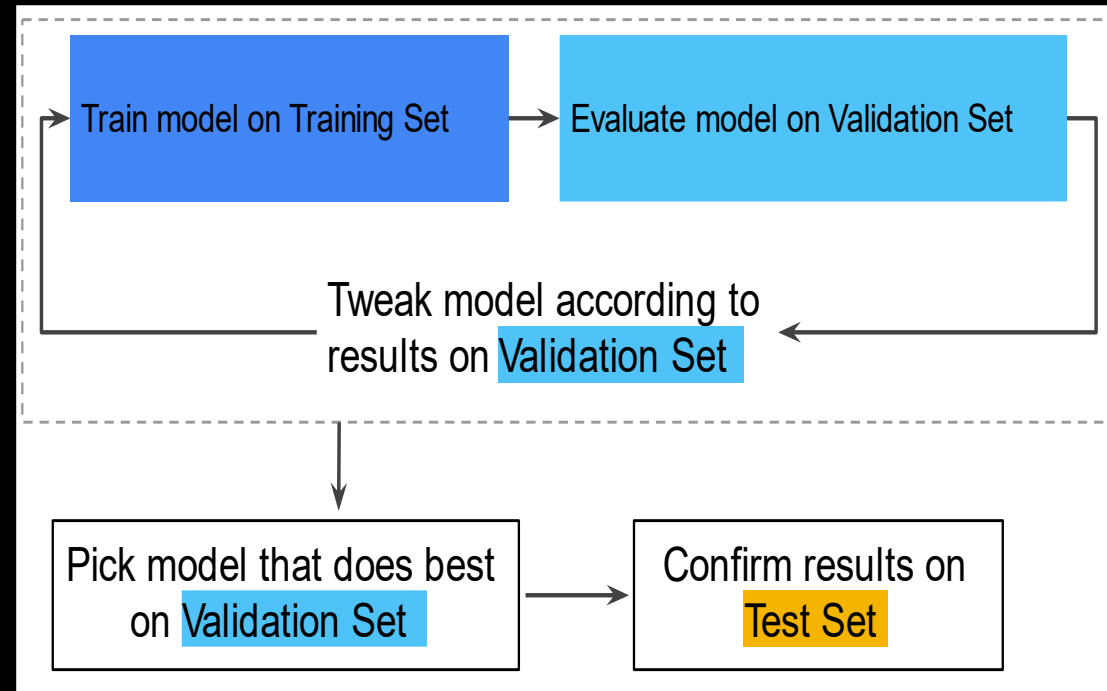
- Slice data into three subsets: Training, validation and test data



- Make sure that your subsets meet the following conditions:
 - Large enough to yield statistically meaningful results.
 - Representative of the data set as a whole.E.g. don't pick a test set with different characteristics than the training set.

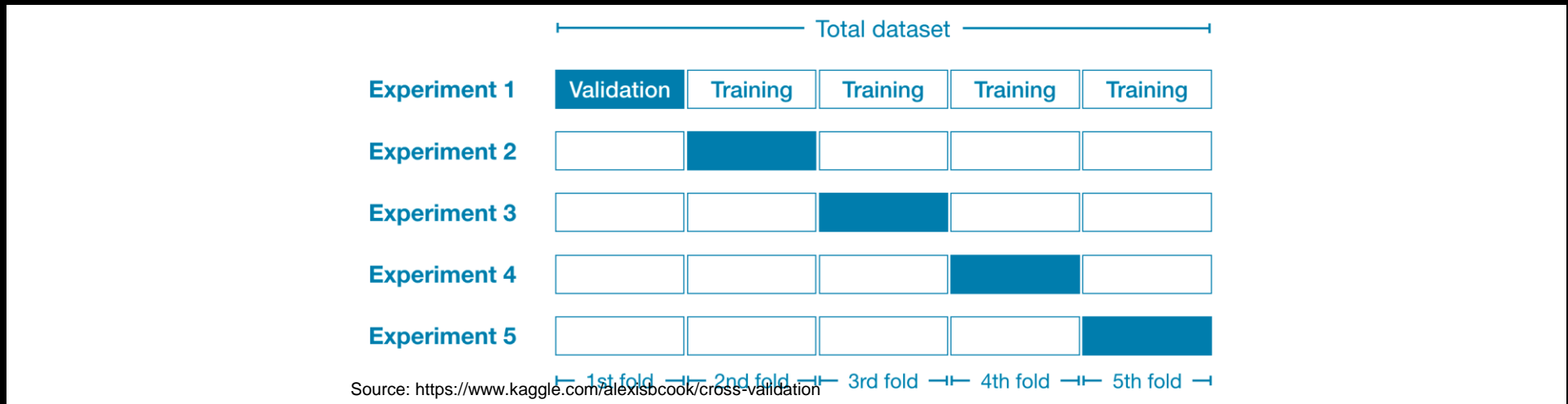
TRAINING, VALIDATION, TESTING

- Never train on test data!



CROSS-VALIDATION

- Estimate of a model's generalization performance
- Break the data into folds



- For small datasets, where extra computational burden isn't a big deal, you should run cross-validation.

SKLEARN CLASSIFICATION REPORT

```
>>> from sklearn.metrics import classification_report
>>> y_true = [0, 1, 2, 2, 2]
>>> y_pred = [0, 0, 2, 2, 1]
>>> target_names = ['class 0', 'class 1', 'class 2']
>>> print(classification_report(y_true, y_pred, target_names=target_names))
```

	precision	recall	f1-score	support
class 0	0.50	1.00	0.67	1
class 1	0.00	0.00	0.00	1
class 2	1.00	0.67	0.80	3
accuracy			0.60	5
macro avg	0.50	0.56	0.49	5
weighted avg	0.70	0.60	0.61	5

```
>>> y_pred = [1, 1, 0]
>>> y_true = [1, 1, 1]
>>> print(classification_report(y_true, y_pred, labels=[1, 2, 3]))
```

	precision	recall	f1-score	support
1	1.00	0.67	0.80	3
2	0.00	0.00	0.00	0
3	0.00	0.00	0.00	0
micro avg	1.00	0.67	0.80	3
macro avg	0.33	0.22	0.27	3
weighted avg	1.00	0.67	0.80	3

GENERALIZATION