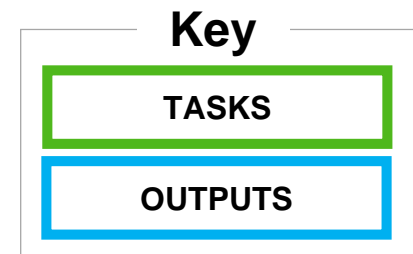
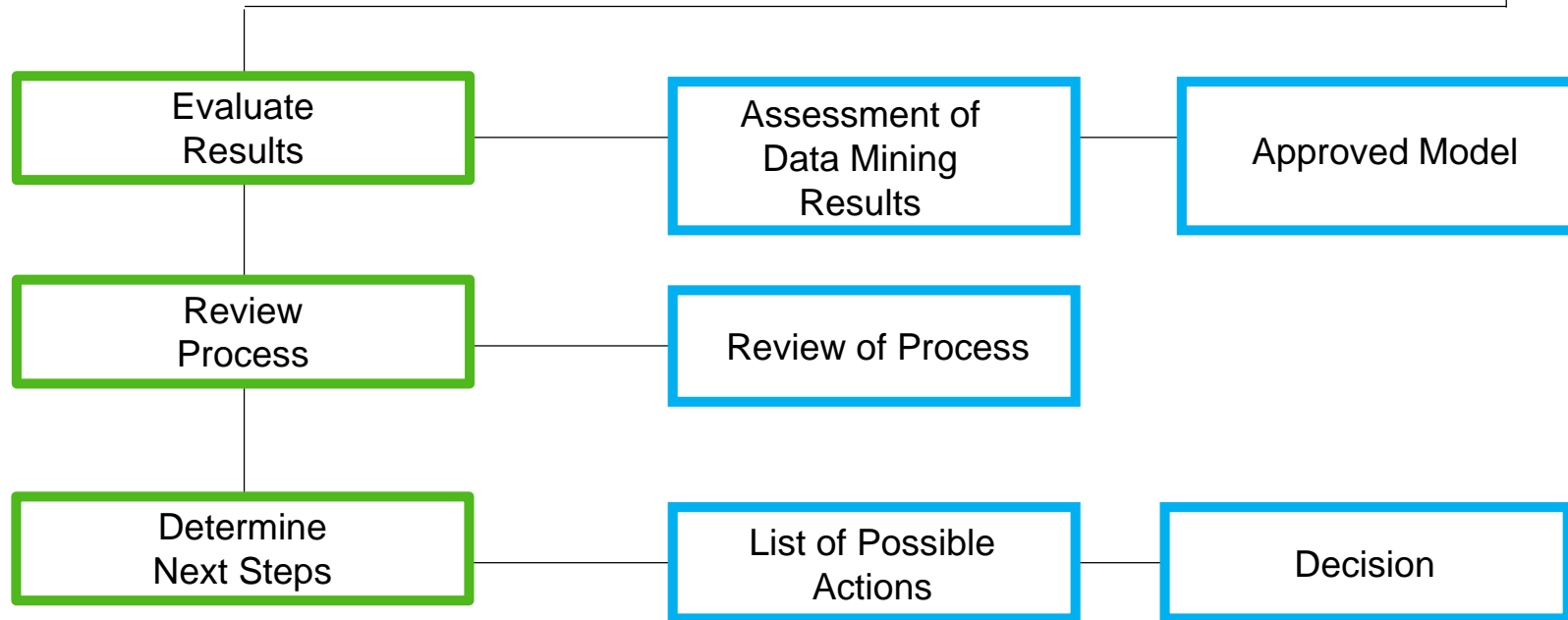


Week 5 Unit 1: Evaluation Phase – Overview



Evaluation Phase – Overview

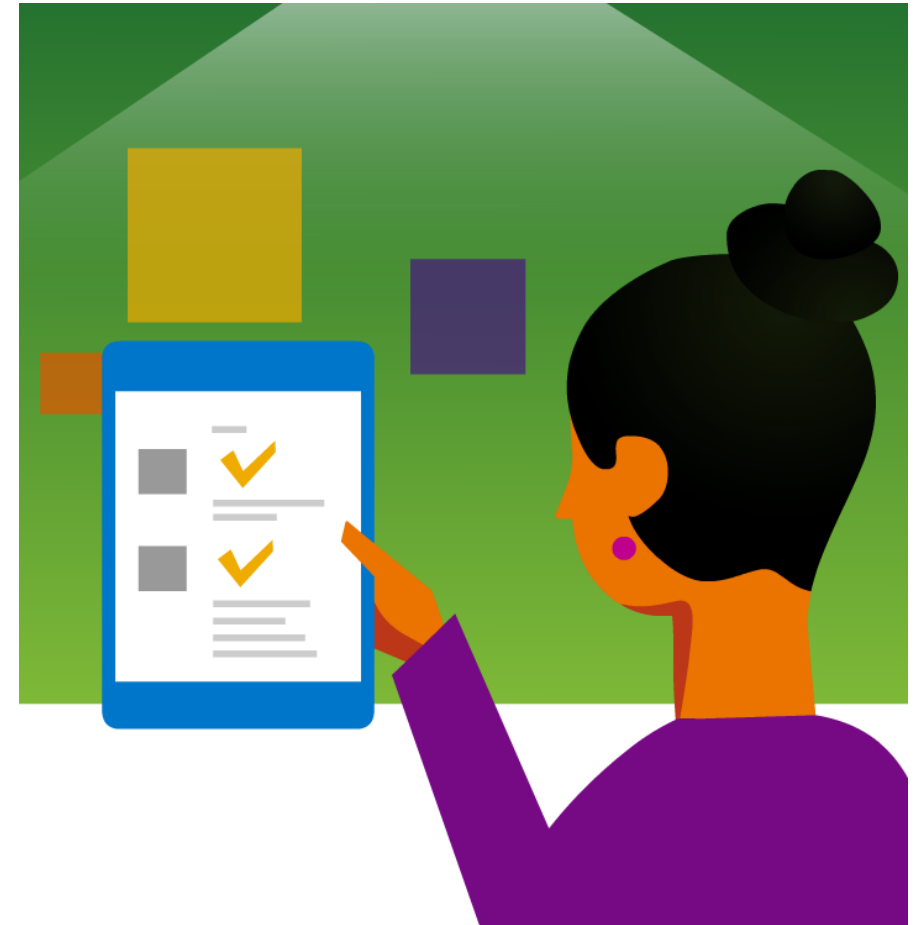
CRISP-DM – Phase 5: Evaluation



Evaluation Phase – Overview

Phase 5.1: Evaluate Results

- **Task**
 - Assess the degree to which the model meets the business objectives.
 - Test the model(s) on test applications if time and budget constraints permit.
- **Output – Assessment of Data Mining Results with respect to business success criteria**
- **Output – Approved Model**



Evaluation Phase – Overview

Phase 5.2: Review Process

- **Task**
 - Conduct a more thorough review of the data mining engagement to determine if there is any important factor or task that has somehow been overlooked.
 - Identify any quality assurance issues.
- **Output – Review of Process**
 - Summarize the process review and highlight activities that have been missed and/or should be repeated.



Evaluation Phase – Overview

Phase 5.3: Determine Next Steps

- **Task**
 - Assess how to proceed with the project.
- **Output – List of Possible Actions**
 - List the potential further actions along with the reasons for and against each option.
- **Output – Decision**
 - Describe the decision on how to proceed.





Thank you

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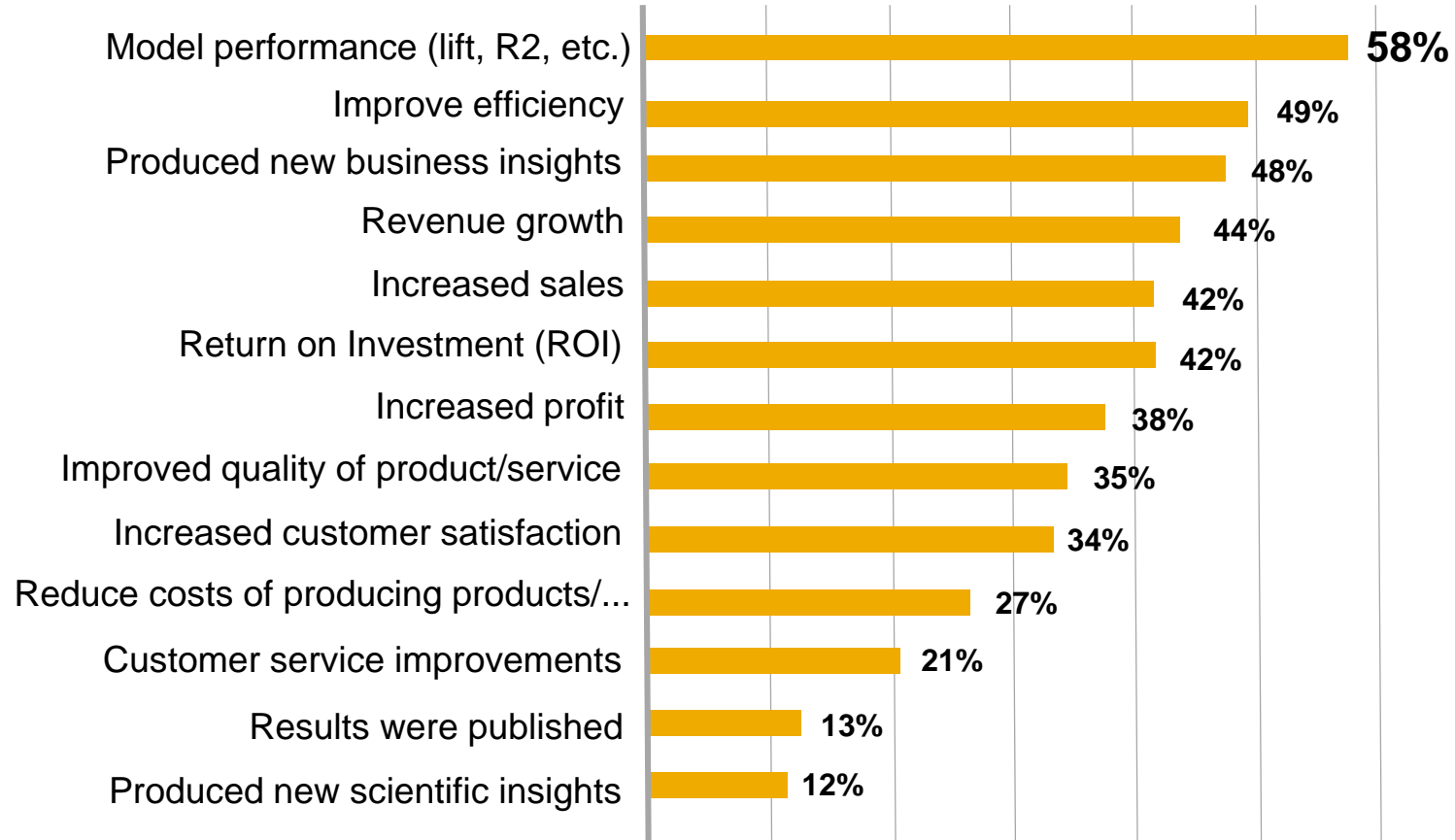
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Week 5 Unit 2: Model Performance Metrics



Model Performance Metrics

Introduction

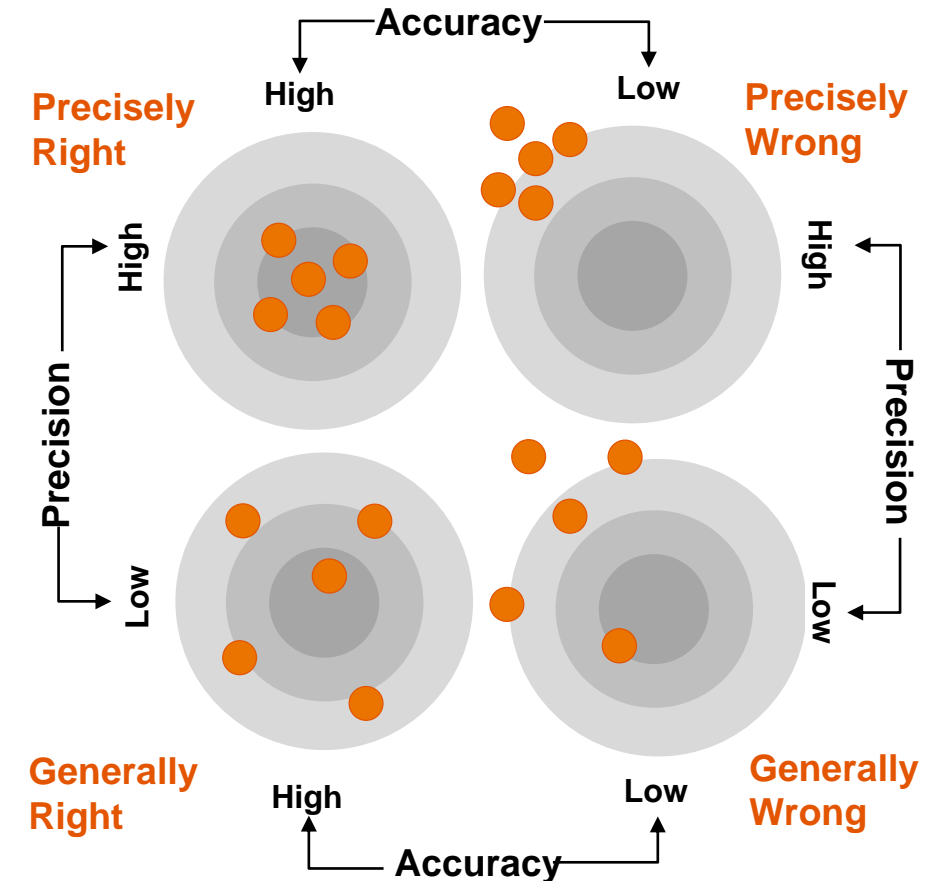


From Rexer Analytics 3rd Annual Data Miner Survey by Karl Rexer, PhD, Heather N. Allen, PhD and Paul Gearan

Model Performance Metrics

Success criteria for classification models

- The following performance metrics are often used to assess classification model success:
 - Confusion matrices summarize Type I and Type II errors
 - Lift, gains, ROC, and area under the curve (AUC)
 - SAP has developed predictive power (KI) and prediction confidence (KR) metrics



Model Performance Metrics

Confusion matrix

		Predicted Class		Total
		1	0	
Actual Class	1	TP	FN	P
	0	FP	TN	N

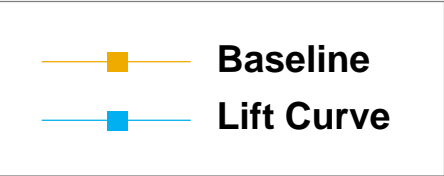
Model Performance Metrics

Confusion matrix: costs and benefits

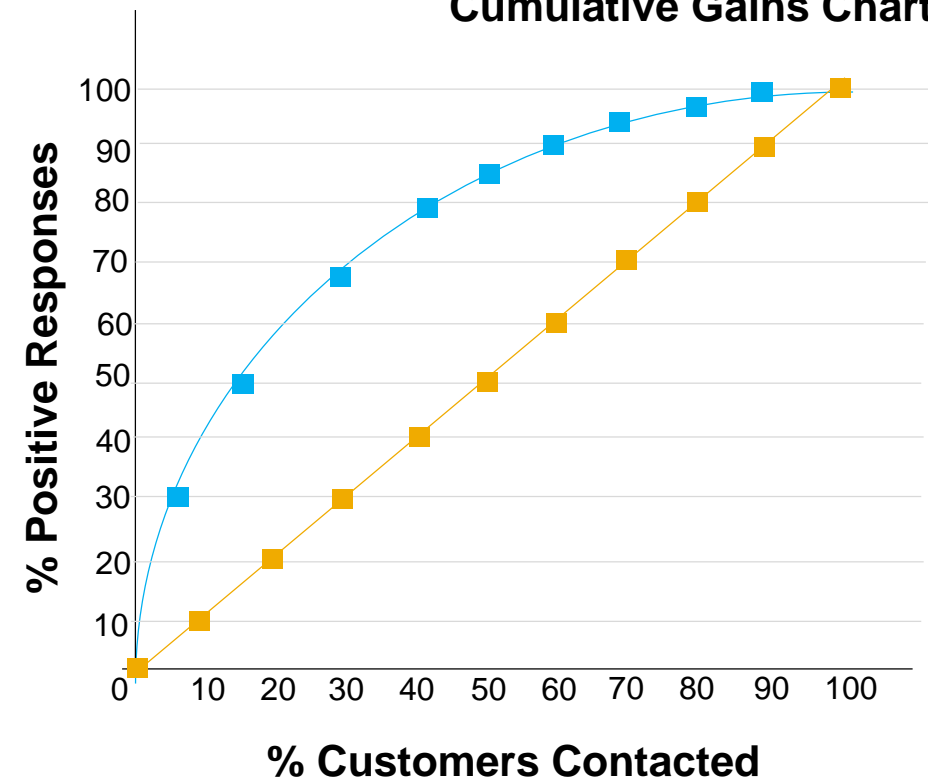
		Predicted		
		Yes	No	
Actual	Yes	+ Benefit Value	– Benefit Value	Actual Total Positive
	No	Cost	N/A	Actual Total Negative
		Total Predicted Positive	Total Predicted Negative	

Model Performance Metrics

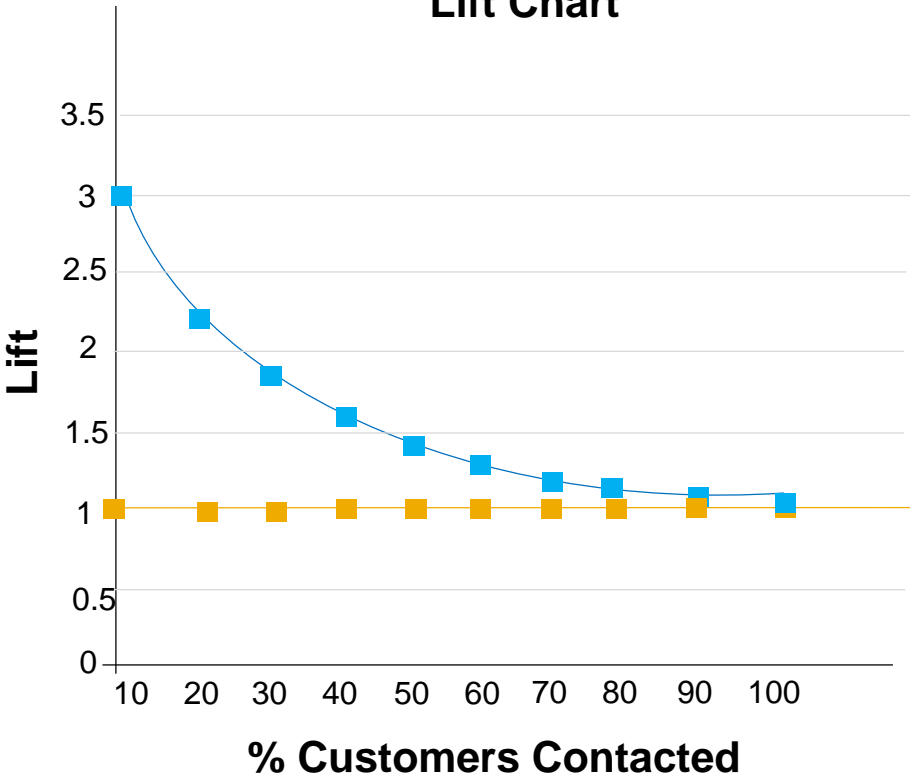
Lift and gains charts



Cumulative Gains Chart



Lift Chart

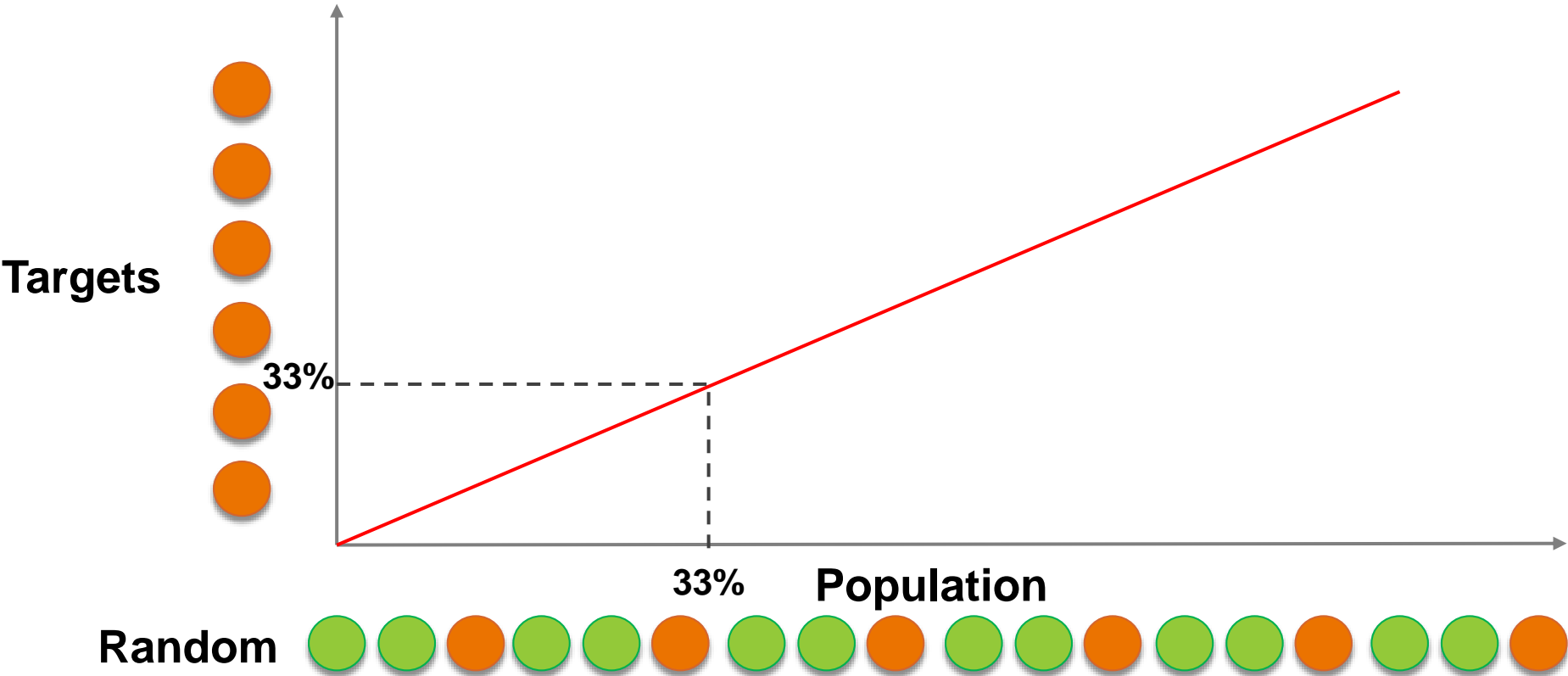


Model Performance Metrics

Gains (detected) chart – Random model

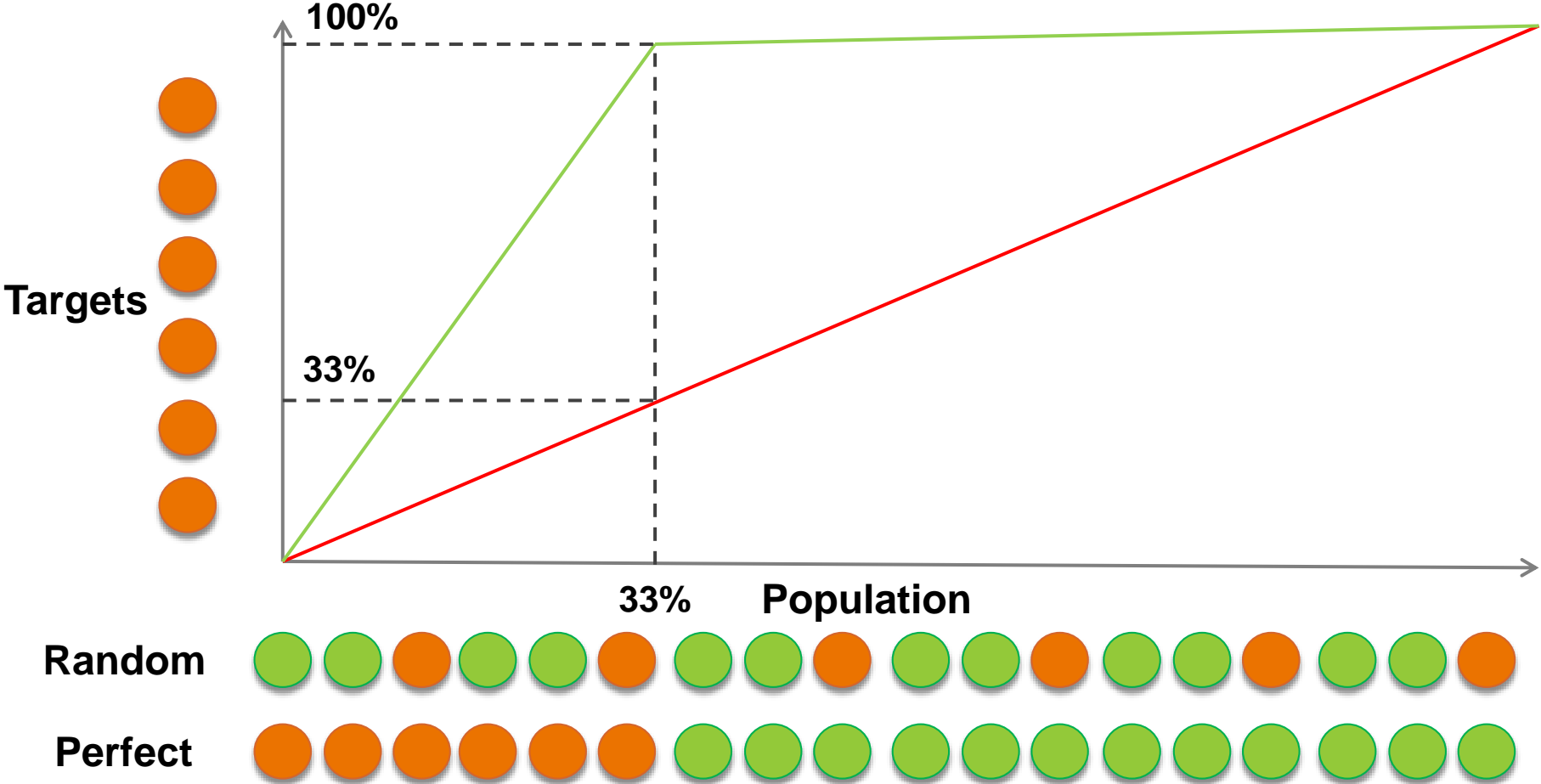
Targets

Non-Targets



Model Performance Metrics

Gains (detected) chart – Perfect model



Model Performance Metrics

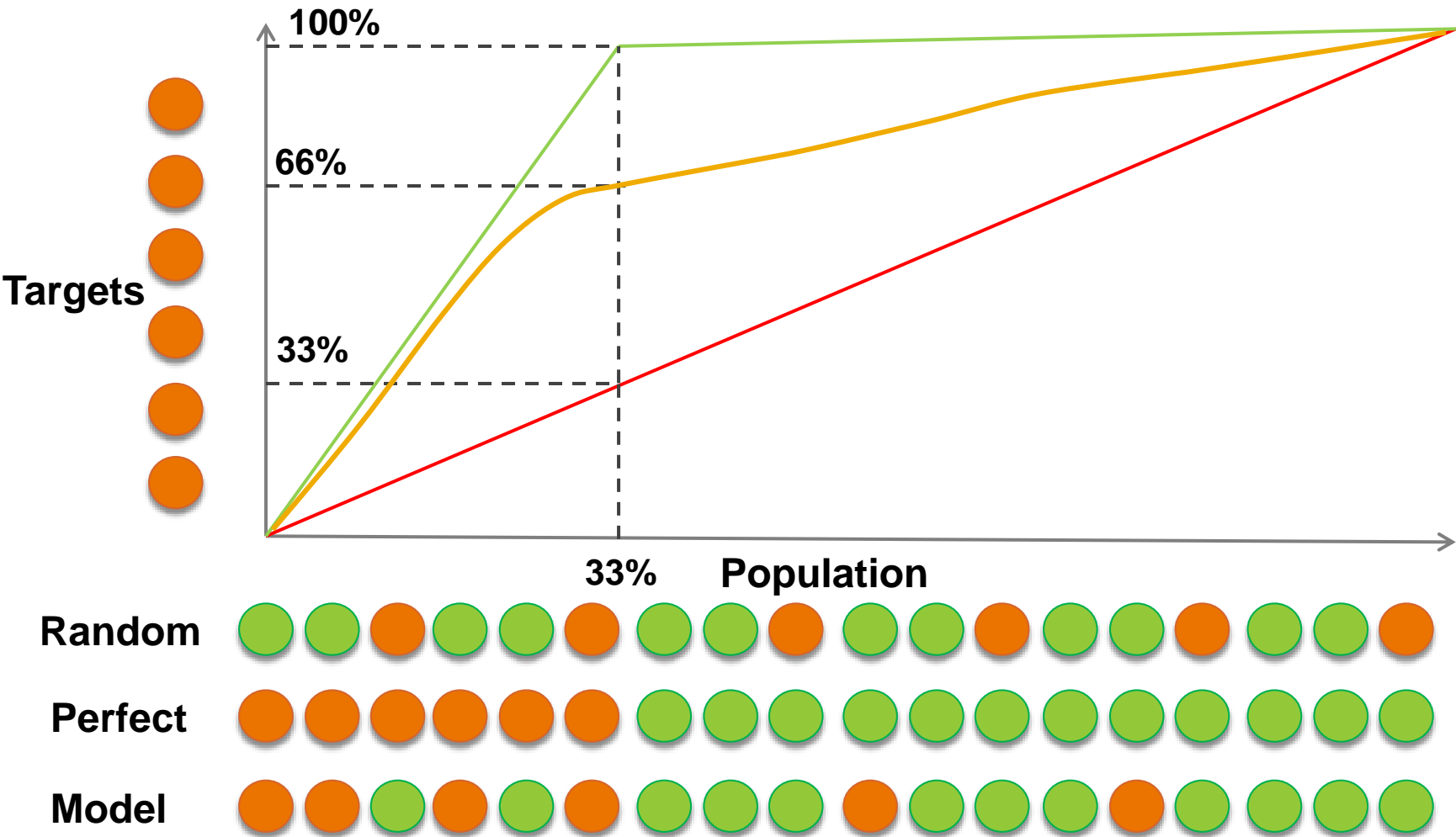
Gains (detected) chart – Our model

●

Targets

●

Non-Targets

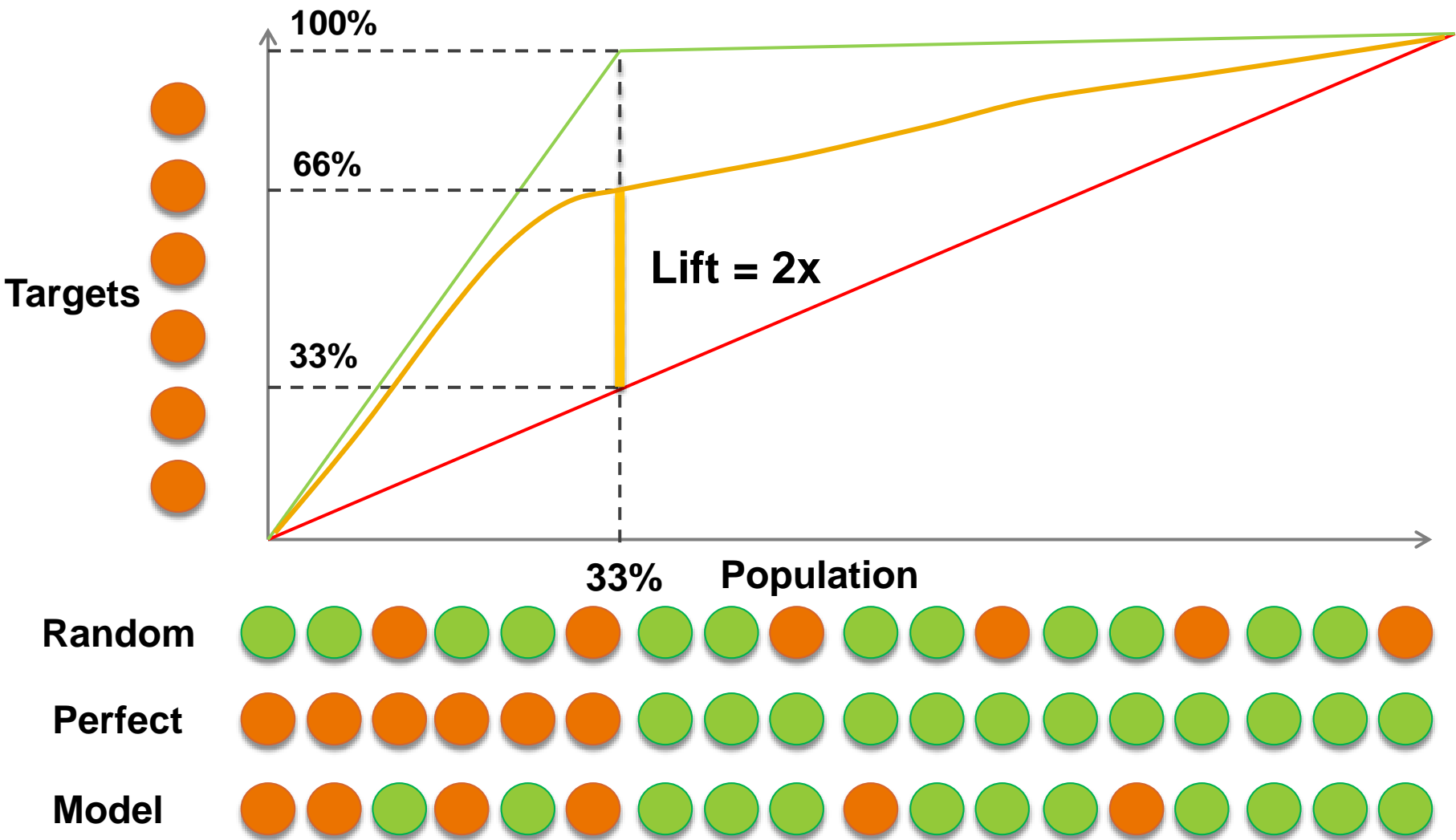


Model Performance Metrics

Lift chart

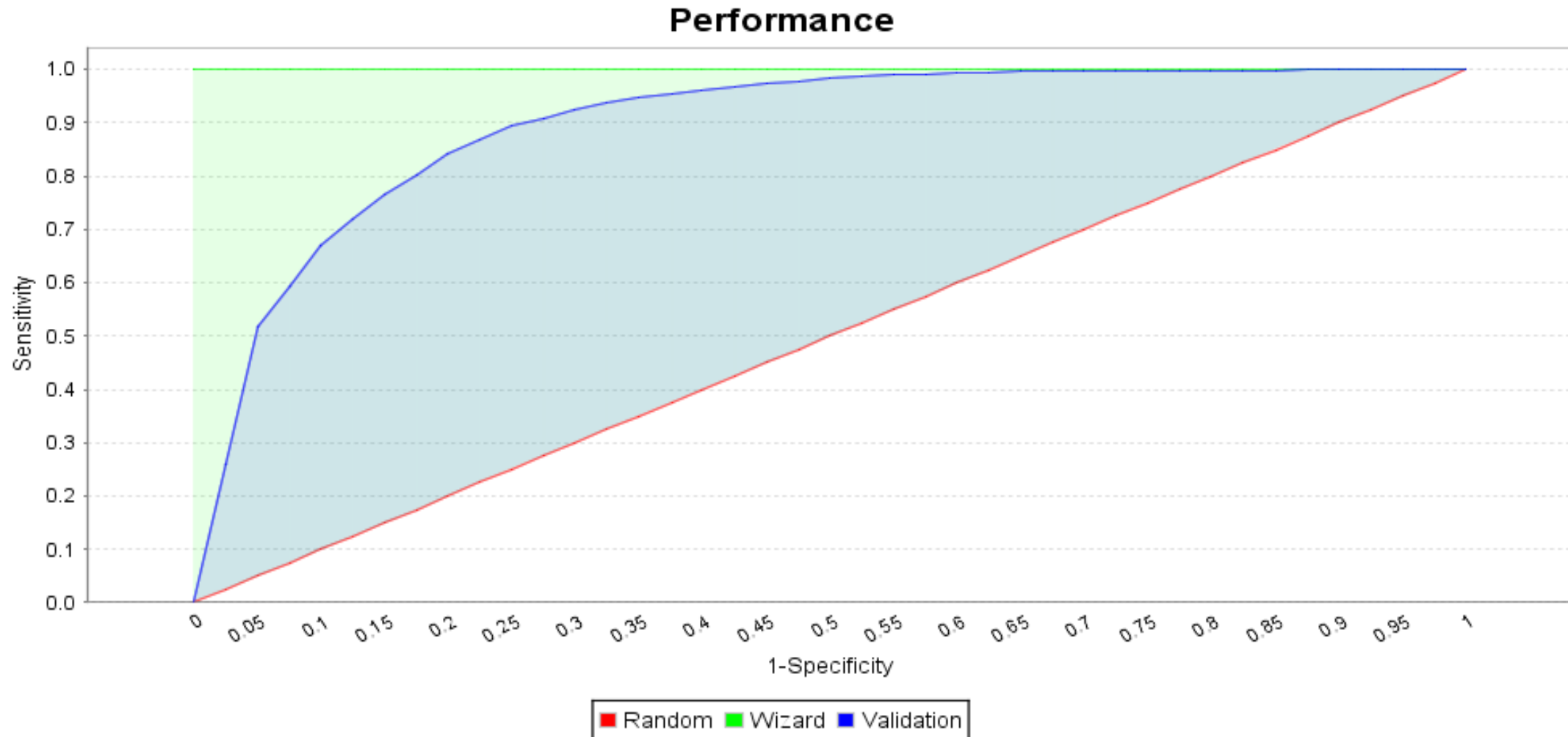
Targets

Non-Targets



Model Performance Metrics

ROC curve

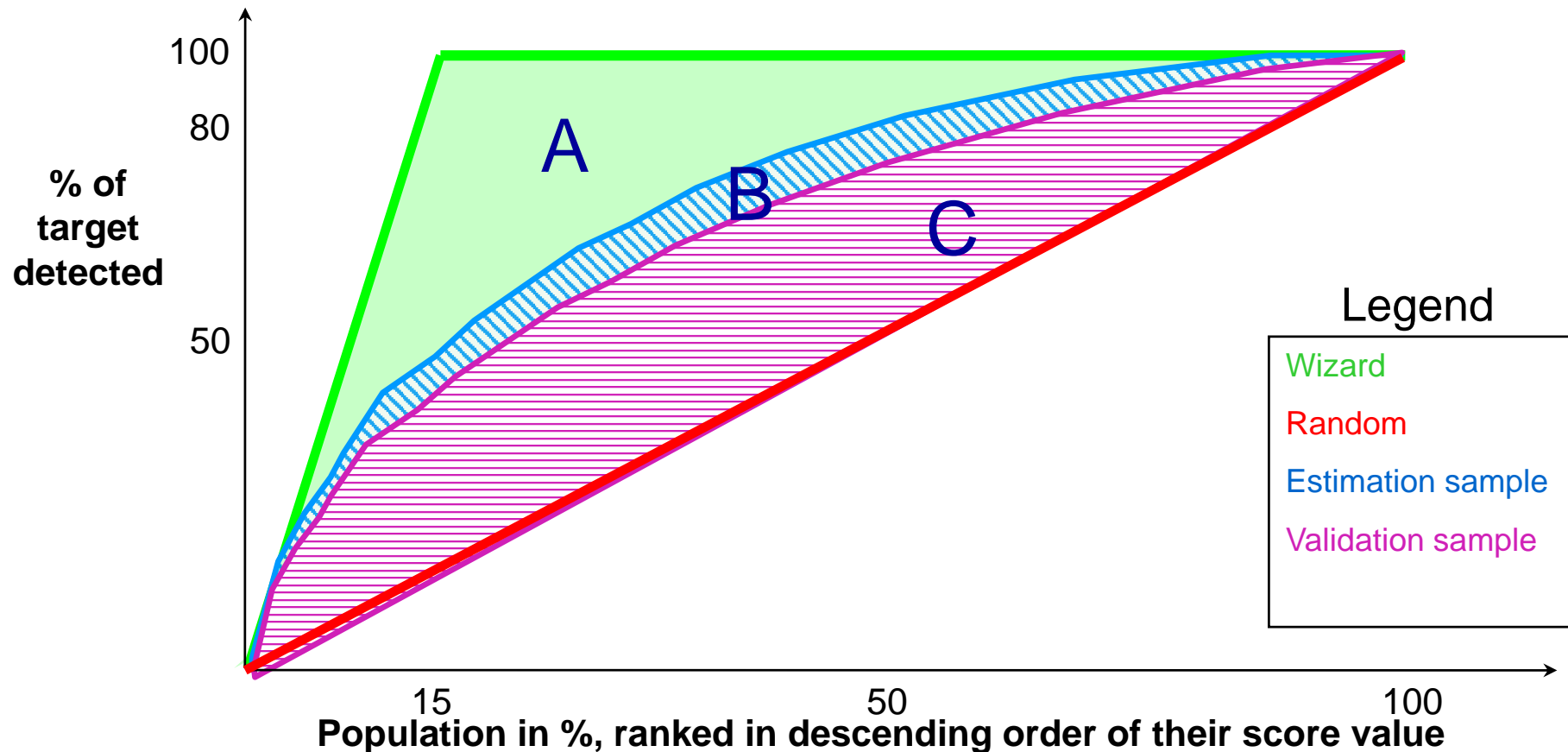


Model Performance Metrics

SAP metrics – Predictive power (KI) and prediction confidence (KR)

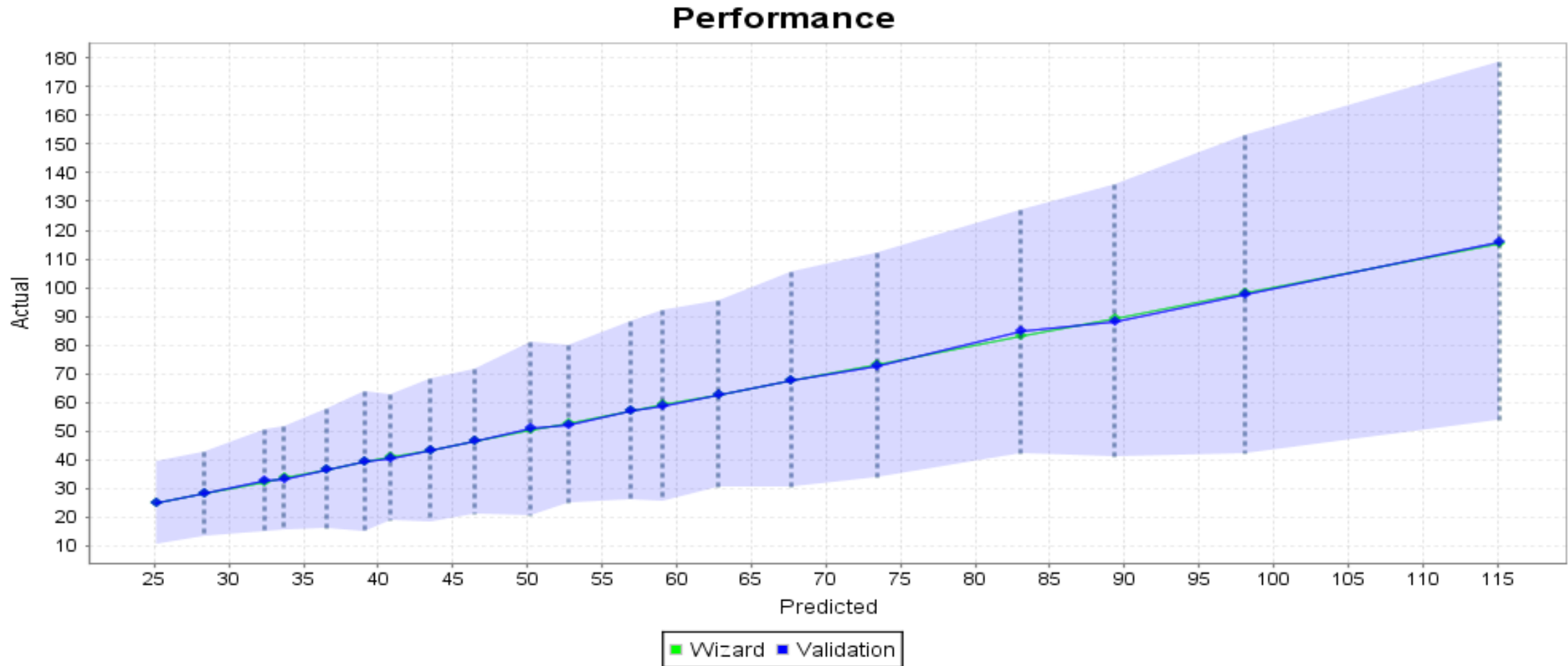
Predictive Power (KI)_{Validation} $\approx C/(A+B+C)$
Predictive Power (KI)_{Estimation} $\approx (B+C)/(A+B+C)$

Prediction Confidence (KR) $\approx 1 - B/(A+B+C)$



Model Performance Metrics

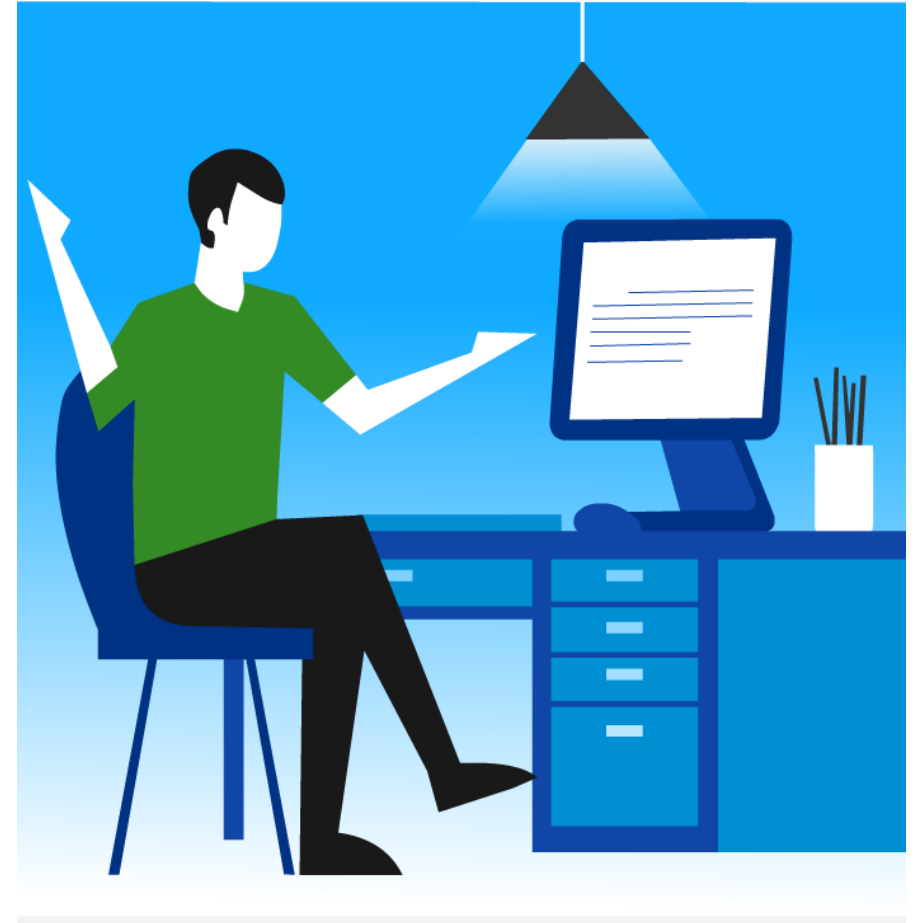
Success criteria for regression models



Model Performance Metrics

Summary

- The choice of performance metric should be the one that most closely matches the business objectives defined at the beginning of the project during the Business Understanding phase.
- The metric used for model selection is of critical importance, because the model selected based on one metric may not be a good model for a different metric.





Thank you

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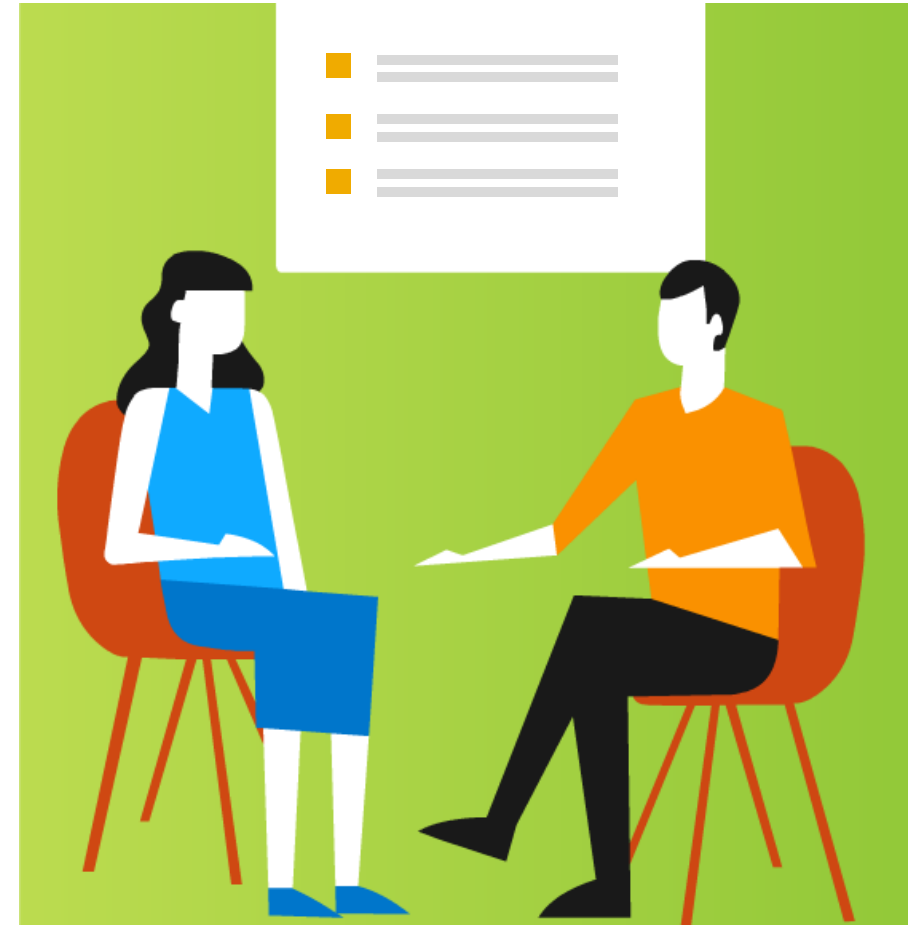
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Model Performance Metrics

Appendix

Additional Material

- Confusion Matrix
 - Commonly used metrics
- Gains and Lift Chart Examples
 - Decision tree
 - Logistic regression
- ROC Curve
 - Sensitivity and specificity
- Success Criteria for Regression Models
 - Notation
 - Mean absolute error (L1)
 - Mean square error (L2)
 - Maximum error (Linf)
 - Coefficient of determination (R2)



Model Performance Metrics

Confusion matrix – Commonly used metrics

True Positive Rate, Hit Rate, Recall, Sensitivity	TP/P	The proportion of positive instances that are correctly classified as positive
False Positive Rate, False Alarm Rate	FP/N	The proportion of negative instances that are erroneously classified as positive
False Negative Rate	FN/P	The proportion of positive instances that are erroneously classified as negative = $1 - \text{True Positive Rate}$
True Negative Rate	TN/N	The proportion of negative instances that are correctly classified as negative

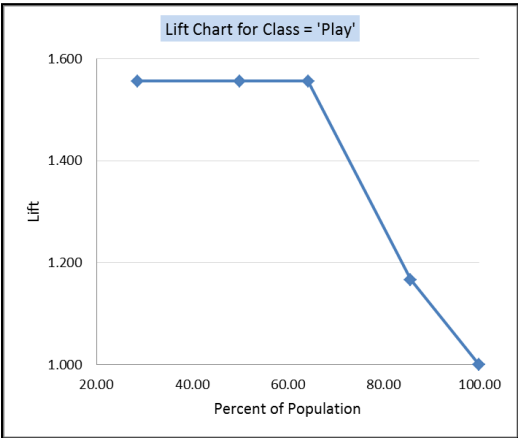
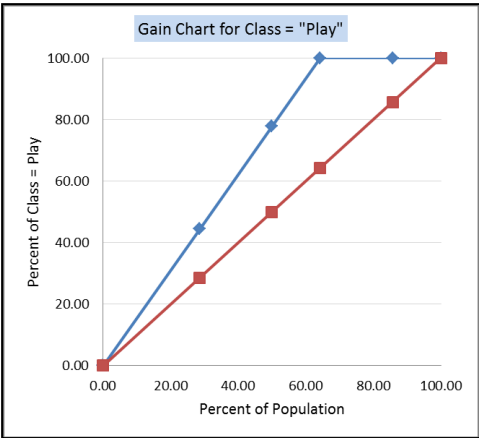
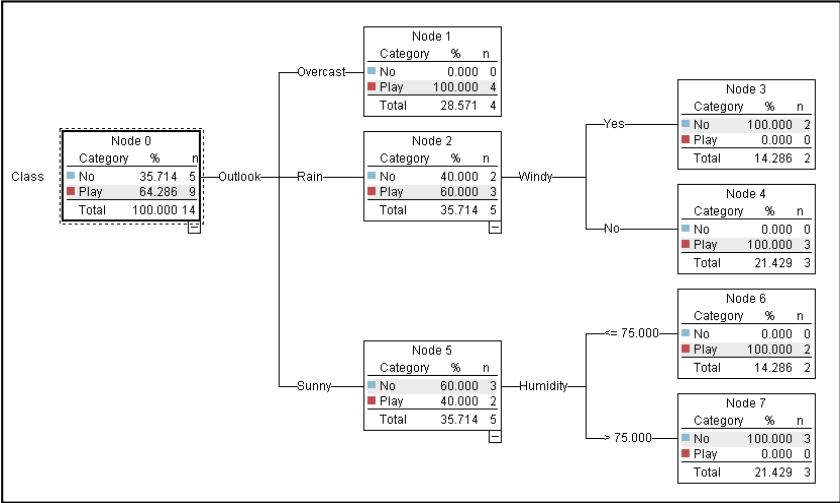
Precision, Positive Predicted Value	$TP/(TP + FP)$	Proportion of instances classified as positive that are really positive
F1 Score, Harmonic Mean	$(2 \times \text{Precision} \times \text{Recall}) / (\text{Precision} + \text{Recall})$	A measure combining Precision and Recall
Accuracy, Predictive Accuracy	$(TP + TN) / (P + N)$	The proportion of instances that are correctly classified
Error Rate	$(FP + FN) / (P + N)$	The proportion of instances that are incorrectly classified

Model Performance Metrics

Gains chart and lift chart – Classification tree example

Scenario	Outlook	n	perature	Humidity	Windy	Class	Predicted
1	Sunny		75	70	Yes	Play	Play
2	Sunny		80	90	Yes	No	No
3	Sunny		85	85	No	No	No
4	Sunny		72	95	No	No	No
5	Sunny		69	70	No	Play	Play
6	Overcast		72	90	Yes	Play	Play
7	Overcast		83	78	No	Play	Play
8	Overcast		64	65	Yes	Play	Play
9	Overcast		81	75	No	Play	Play
10	Rain		71	80	Yes	No	No
11	Rain		65	70	Yes	No	No
12	Rain		75	80	No	Play	Play
13	Rain		68	80	No	Play	Play
14	Rain		70	96	No	Play	Play

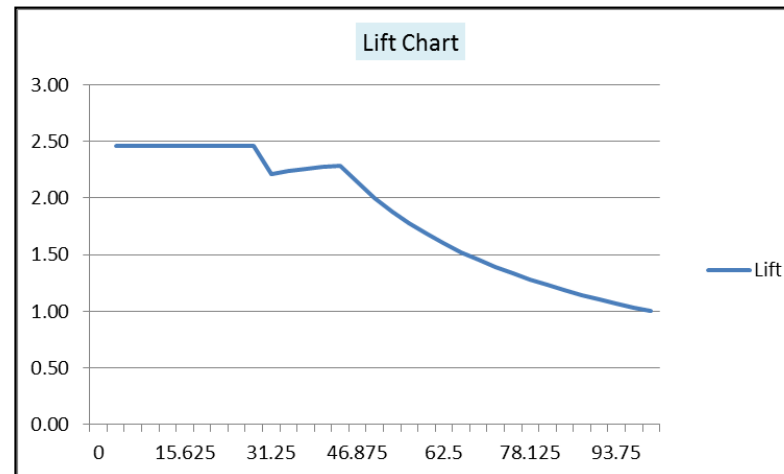
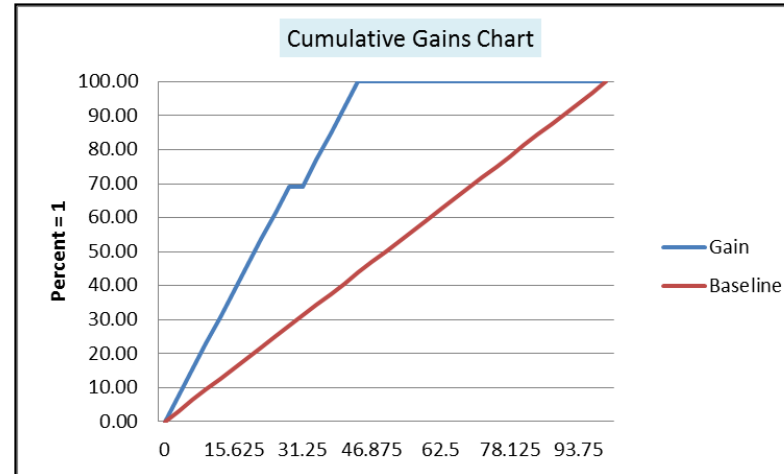
Tree Node	Class % of Bin	Play	No	Total	Cumulative Class	Cumulative Total	Cumulative % Total	Cumulative % Class	Base Line			
												Lift
1	100.00	4	0	4	4	4	28.57	44.44	28.57		28.57	1.556
4	100.00	3	0	3	7	7	50.00	77.78	50.00		50.00	1.556
6	100.00	2	0	2	9	9	64.29	100.00	64.29		64.29	1.556
7	0.00	0	3	3	9	12	85.71	100.00	85.71		85.71	1.167
3	0.00	0	2	2	9	14	100.00	100.00	100.00		100.00	1.000
Total		9	5	14								



Model Performance Metrics

Gains chart and lift chart – Logistic regression example

Record ID	Fitted	Fitted Sorted	Actual	Cumulative			Lift
				Actual	Gain	Baseline	
					0.00	0	
0	0.8423	1.0000	1	1	7.69	3.125	2.46
1	0.4048	0.9995	1	2	15.38	6.25	2.46
2	0.9702	0.9983	1	3	23.08	9.375	2.46
3	0.0417	0.9964	1	4	30.77	12.5	2.46
4	0.0694	0.9924	1	5	38.46	15.625	2.46
5	0.0050	0.9702	1	6	46.15	18.75	2.46
6	0.2480	0.9700	1	7	53.85	21.875	2.46
7	0.0093	0.9437	1	8	61.54	25	2.46
8	0.0410	0.9434	1	9	69.23	28.125	2.46
9	0.0112	0.9211	0	9	69.23	31.25	2.22
10	0.0112	0.8960	1	10	76.92	34.375	2.24
11	0.0005	0.8423	1	11	84.62	37.5	2.26
12	0.0085	0.5857	1	12	92.31	40.625	2.27
13	0.0057	0.4048	1	13	100.00	43.75	2.29
14	0.0000	0.2480	0	13	100.00	46.875	2.13
15	0.0000	0.0694	0	13	100.00	50	2.00
16	0.0000	0.0417	0	13	100.00	53.125	1.88
17	0.9700	0.0410	0	13	100.00	56.25	1.78
18	0.9995	0.0359	0	13	100.00	59.375	1.68
19	0.9983	0.0304	0	13	100.00	62.5	1.60
20	0.9211	0.0155	0	13	100.00	65.625	1.52
21	0.0155	0.0112	0	13	100.00	68.75	1.45
22	0.0304	0.0112	0	13	100.00	71.875	1.39
23	0.0359	0.0093	0	13	100.00	75	1.33
24	0.0028	0.0085	0	13	100.00	78.125	1.28
25	0.9964	0.0057	0	13	100.00	81.25	1.23
26	0.9924	0.0050	0	13	100.00	84.375	1.19
27	1.0000	0.0028	0	13	100.00	87.5	1.14
28	0.9434	0.0005	0	13	100.00	90.625	1.10
29	0.9437	0.0000	0	13	100.00	93.75	1.07
30	0.8960	0.0000	0	13	100.00	96.875	1.03
31	0.5857	0.0000	0	13	100.00	100	1.00



Model Performance Metrics

ROC curve, sensitivity, and specificity



Score Max

30%

Score Min

Churners are depicted in red

Non-churners are black

		Customers in the Top 30%	Customers not in the Top 30%	Total
Actual	Churner	5	6	11
	Non-Churner	4	16	20
Total		9	22	31

$$\text{Sensitivity (30\%)} = \frac{\text{Number of churners in top 30\%}}{\text{Total number of churners}} = \frac{5}{11} = 45\% = \text{Detected (30\%)}$$

$$\text{Specificity (30\%)} = \frac{\text{Number of non-churners not in top 30\%}}{\text{Total number of non-churners}} = \frac{16}{20} = 80\%$$

Model Performance Metrics

ROC curve, sensitivity, and specificity



2 Extreme cases :

TOP 0%		Customers in the Top 0%	Customers not in the Top 0%	Total	TOP 100%		Customers in the Top 100%	Customers not in the Top 100%	Total
Actual	Churner	0	11	11	Actual	Churner	11	0	11
	Not Churner	0	20	20		Not Churner	20	0	20
Total		0	31	31	Total		31	0	31

Sensitivity(0%) = 0
1-Specificity(0%) = 0

Sensitivity(100%) = 1
1-Specificity(0%) = 1

Model Performance Metrics

Success criteria for regression models

If we use the following notation:

- Target (response value): y_i
- Predictor (predictor response value): \hat{y}_i
- Residual: $r_i = y_i - \hat{y}_i$
- Error: $u_i = |y_i - \hat{y}_i| = |r_i|$
- Weight of the tested observation: w_i

$$W = \sum_{i=1}^n w_i$$

- Total weight of the population:

- Target average:
$$\bar{y} = \frac{1}{W} \sum_{i=1}^N w_i y_i$$

- Predictor average:
$$\bar{\hat{y}} = \frac{1}{W} \sum_{i=1}^N w_i \hat{y}_i$$

Model Performance Metrics

Success criteria for regression models – Mean absolute error (L1)

- Definition: mean of the absolute values of the differences between predictions and actual results (city block distance or Manhattan distance)
- Formula:

$$L1 = \frac{1}{W} \sum_{i=1}^N w_i u_i$$

Model Performance Metrics

Success criteria for regression models – Mean square error (L2)

- Definition: square root of the mean of the quadratic errors (Euclidian distance or root mean squared error – RMSE)
- Formula:

$$MSE = \frac{SSE_w}{W} = \frac{1}{W} \sum_{i=1}^N w_i u_i^2$$

Model Performance Metrics

Success criteria for regression models – Maximum error (Linf)

- Definition: maximum absolute difference between predicted and actual values (upper bound) (Chebyshev distance)
- Formula:

$$L_{\infty} = \max_i u_i$$

Model Performance Metrics

Success criteria for regression models – Coefficient of determination (R2)

- Definition: ratio between the variability (sum of squares) of the prediction and the variability (sum of squares) of the data.

- Formula:

$$SSR = \sum_{i=1}^N w_i (\hat{y}_i - \bar{y})^2$$

$$SST = \sum_{i=1}^N w_i (y_i - \bar{y})^2$$

$$R^2 = \frac{SSR}{SST}$$

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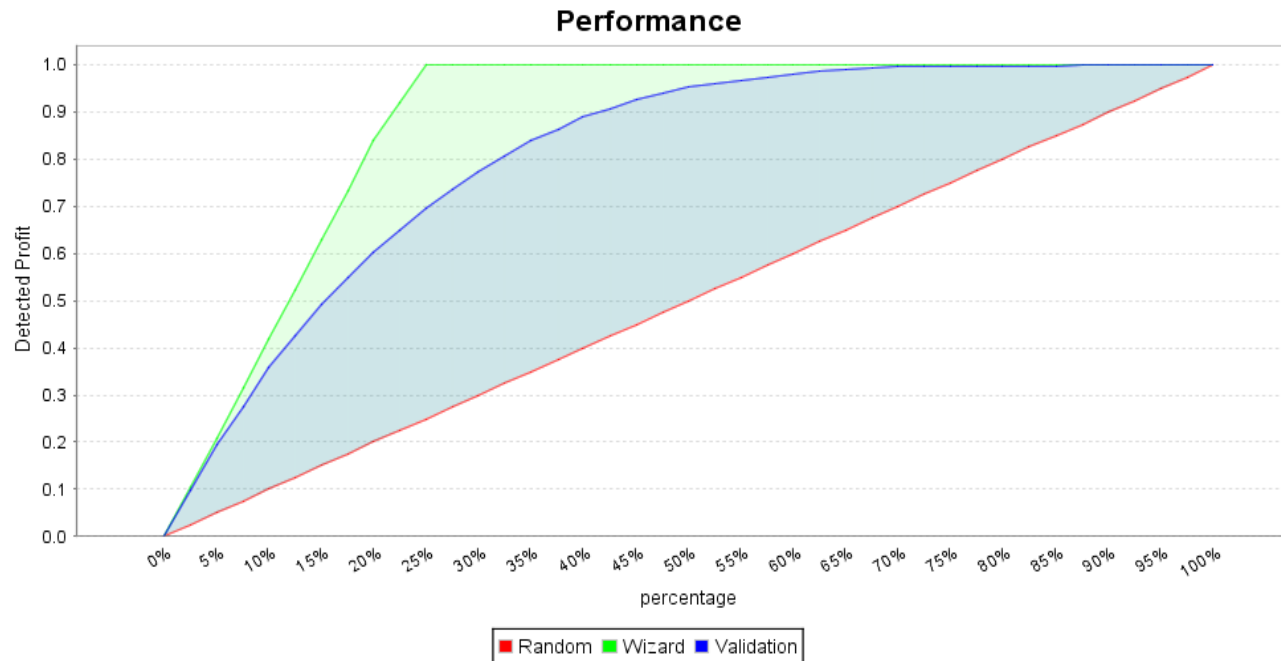
Week 5 Unit 3: Model Testing



Model Testing

Introduction – Gains charts, lift charts, and decile tables

To test the strength of classification models, many data scientists use lift charts and decile tables, which measure the performance of the model against random guessing, or what the results would be if you didn't use any model.



percentage	Random	Wizard	Validation	Estimation
0%	0.00	0.00	0.00	0.00
5%	0.05	0.21	0.20	0.20
10%	0.10	0.42	0.36	0.36
15%	0.15	0.63	0.49	0.50
20%	0.20	0.84	0.60	0.61
25%	0.25	1.00	0.70	0.70
30%	0.30	1.00	0.77	0.77
35%	0.35	1.00	0.84	0.84
40%	0.40	1.00	0.89	0.89
45%	0.45	1.00	0.93	0.93
50%	0.50	1.00	0.95	0.95
55%	0.55	1.00	0.97	0.97
60%	0.60	1.00	0.98	0.98
65%	0.65	1.00	0.99	0.99
70%	0.70	1.00	1.00	0.99
75%	0.75	1.00	1.00	1.00
80%	0.80	1.00	1.00	1.00
85%	0.85	1.00	1.00	1.00
90%	0.90	1.00	1.00	1.00
95%	0.95	1.00	1.00	1.00
100%	1.00	1.00	1.00	1.00

Model Testing

Direct mailing example

- Last year, a company sent out a mail campaign, without using a predictive model, to 10,000 customers.
- It cost the company \$1 for each item mailed.
- The response rate was 20% (there were 2,000 positive responses).

<i>Total Customers Contacted</i>	<i>Positive Responses</i>	<i>Cost (\$)</i>
10000	2000	10000

Model Testing

Direct mailing example

Decile	Total Customers Contacted	Positive Responses per decile	Cumulative Positive Responses	Cost (\$) to contact
1	1000	600	600	1000
2	2000	400	1000	2000
3	3000	300	1300	3000
4	4000	280	1580	4000
5	5000	120	1700	5000
6	6000	100	1800	6000
7	7000	80	1880	7000
8	8000	60	1940	8000
9	9000	40	1980	9000
10	10000	20	2000	10000

High Score

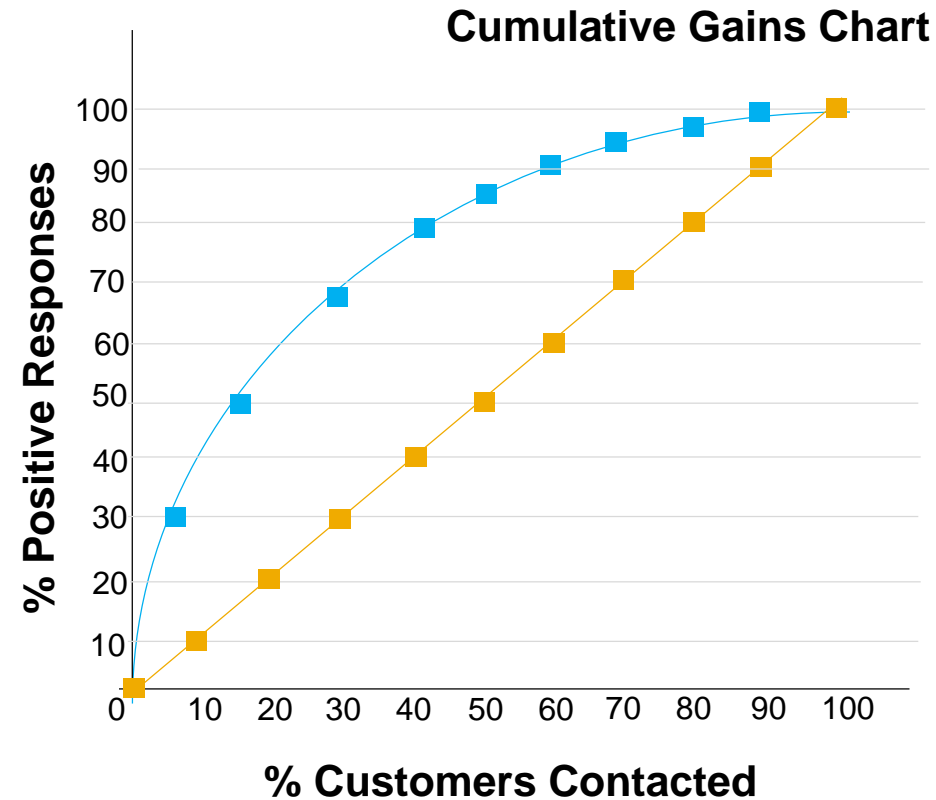


Low Score

Model Testing

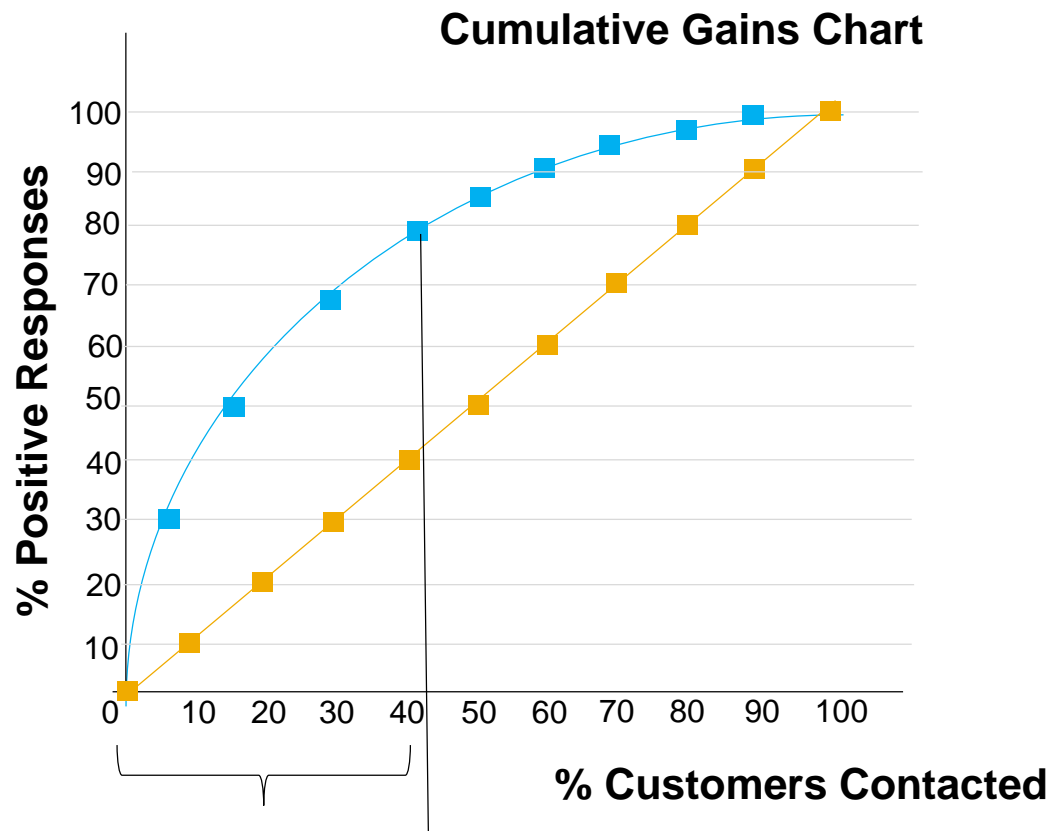
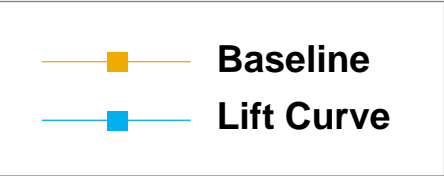
Direct mailing example – Cumulative gains chart

—■— Baseline
—■— Lift Curve



Model Testing

Direct mailing example – Cumulative gains chart



Top 40% of customers with the highest model scores

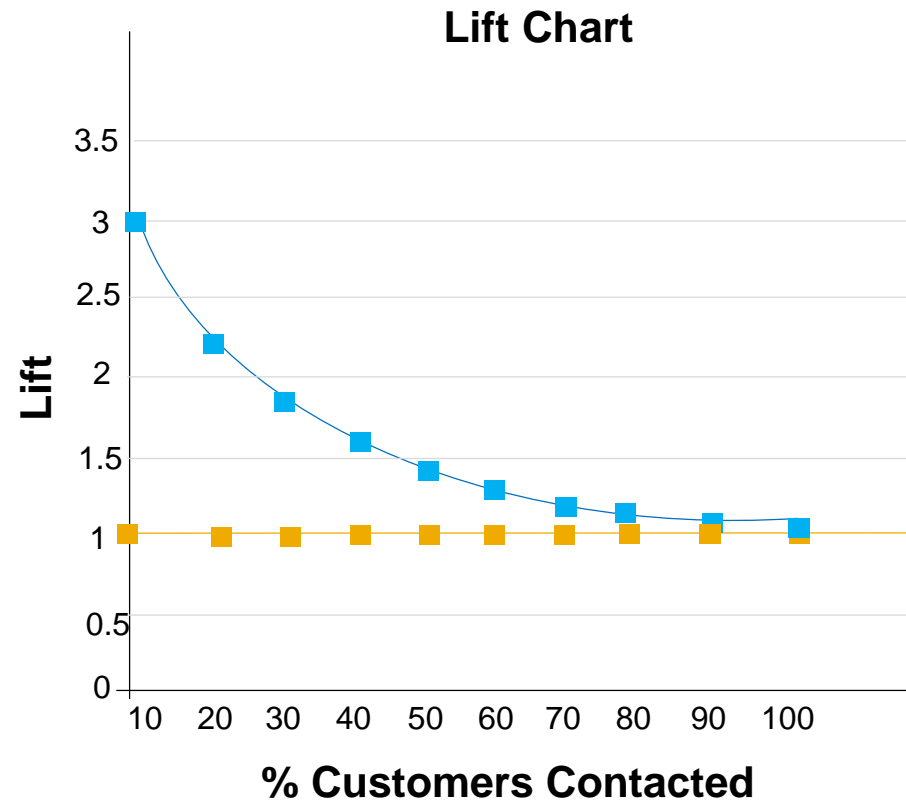
Decile	Total Customers Contacted	Positive Responses per decile	Cumulative Positive Responses	Cost (\$) to contact
1	1000	600	600	1000
2	2000	400	1000	2000
3	3000	300	1300	3000
4	4000	280	1580	4000
5	5000	120	1700	5000
6	6000	100	1800	6000
7	7000	80	1880	7000
8	8000	60	1940	8000
9	9000	40	1980	9000
10	10000	20	2000	10000

Decile Table

Model Testing

Direct mailing example – Lift chart

—■— Baseline
—■— Lift Curve





Thank you

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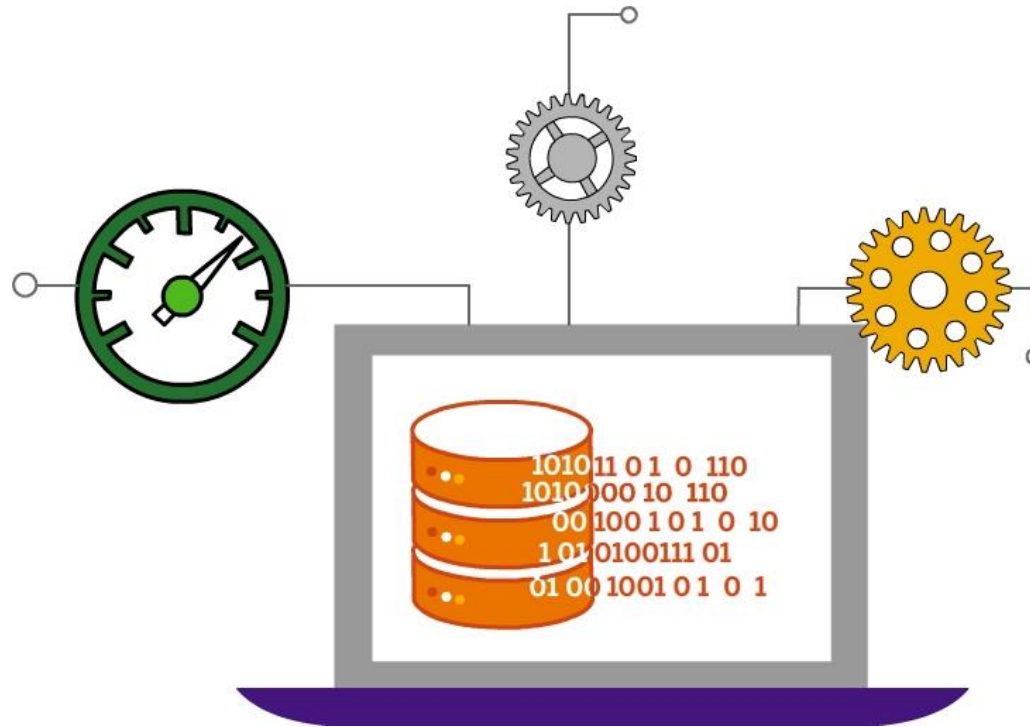
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Week 5 Unit 4: Improving Model Performance



Improving Model Performance

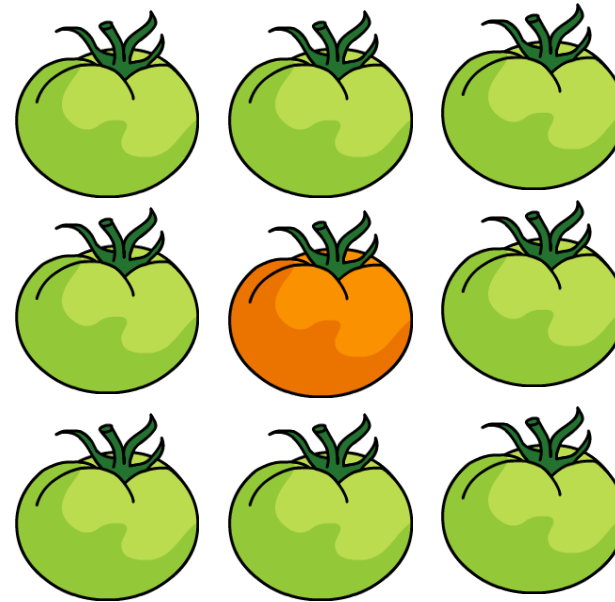
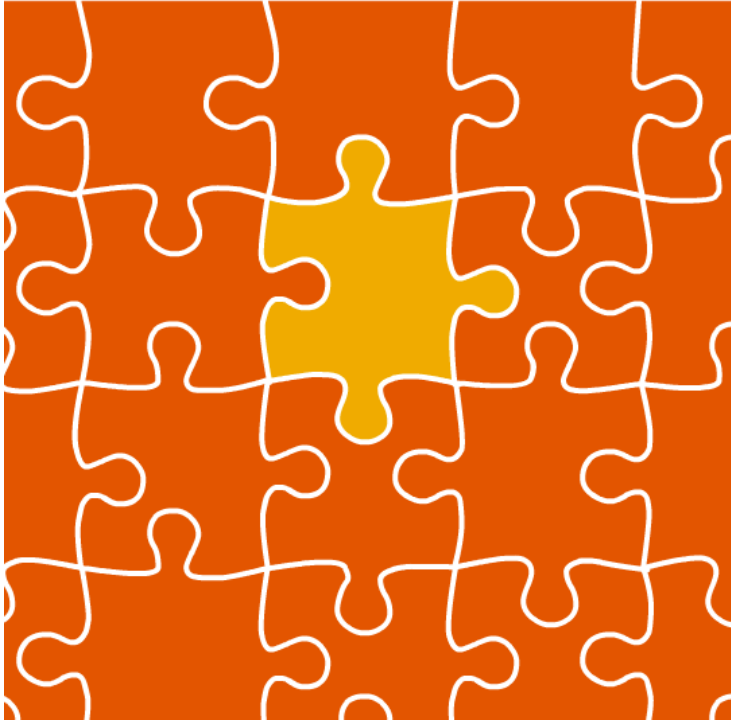
Add more data



more data = more accurate models

Improving Model Performance

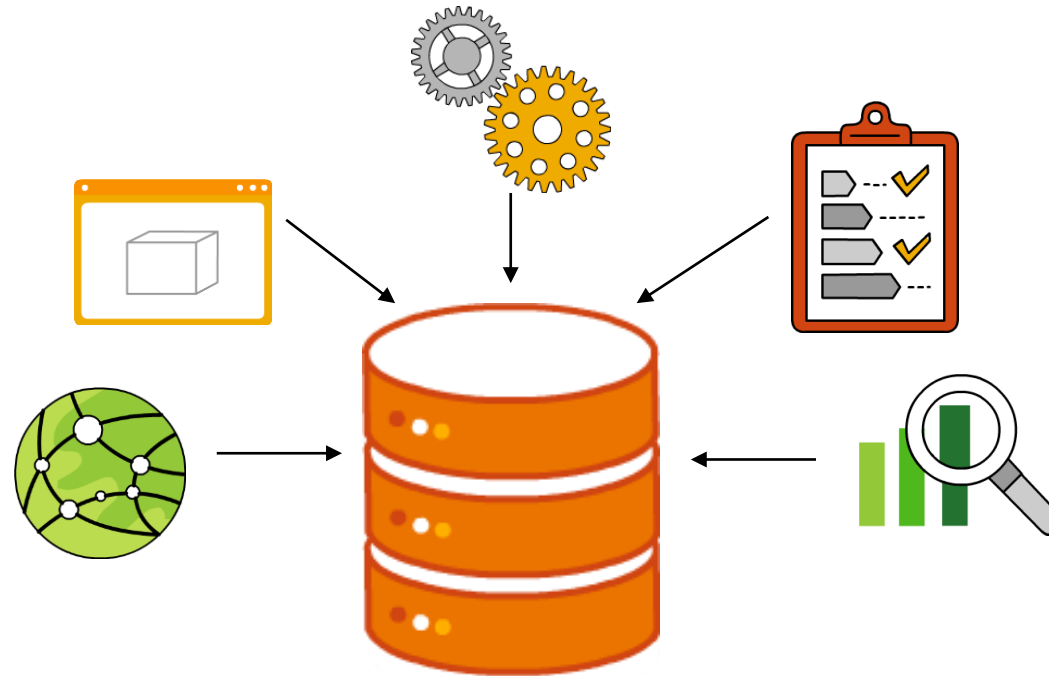
Improve data quality – Missing values and outliers



**Missing and outlier values in training data can
reduce accuracy**

Improving Model Performance

Feature engineering

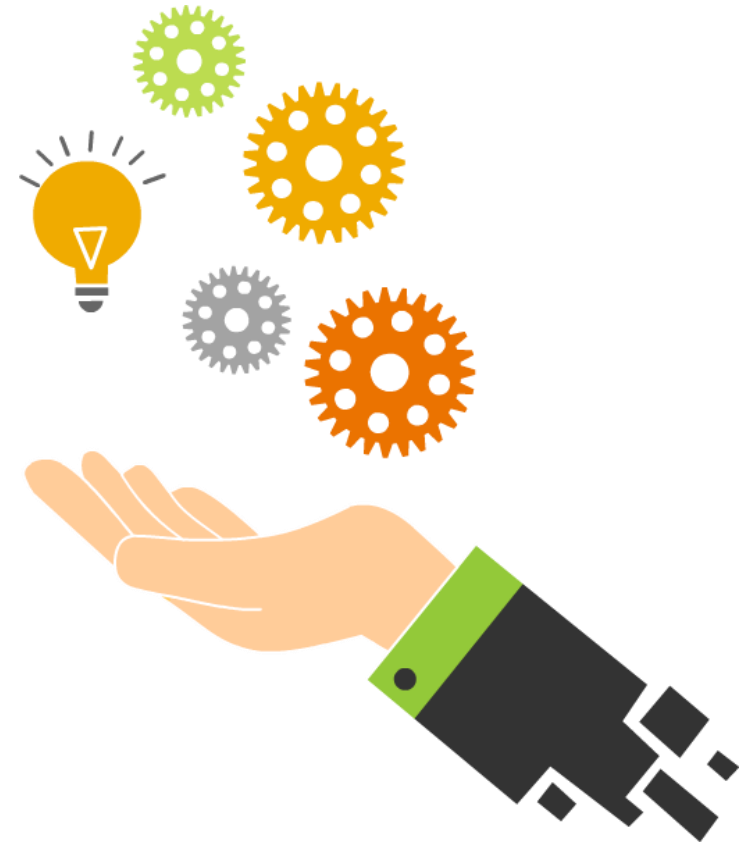


Creating new features can improve accuracy

Improving Model Performance

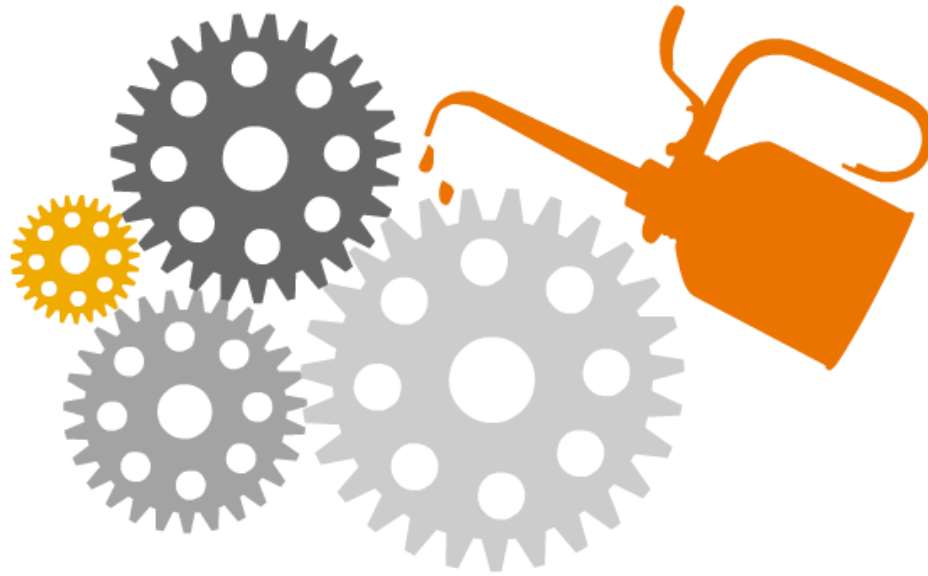
Using multiple algorithms

**Multiple algorithms might
increase accuracy**



Improving Model Performance

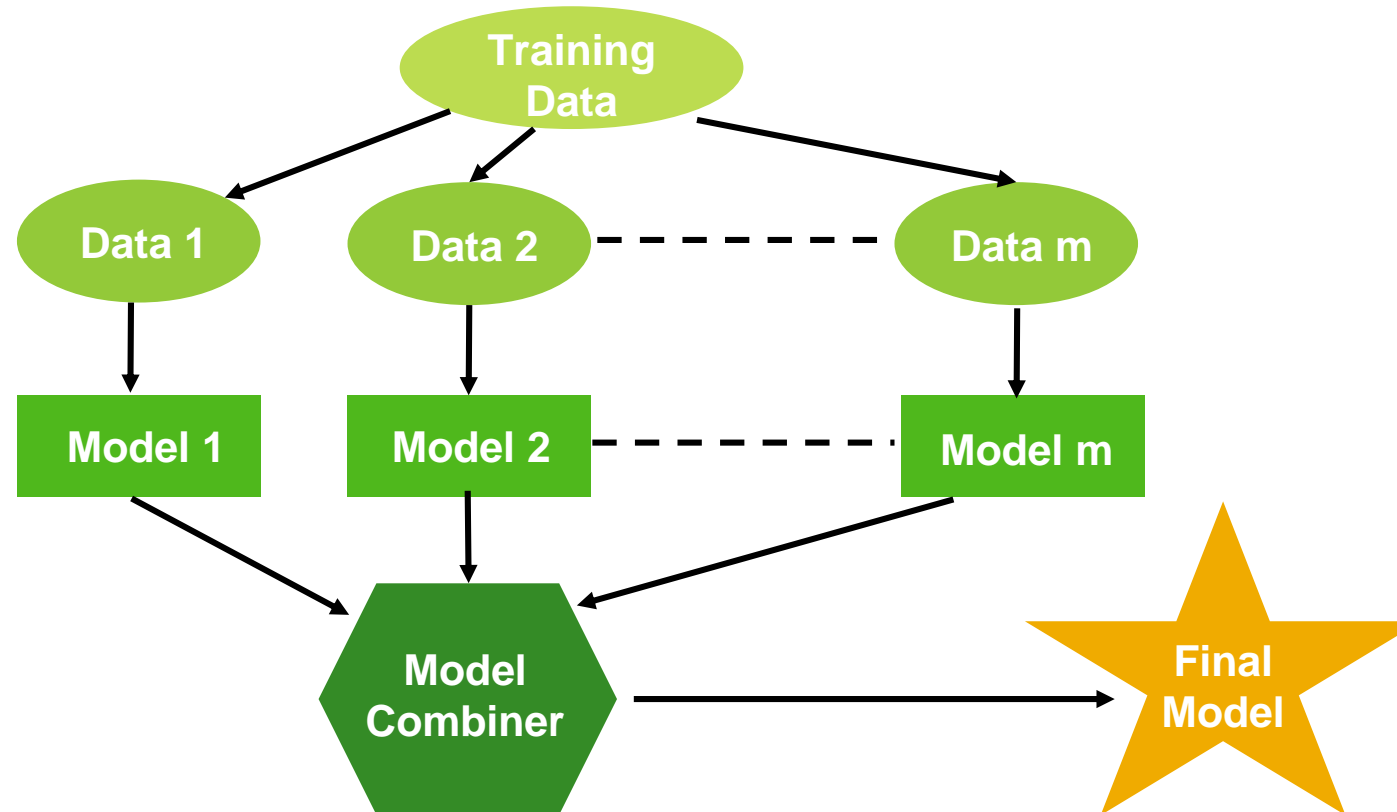
Algorithm tuning



Find the optimum value for each parameter to improve accuracy

Improving Model Performance

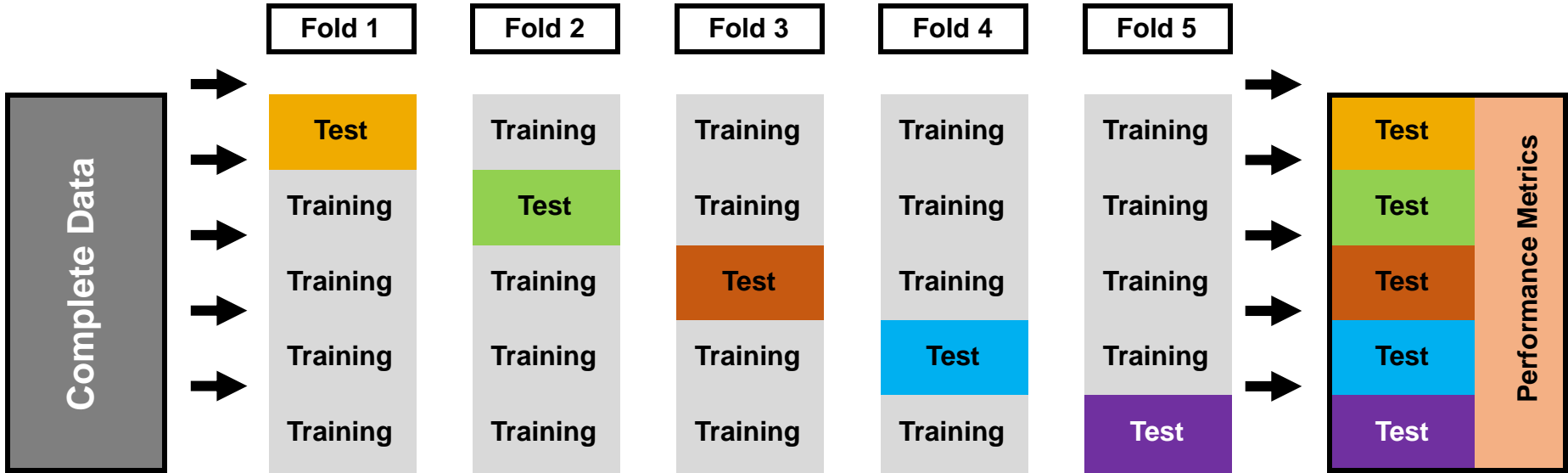
Ensemble methods



Ensemble methods can improve accuracy and robustness

Improving Model Performance

Cross-validation



Cross-validation assesses how the model will generalize



Thank you

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