Business requirements

MACHINE LEARNING FOR BUSINESS



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Scoping business needs

- 1. What is the business **situation**?
 - The company plans to expand to new markets
- 2. What is the business opportunity and how big is it?
 - Identify the right markets with the biggest demand
- 3. What are the business actions we will take?
 - Prioritize and invest more in the markets with higher predicted demand

Business scope - fraud example

- Situation The fraud rate has started increasing
- 2. Opportunity Reduce fraud rate by X %, resulting in Y USD savings
- 3. **Action** Work on improving fraud detection system, reduce fraud drivers, and manually review transactions at risk



Business scope - churn example

- Situation The customers started to churn more
- 2. Opportunity Reduce churn rate by X %, resulting in Y USD revenue saved
- 3. Action Work on identifying and improving churn drivers (website errors, too much/little advertising, customer service issues etc.); identify customers at risk and introduce retention campaigns



Business situation - asking the right question

Always start with inference questions

- Why has churn started increasing?
- Which information indicates a potential transaction fraud?
- How are our most valuable customers different from others?

Build on inference question to define prediction questions

- Can we identify customers at risk of churning?
- Can we flag potentially risky transactions?
- Can we predict early on which customers are likely to become highly valuable?

Business opportunity

Would you spend 1 million USD to earn extra 5000 USD each year? (~200 year return on investment)

- Size up the opportunity
- Once you know the drivers of the outcome, how much will it cost changing them, and what will be the value of doing that?
- Finally, how do you know if you can affect the predicted outcome? (hint experiments, experiments, and more experiments)

Actionable machine learning

Finally, how do you know if you can affect the predicted outcome? (hint - experiments, experiments, and more experiments)

- First, look at historical levels (churn, fraud, # of high value customers)
- Run experiments e.g. target customers at risk with a discount, manually review top 10% riskiest transactions. Repeat experiments multiple times, see if you get a repeated pattern of desired results
- If yes, use that to calculate opportunity and make decision if it's a worthwhile investment
- If no 1) collect more data, 2) qualitative research, 3) narrow down business question

Let's practice!

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Model training

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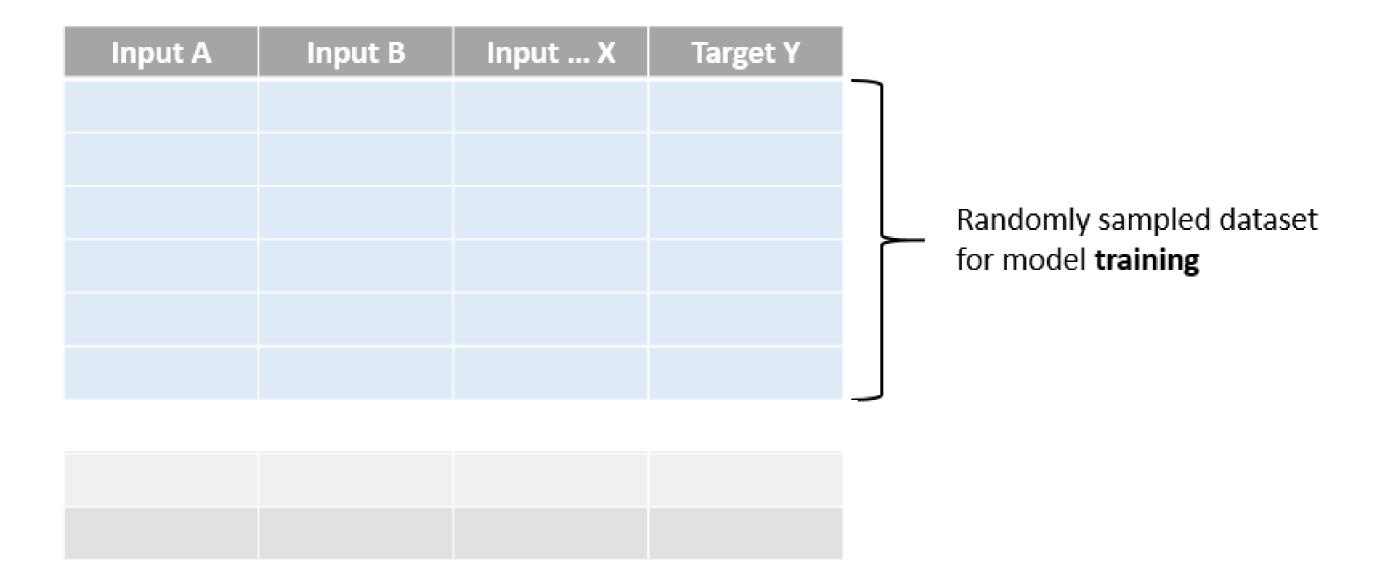
Modeling dataset

Input A	Input B	Input X	Target Y

Full dataset

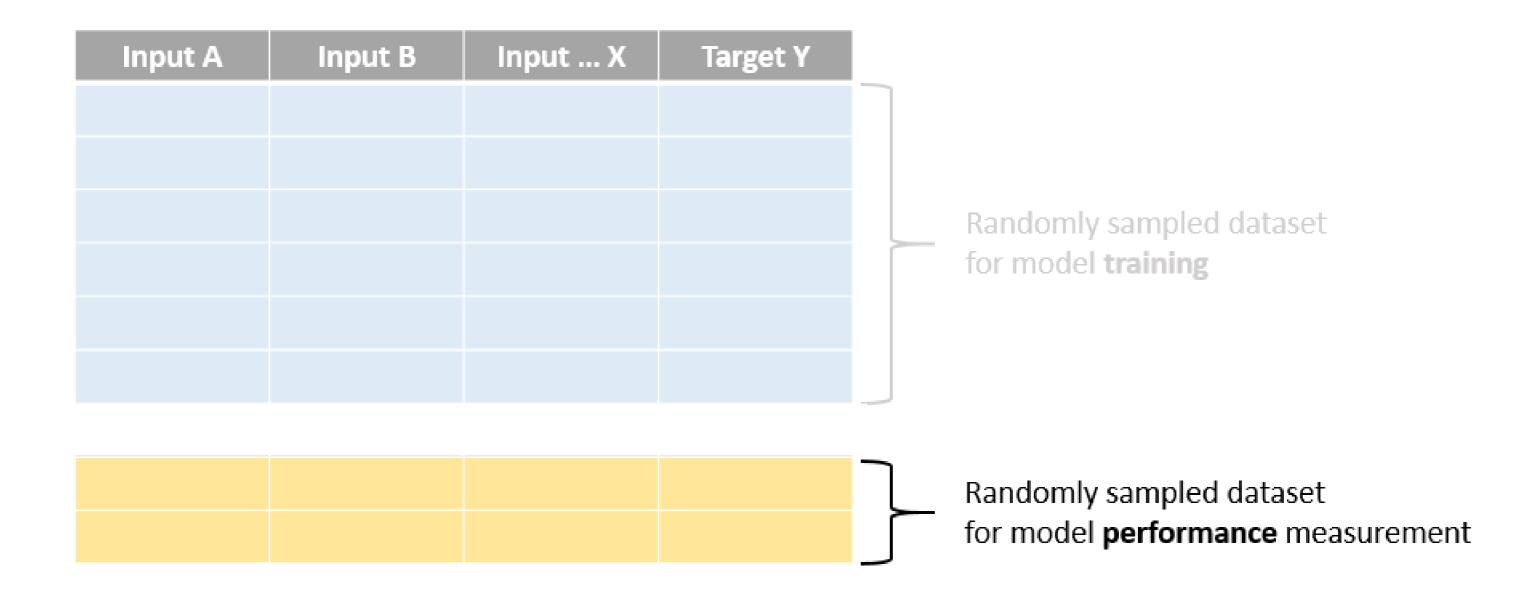
Input A	Input B	Input X	Target Y		
				_	Full dataset

Splitting data for training

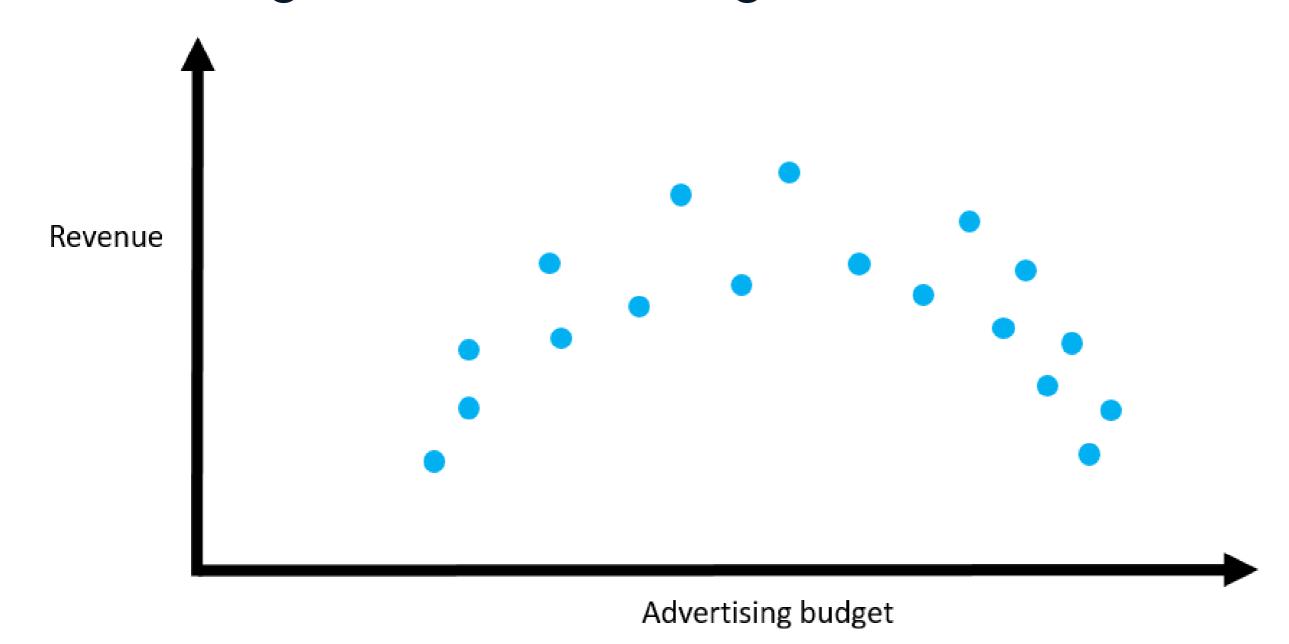




Test

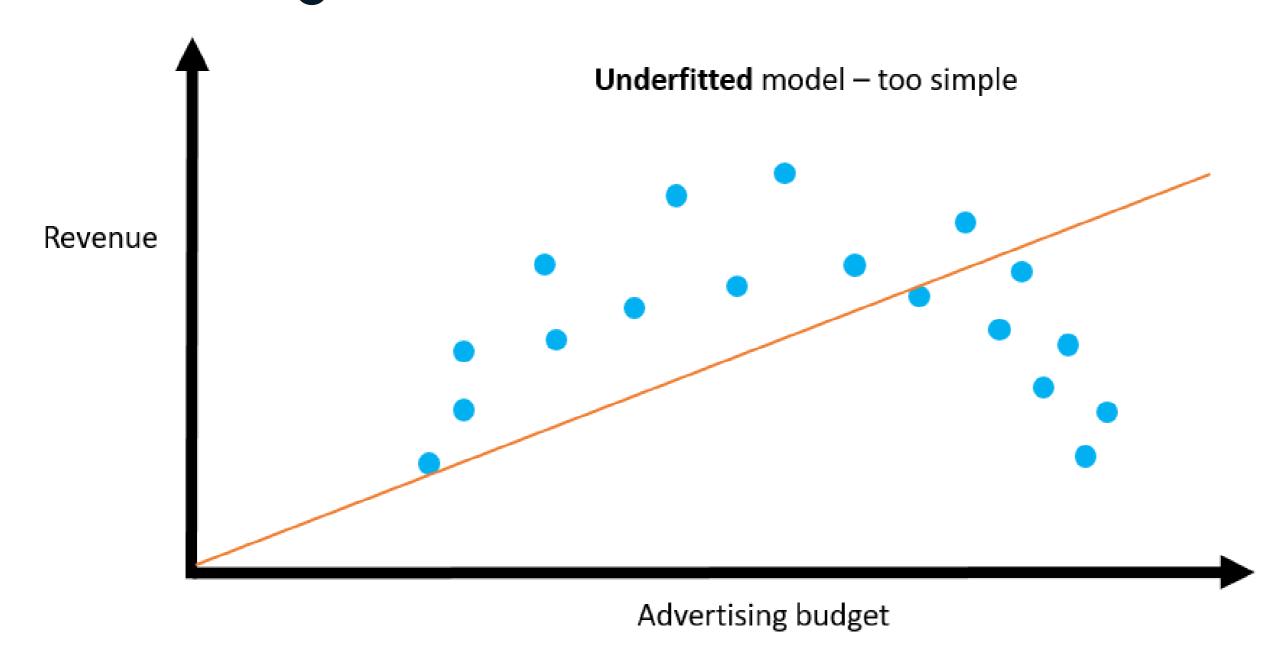


Overfitting and underfitting



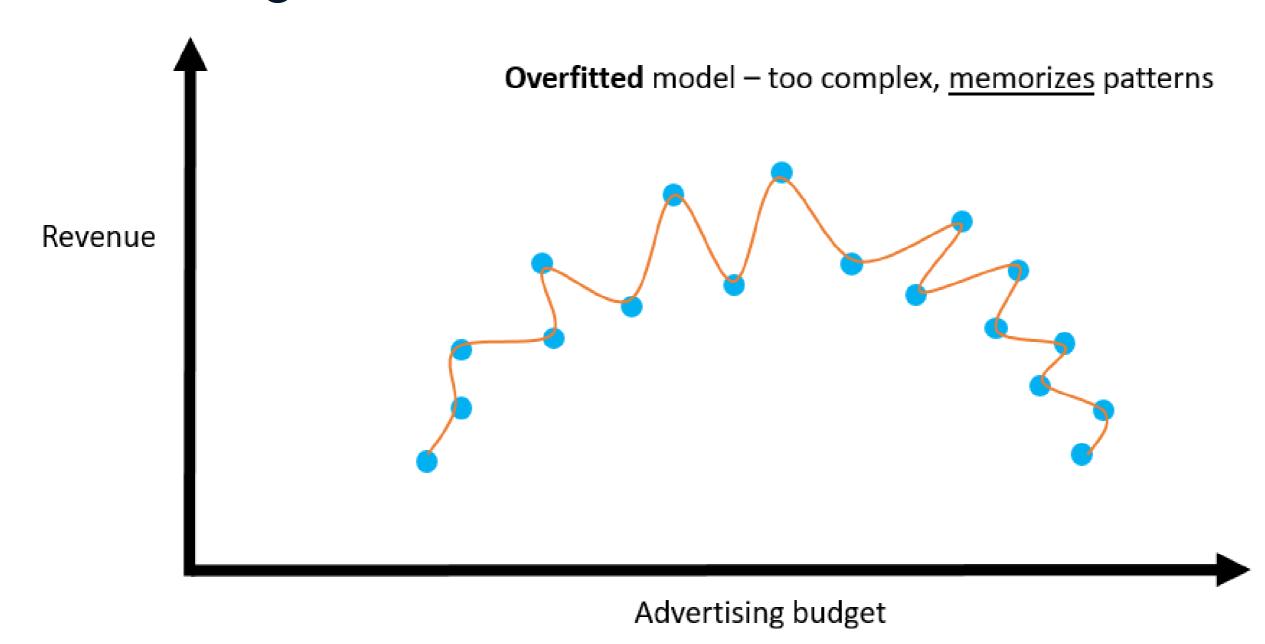


Underfitting



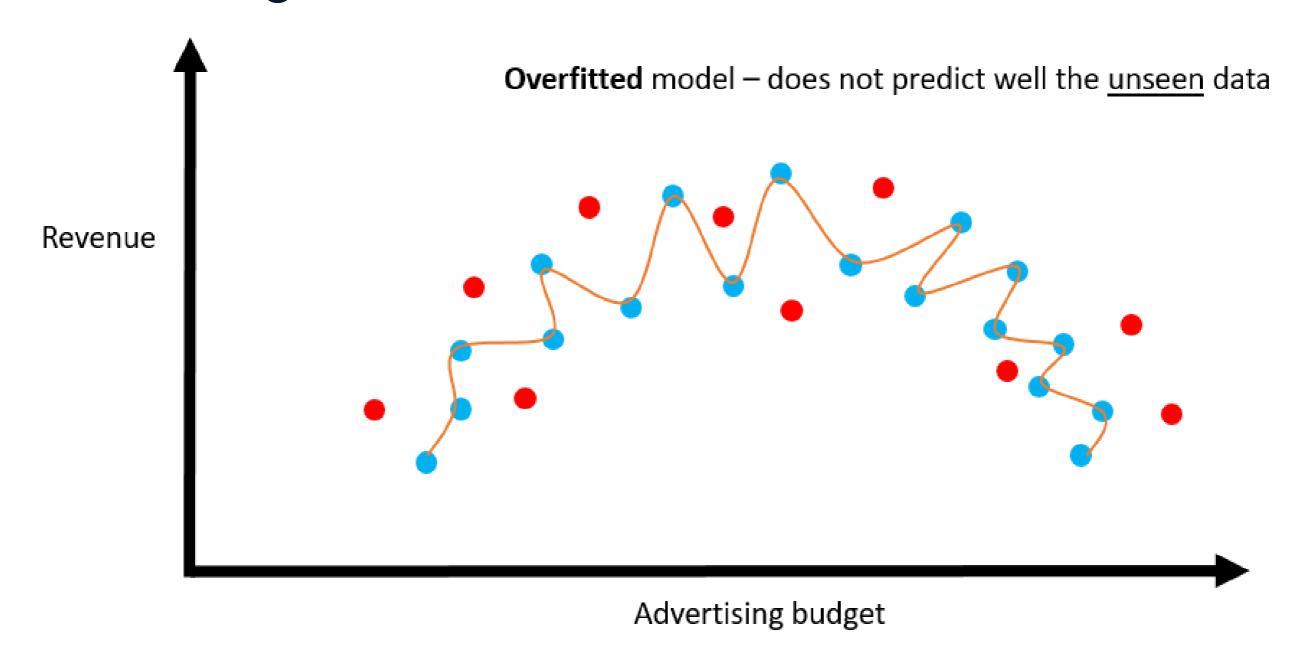


Overfitting 1



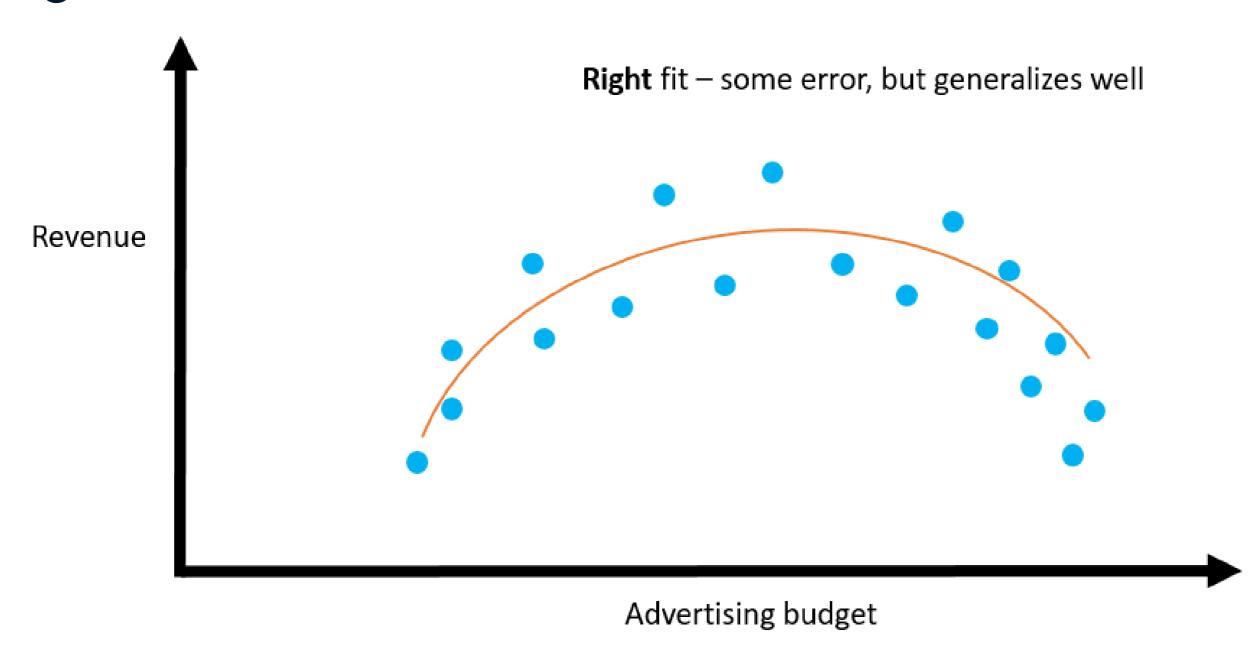


Overfitting 2



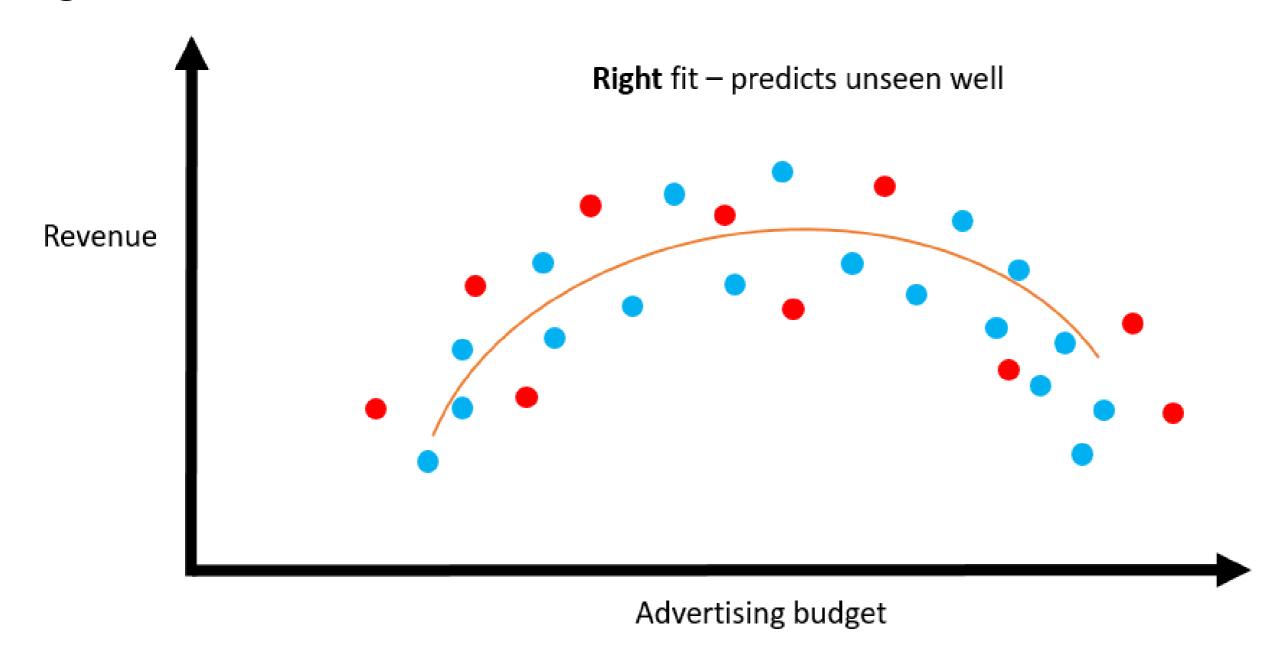


Right model fit 1



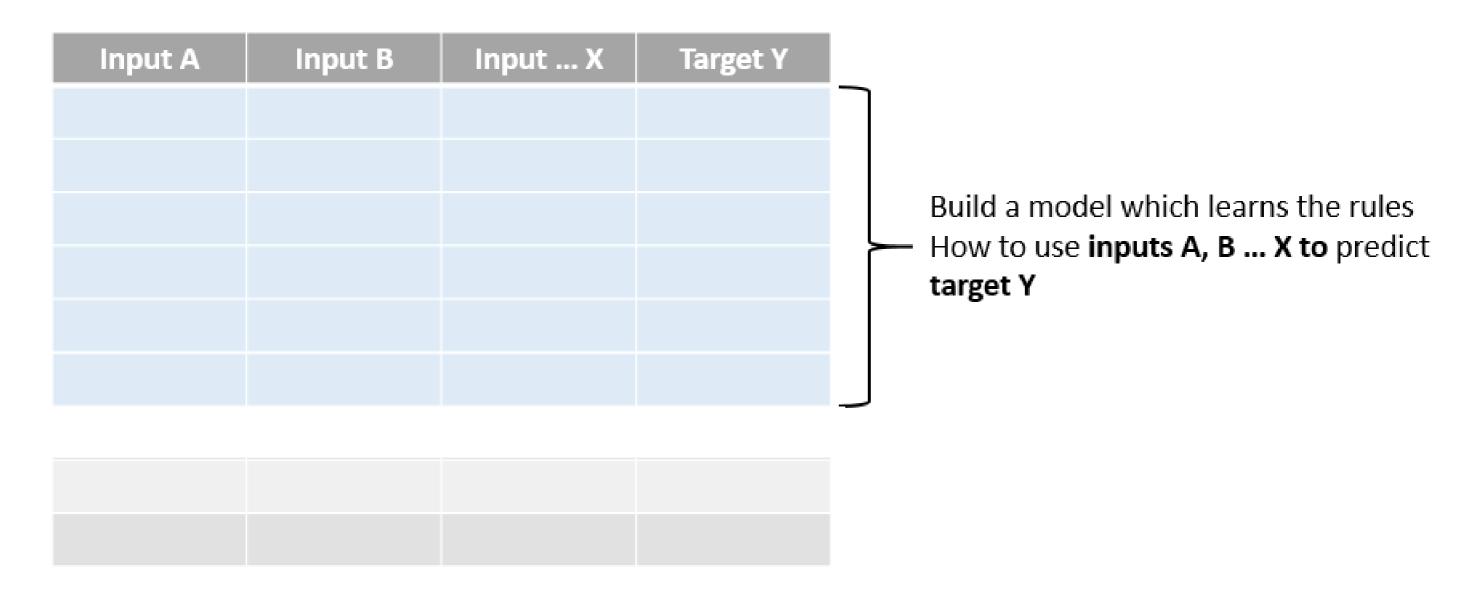


Right model fit 2

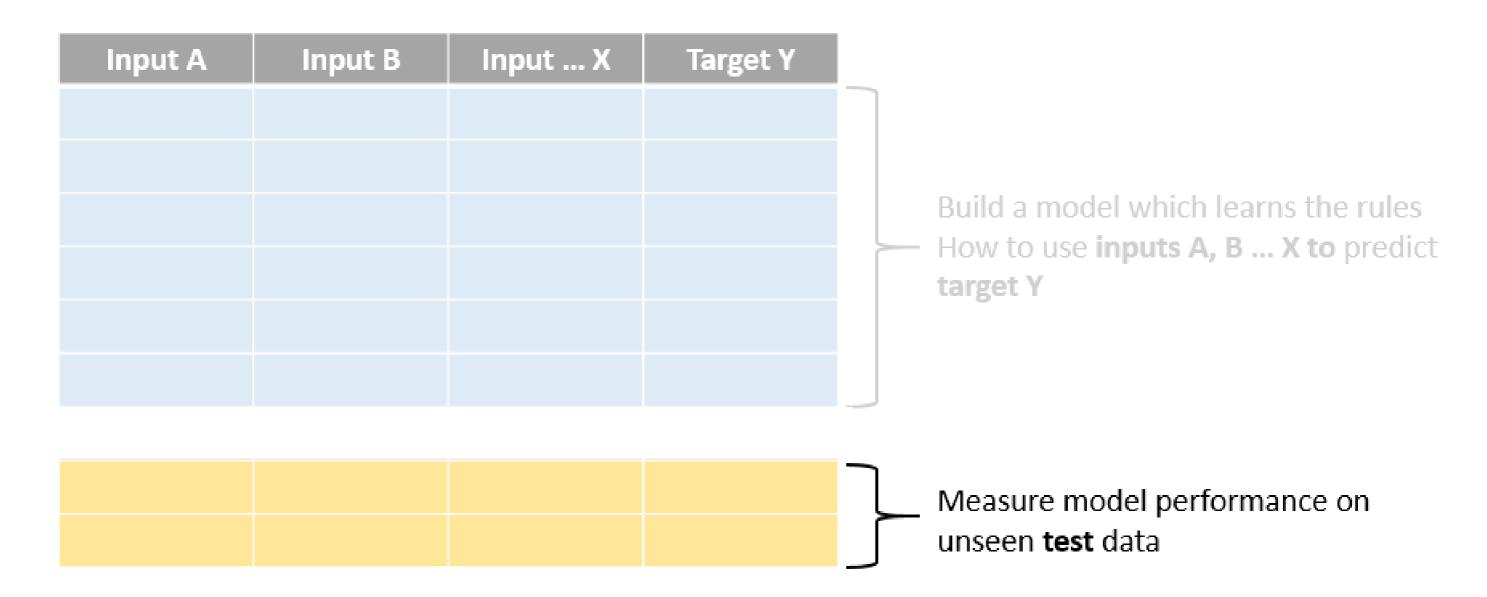




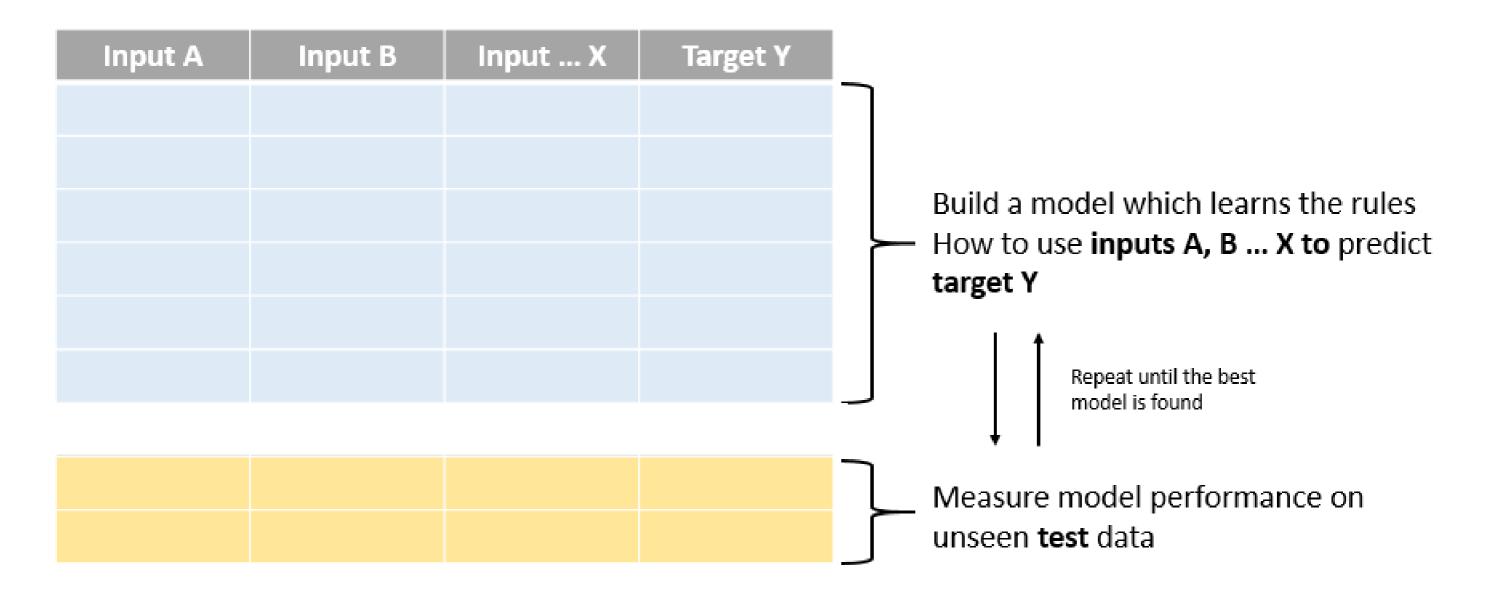
Model training



Assess model performance on test



Try a few models



Let's practice!

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Model performance measurement

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Performance measurement types

Two key supervised learning metrics:

- 1. Accuracy --> classification
- 2. Error --> regression

Classification performance

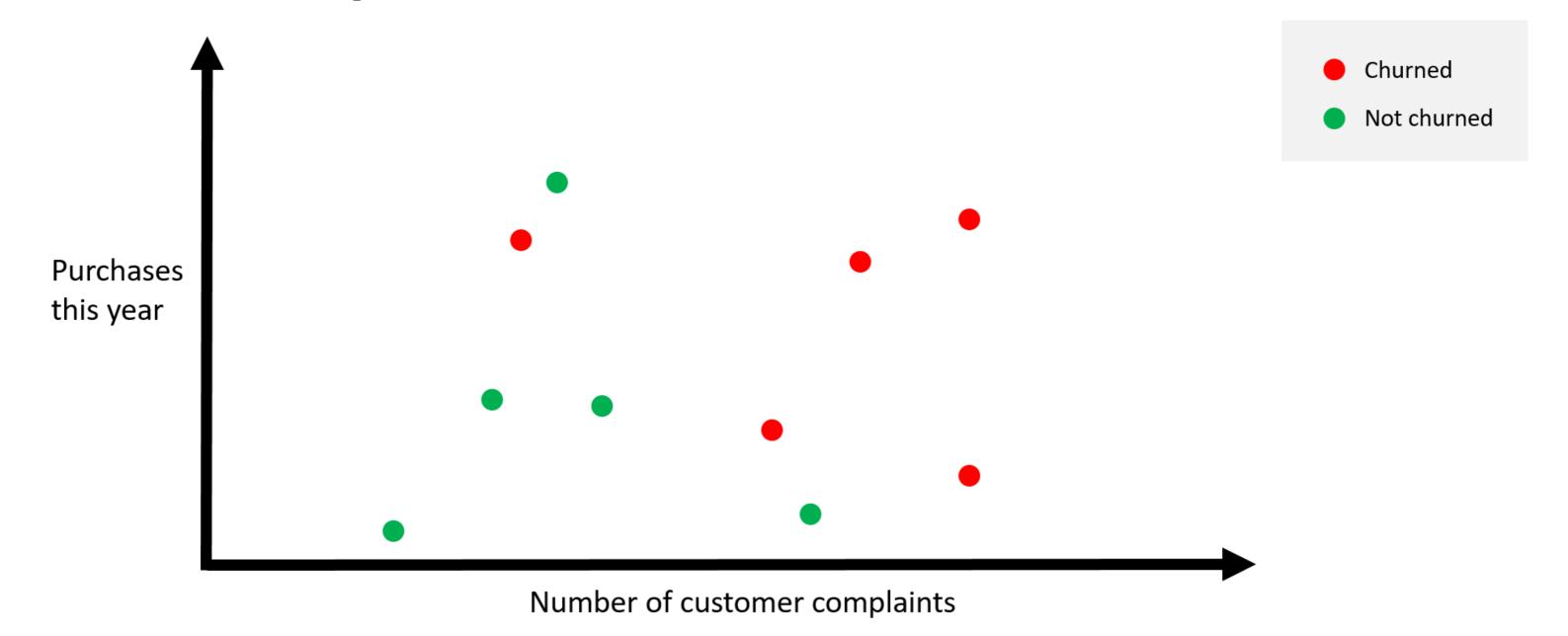
- Accuracy
- Recall
- Precision

Churn example

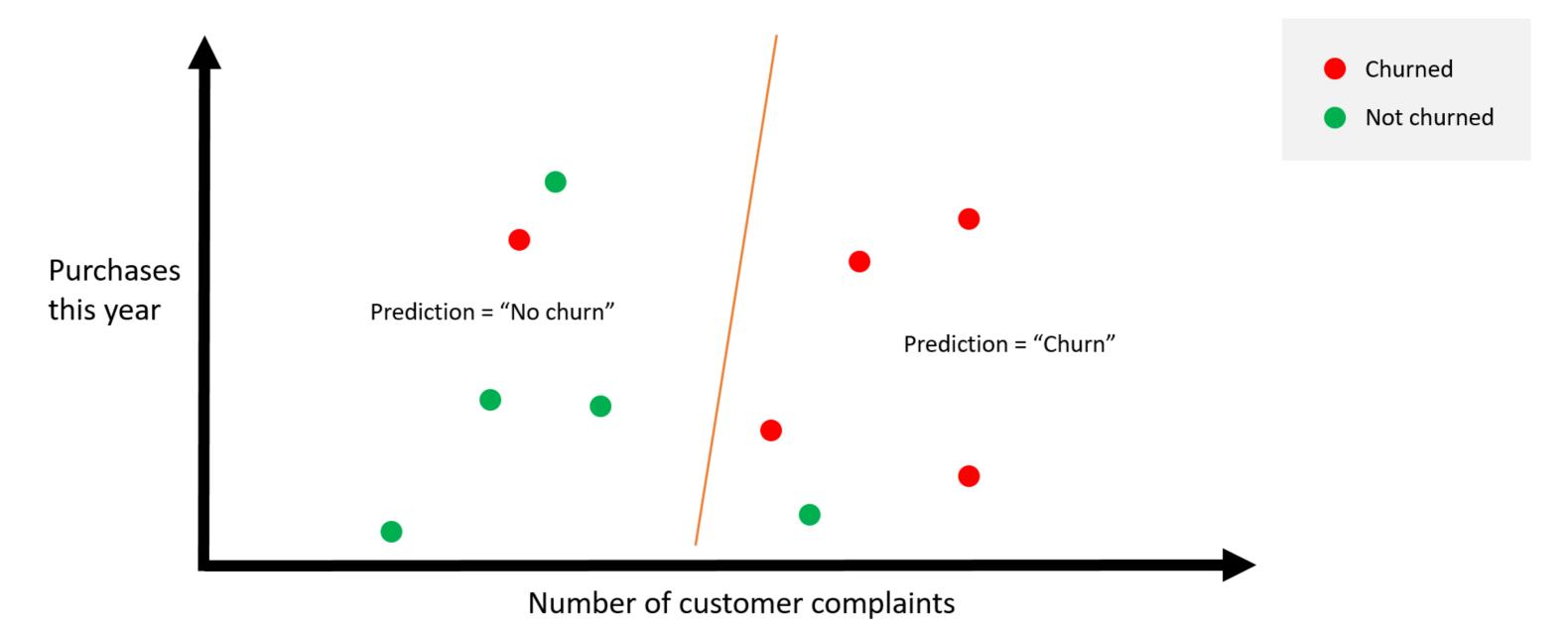




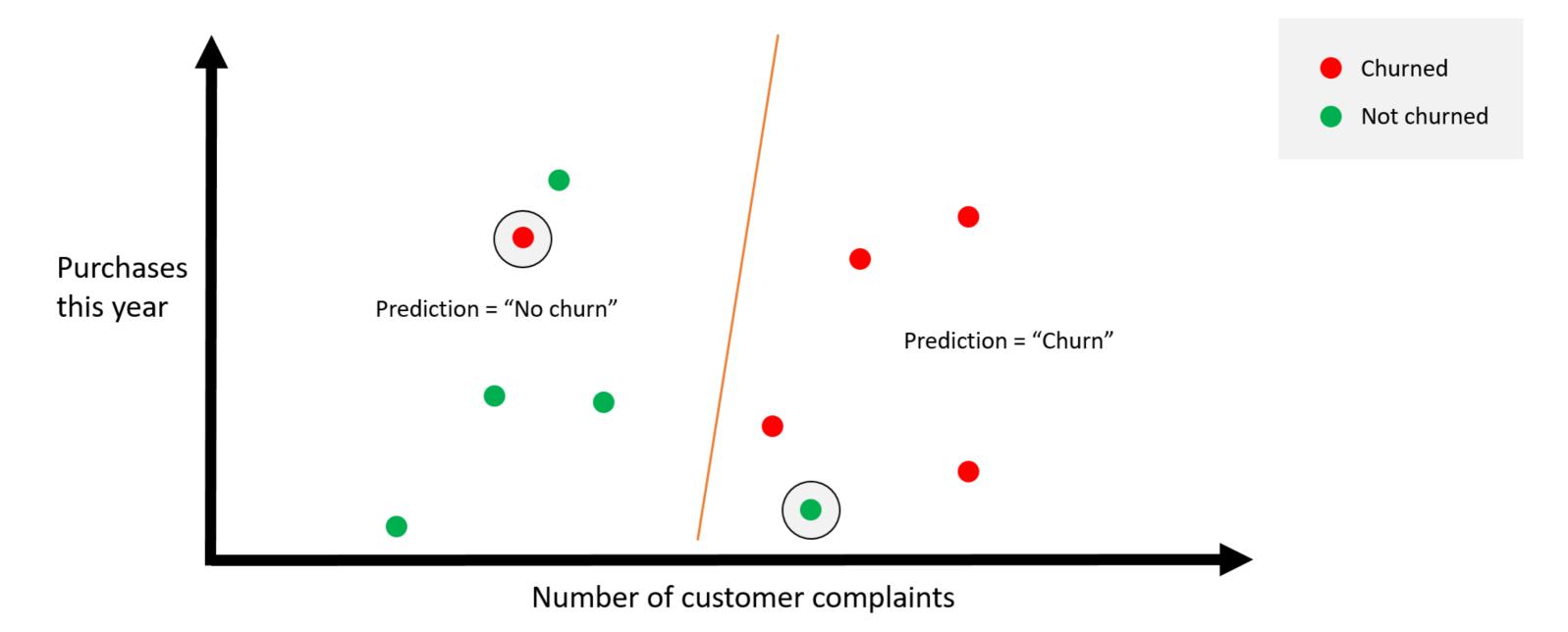
Churn example



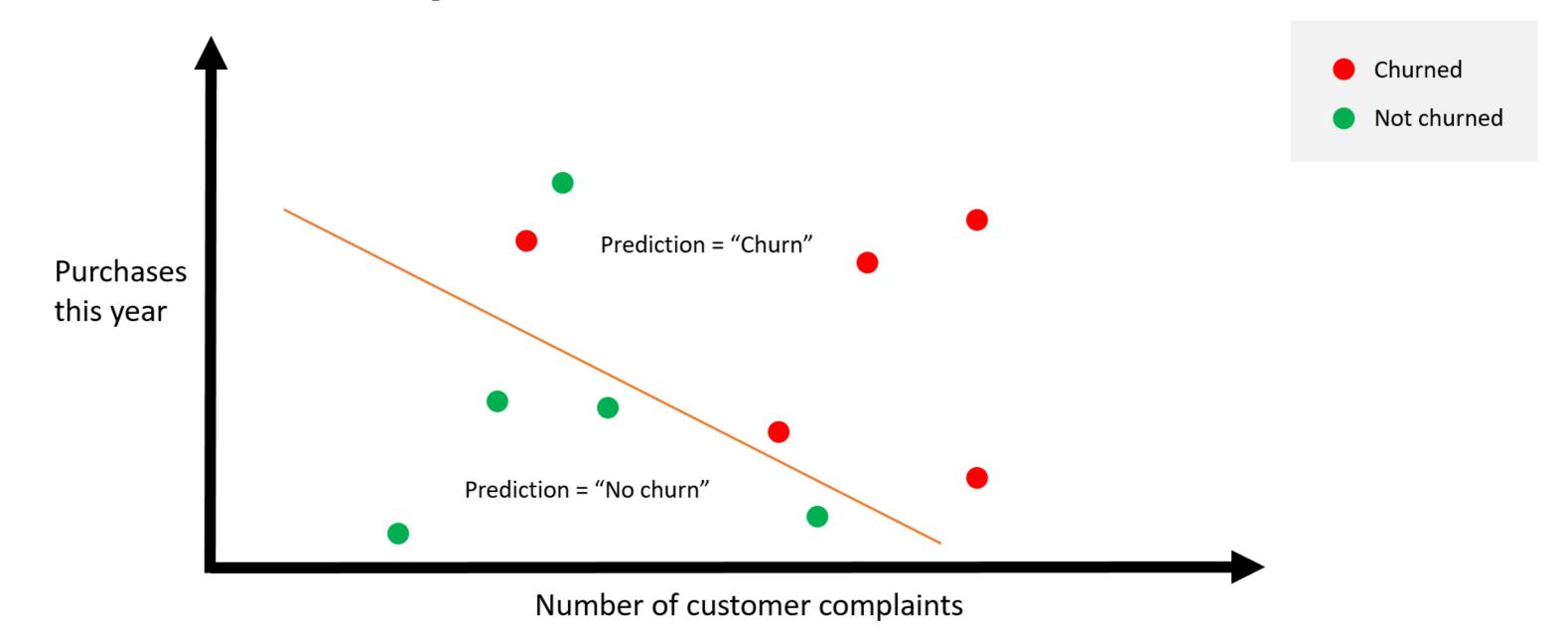
Churn prediction



Mis-classified items

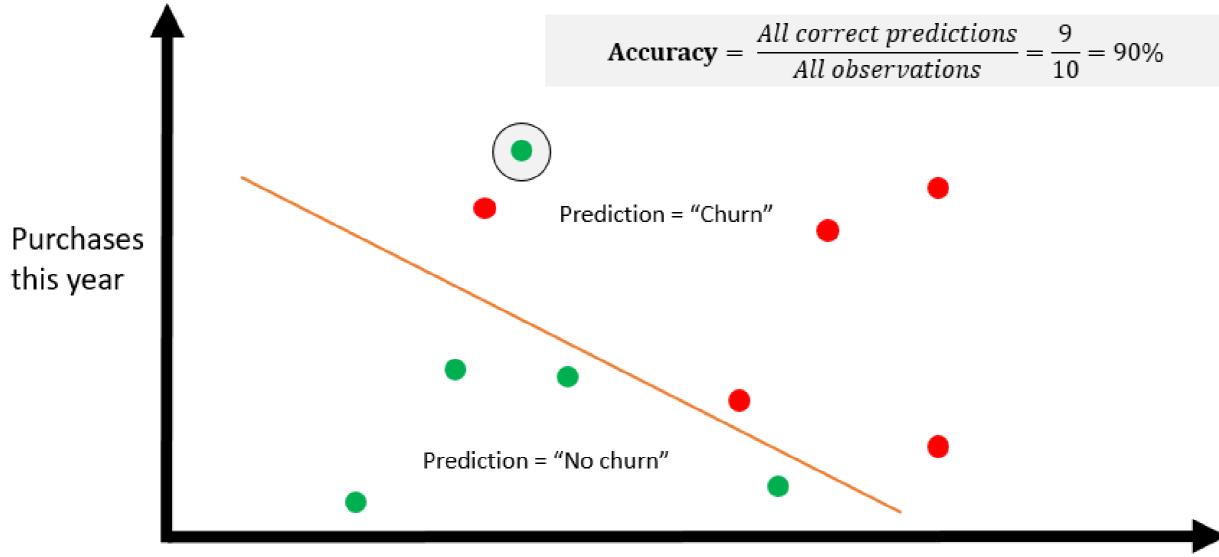


Another churn prediction





Accuracy

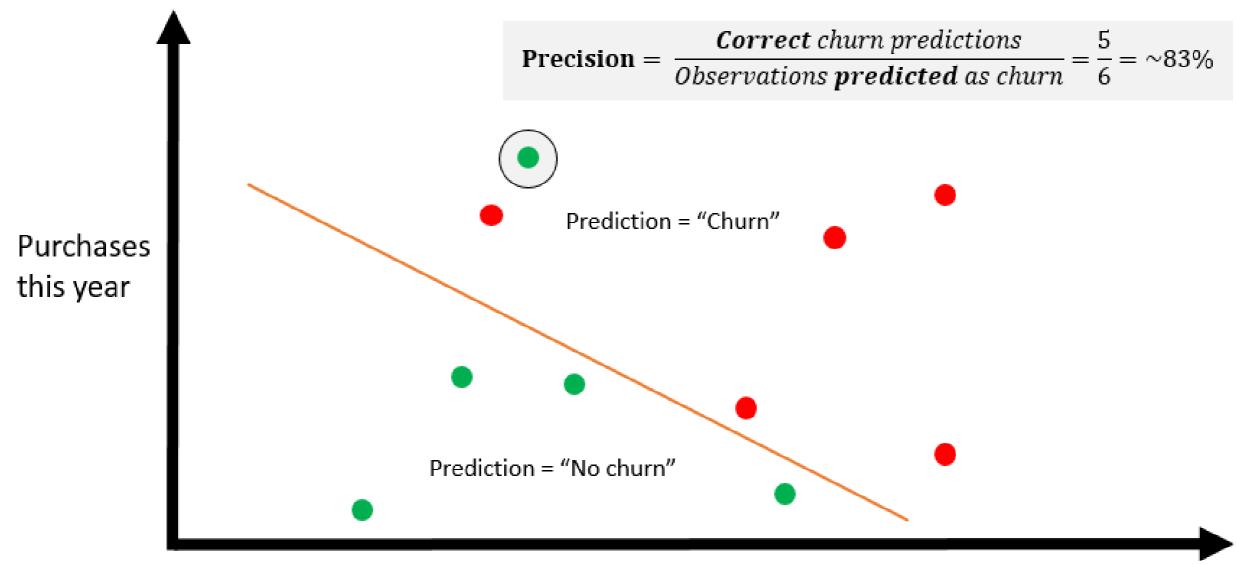


Churned

Not churned

Number of customer complaints

Precision

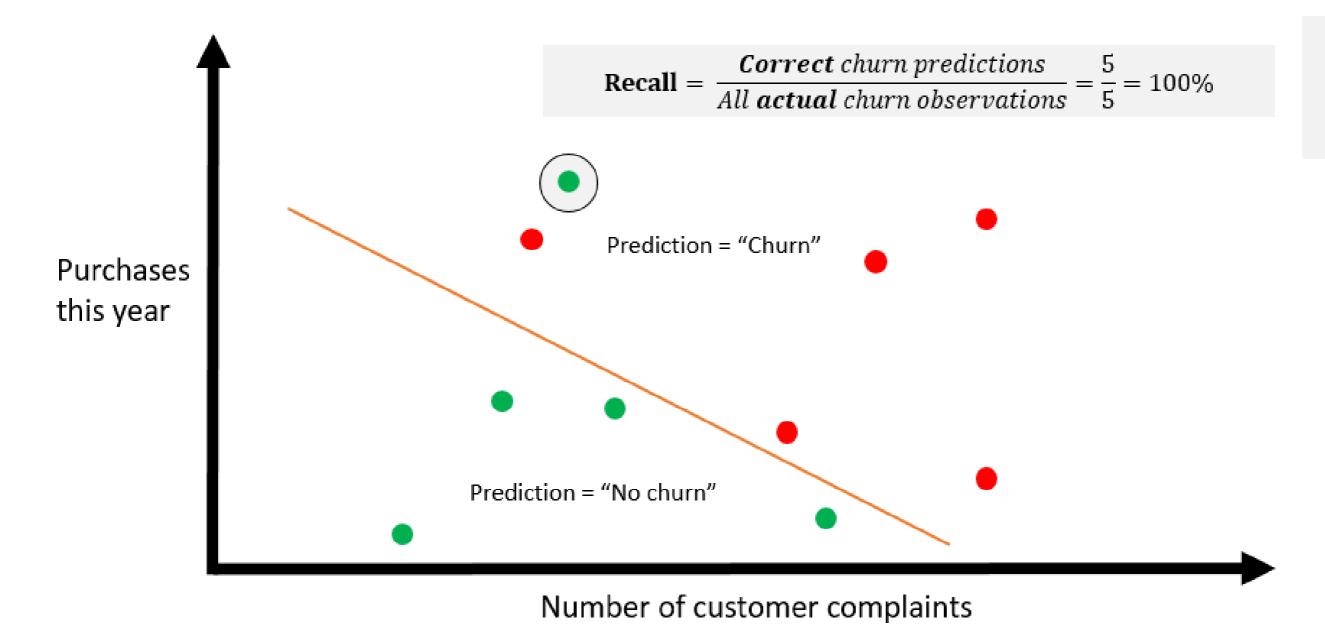


Churned

Not churned

Number of customer complaints

Recall



Churned

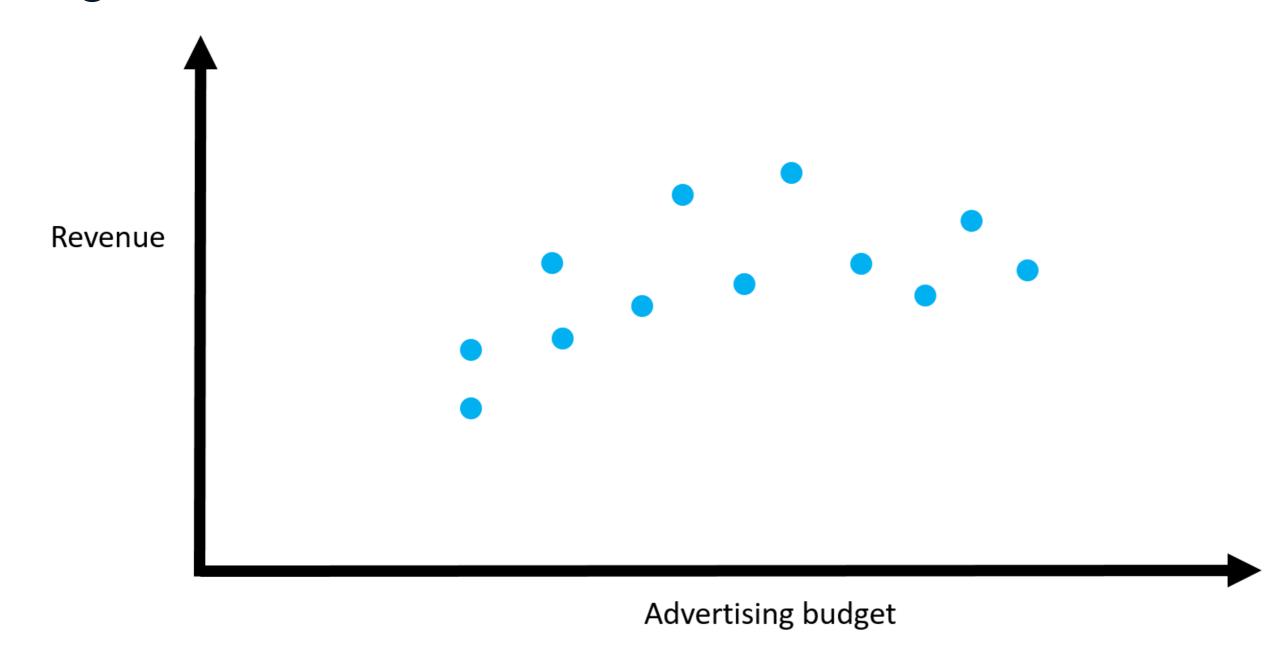
Not churned

Regression performance

• Error

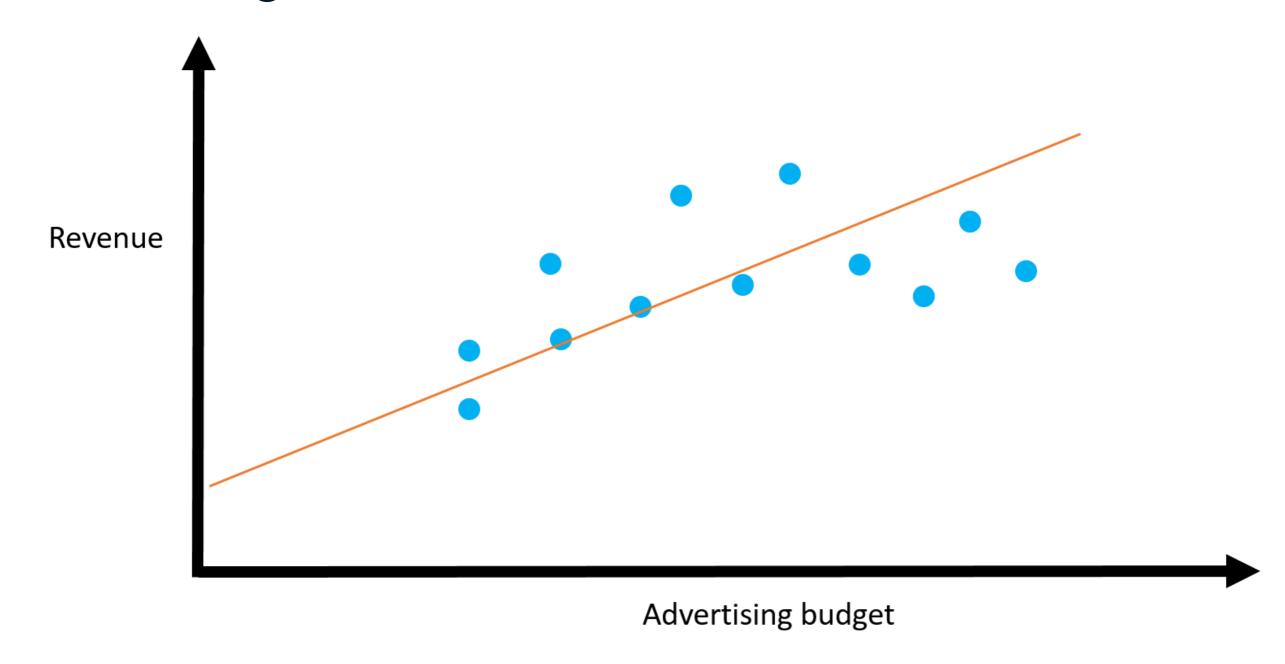


Regression example



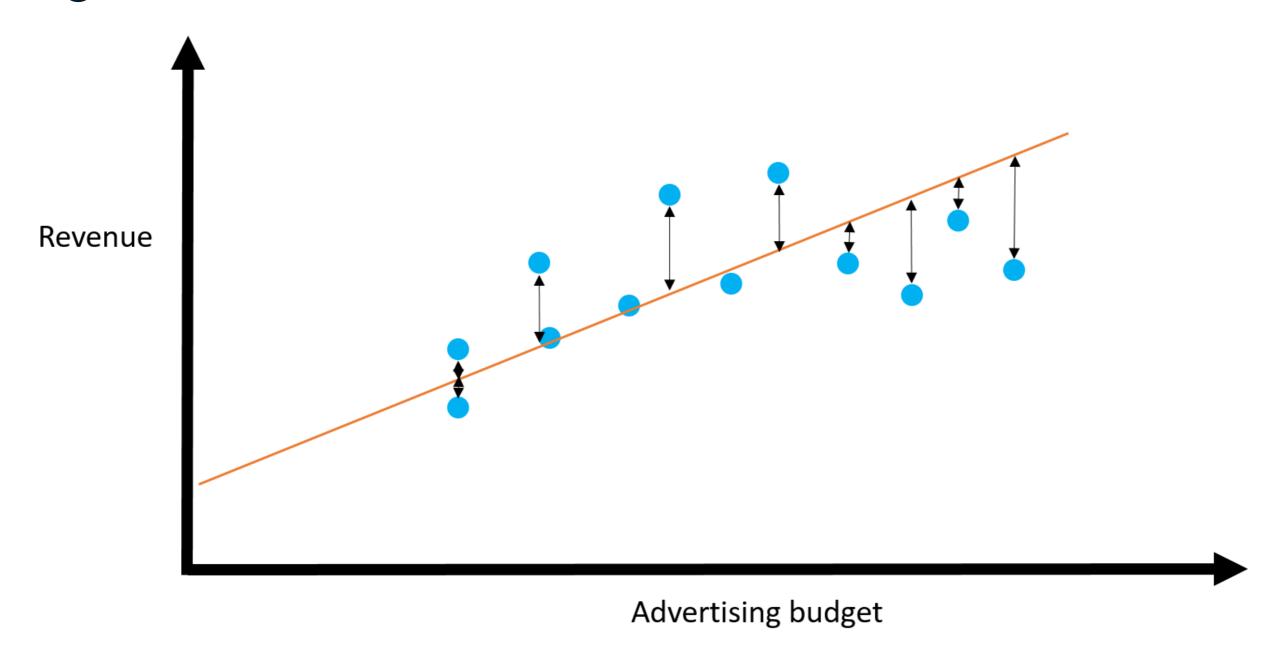


Predicting revenue with a line



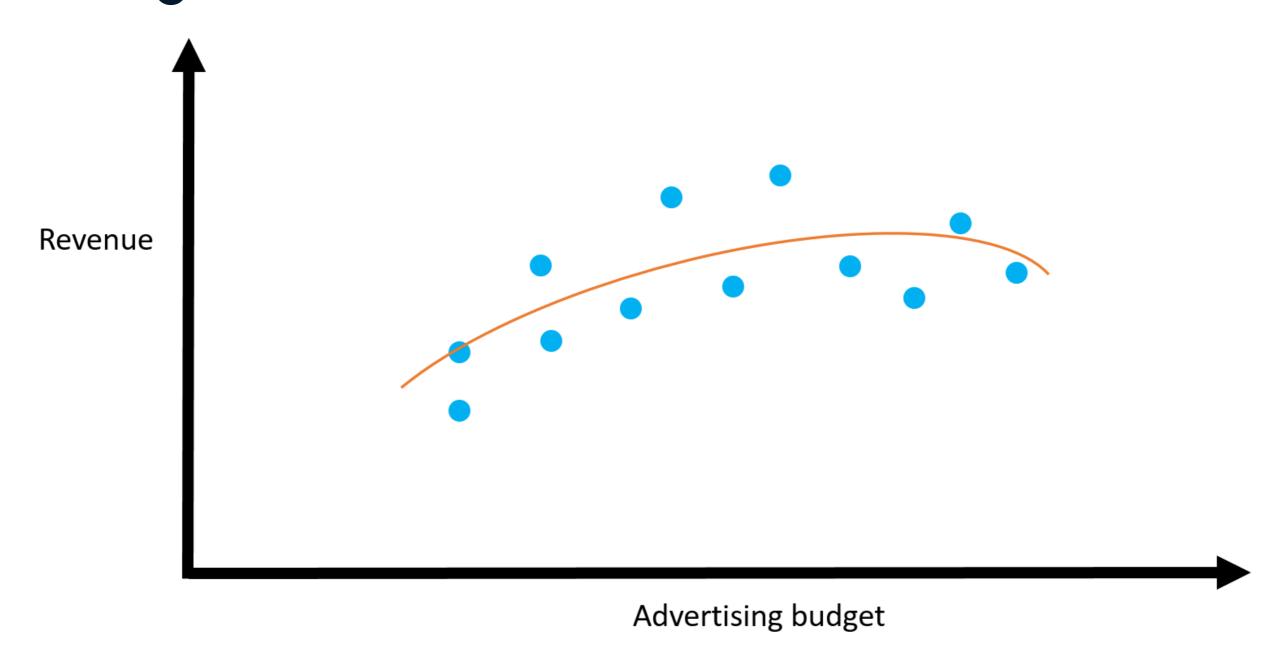


Regression error



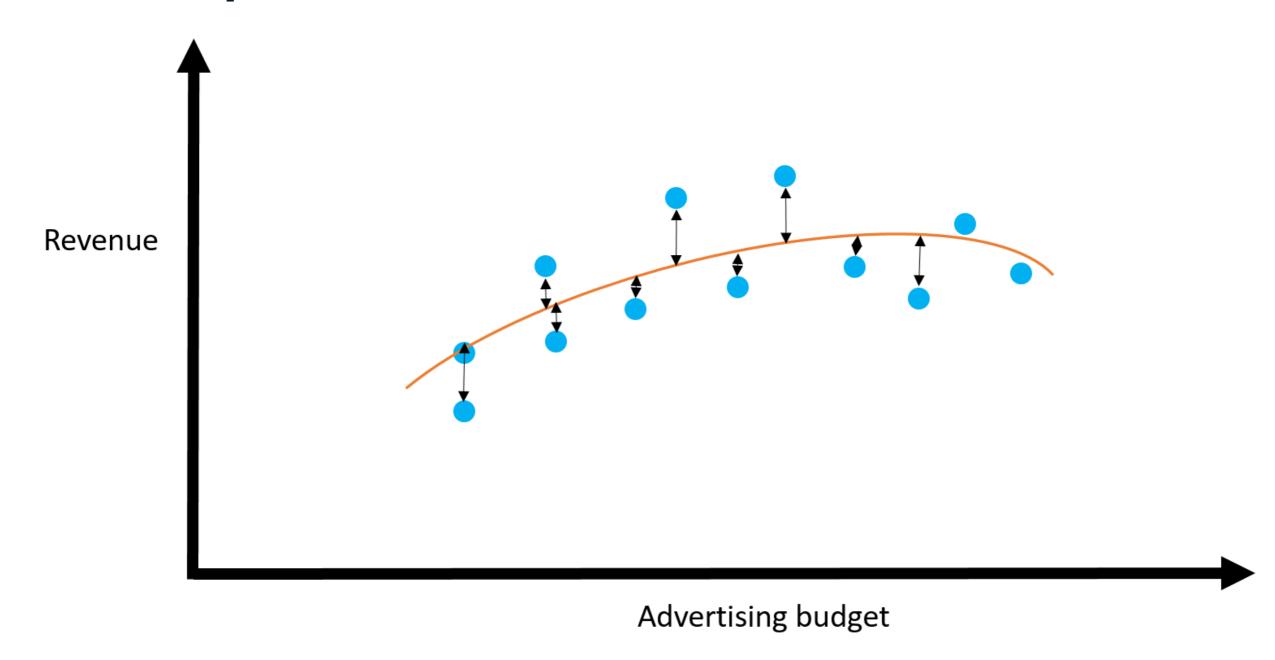


Testing non-linear models





Error improvements





Actionable models - A/B testing

Good models are not always actionable:

Churn prediction, purchase prediction, machine failure prediction

Test if using models helps improve outcomes:

- Target customers predicted to churn with incentives (discounts, coupons, promotions)
- Send reminder emails and product details to customers likely to purchase

_Did this result in decreased churn, increased purchase rate and reduced machine failures? If **yes**, build into automated process. If **no**, collect more data, improve models, and test again. _

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Machine learning risks

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Poor performance

Some models perform poorly (make sure you review test performance, not training):

- Low precision
- Low recall
- Large error

Low precision

Low precision - a lot of misclassified items in the class of interest = a lot of false positives

Example - only 10% of customers identified as likely to purchase actually purchased the product



Low recall

Low recall - only a small fraction of all observations in the class have been correctly captured (recalled) by the model

Example - only 25% of all fraudulent transactions identified by the model



Large error

Large error - large differences between predicted and actual values

Example - the average error for the customer satisfaction rating prediction is 3.5 units or 70% in percentage points



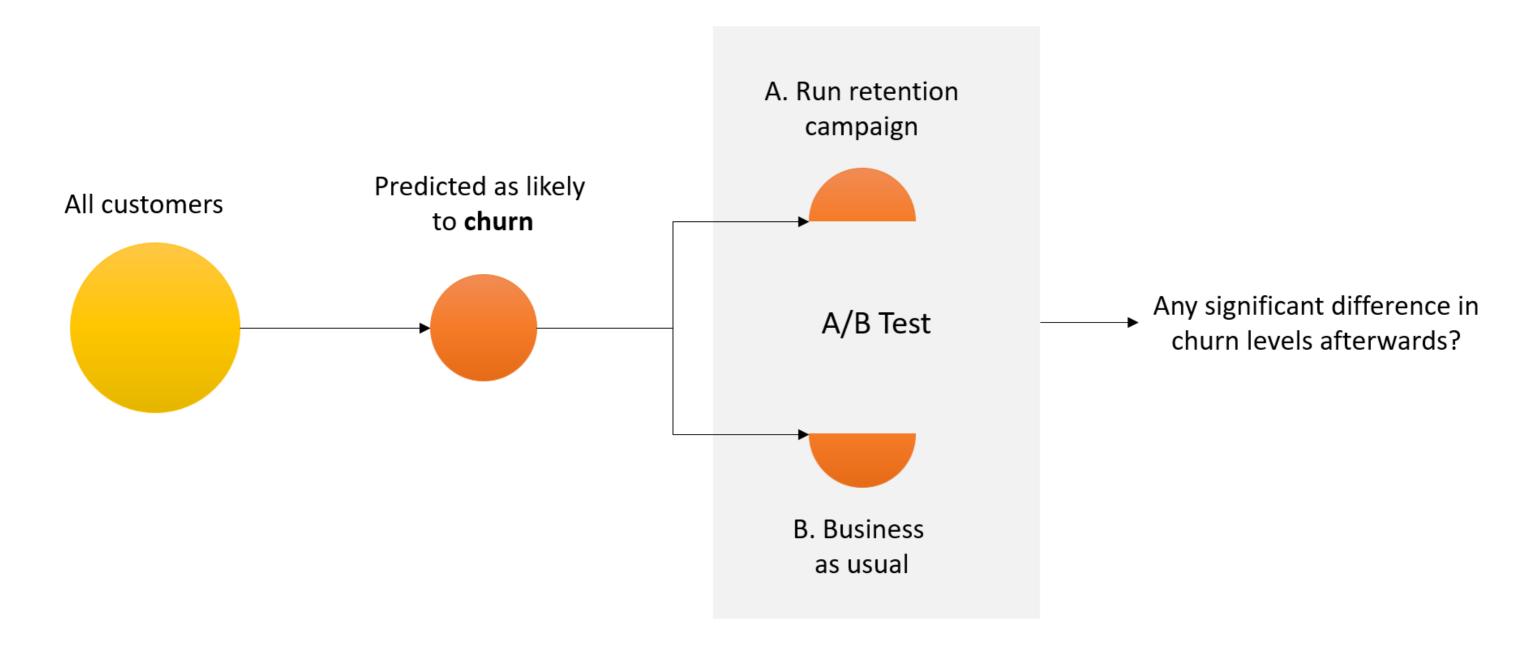
Non-actionable model use cases

Q: How to test the models correctly?

A: Run tests / experiments to validate their performance e.g. churn prevention emails, product promotions, manual machine maintenance, manual transaction review



A/B testing





What if tests don't work?

- Get more data business has to be involved
- Build causal models to understand drivers
- Run qualitative research (surveys etc.)
- Change the scope of the problem
 - Narrow
 - Widen
 - Different question

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