# Business requirements

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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**

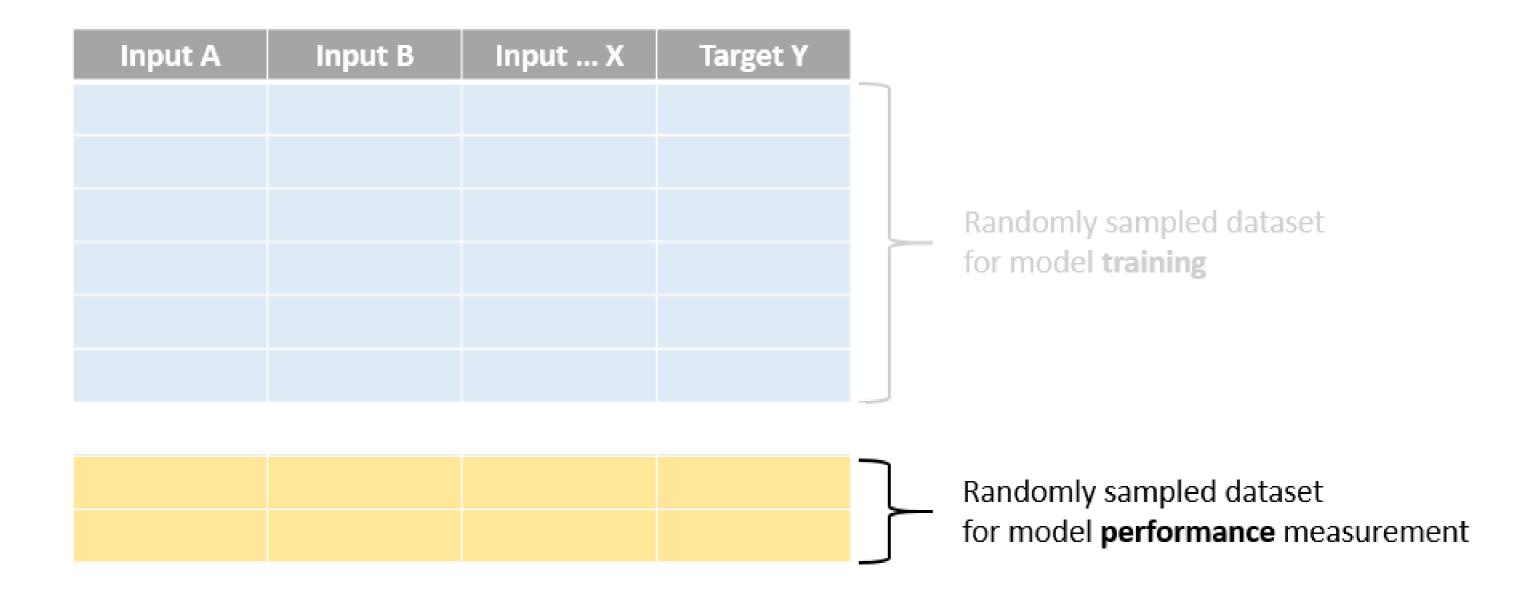
Input A	Input B	Input X	Target Y		
				_	Full dataset

## Splitting data for training

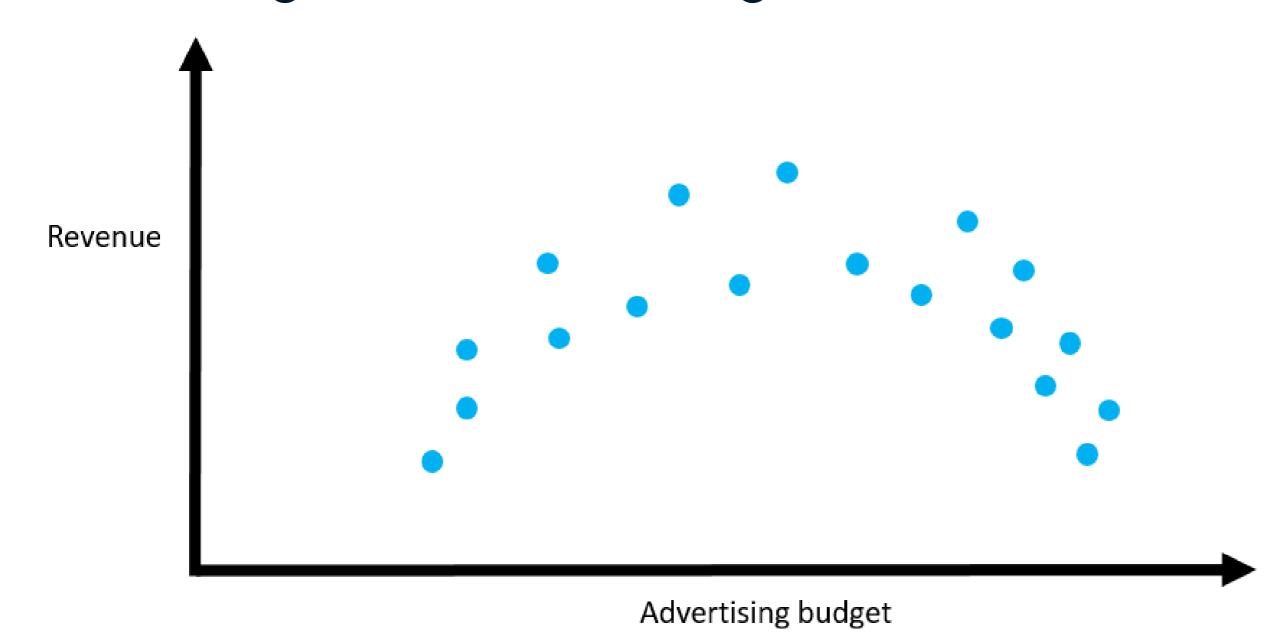
Input A	Input B	Input X	Target Y	
				Randomly sampled dataset
				for model <b>training</b>



#### **Test**

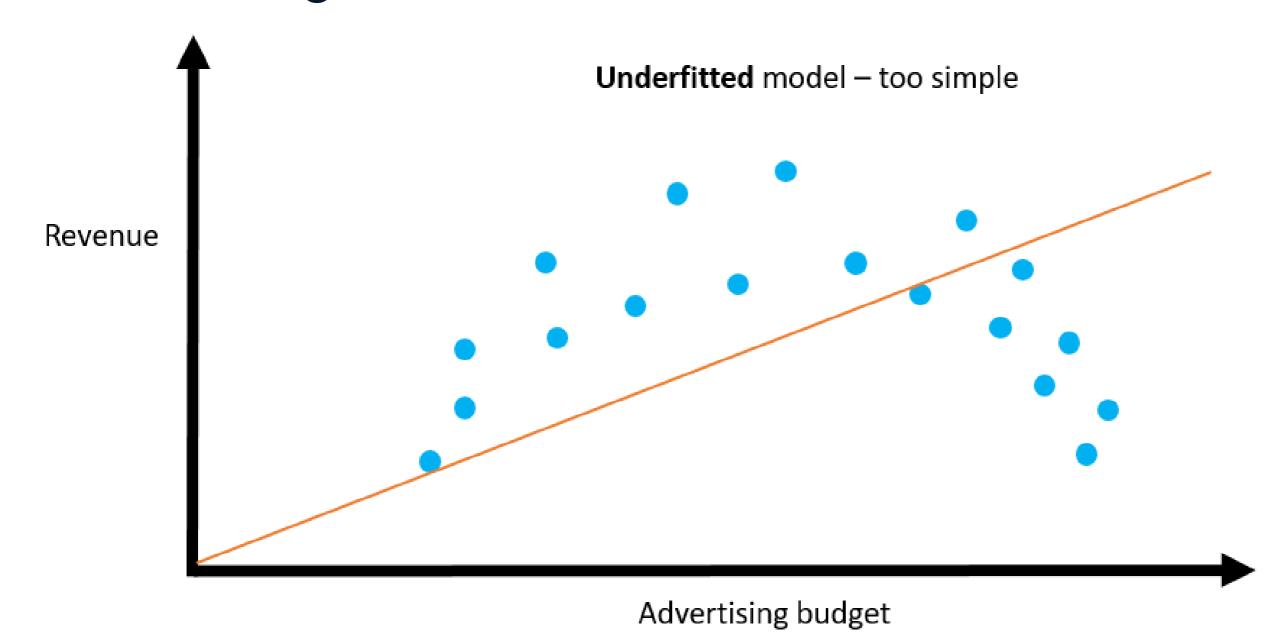


## Overfitting and underfitting



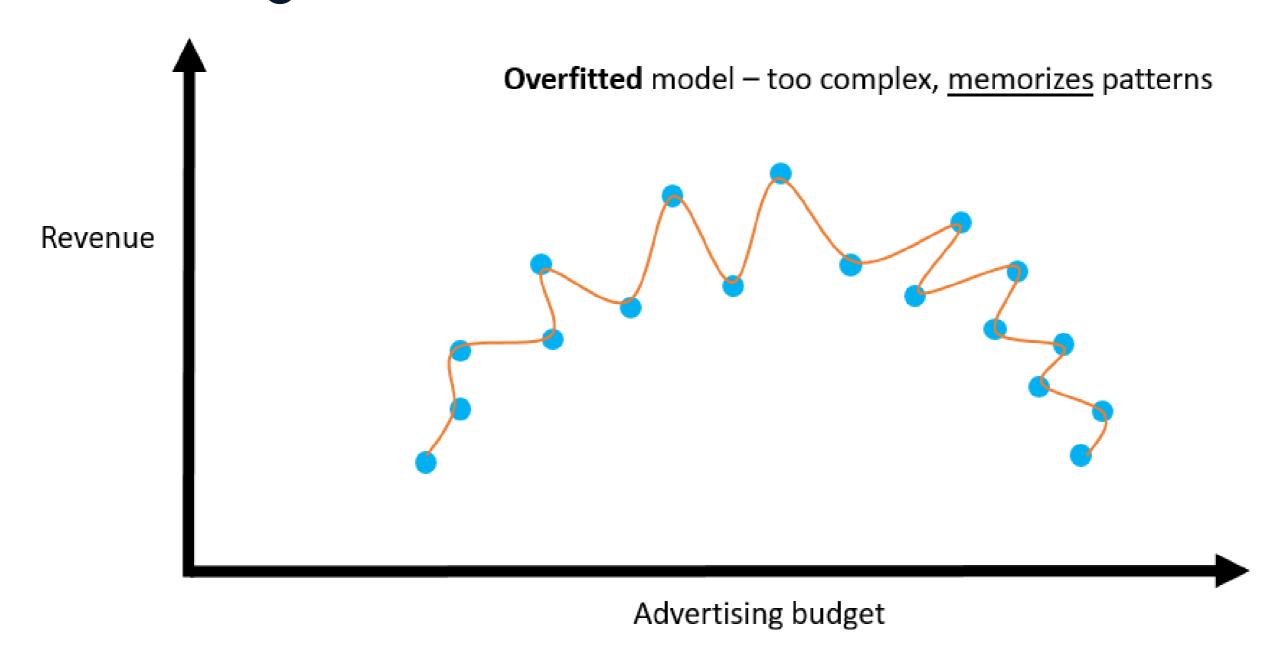


### Underfitting



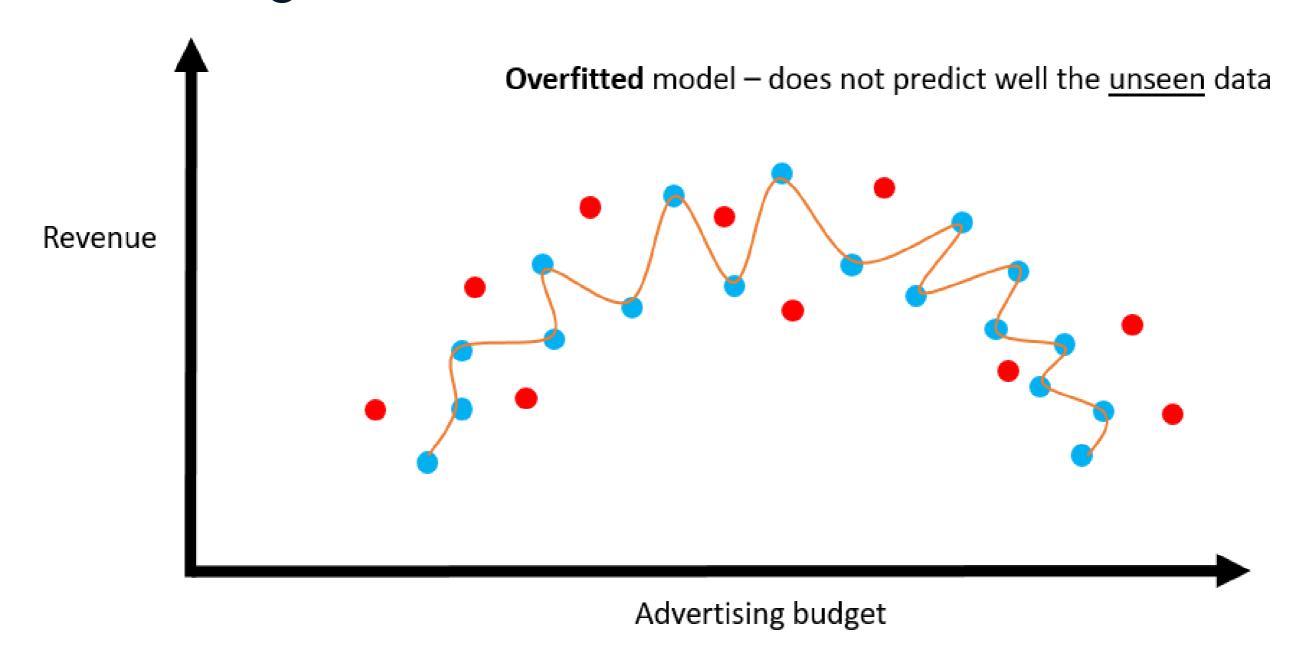


### Overfitting 1



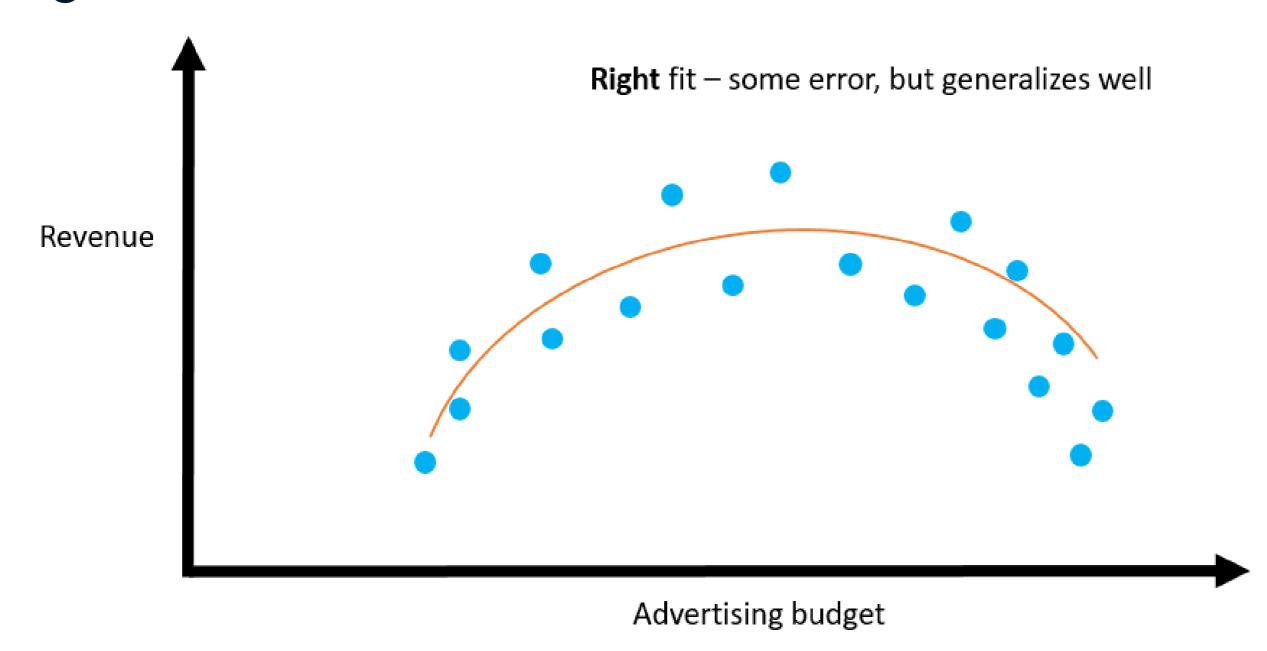


### Overfitting 2



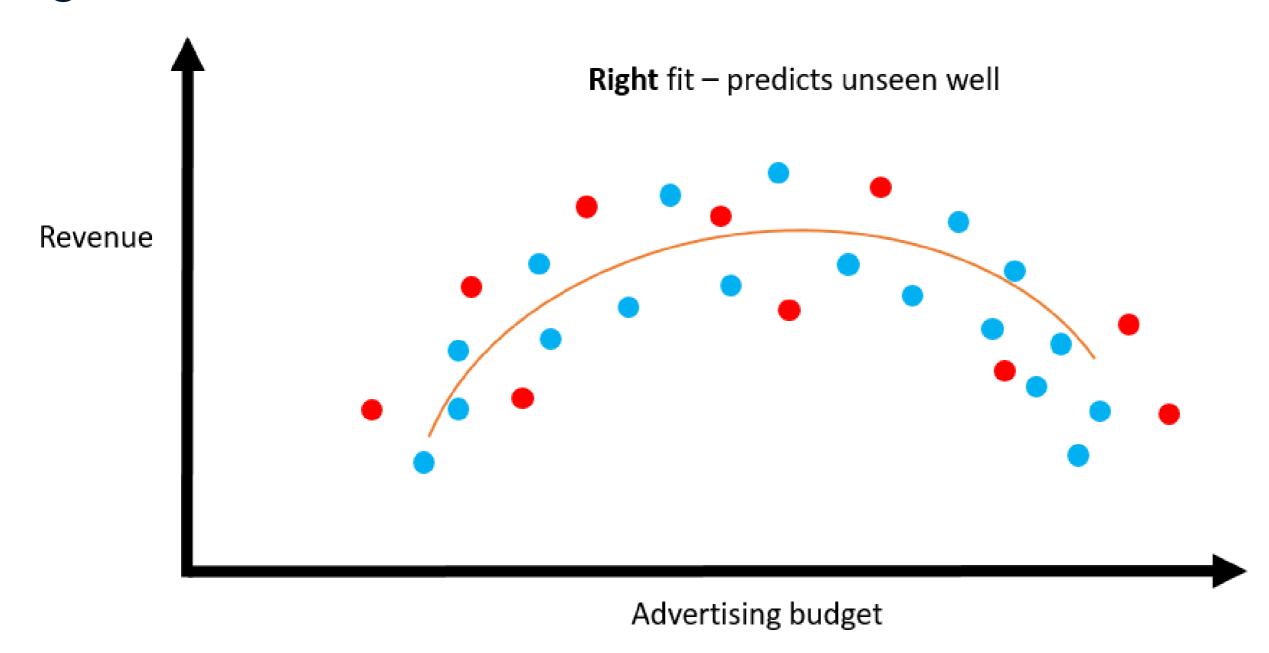


#### Right model fit 1



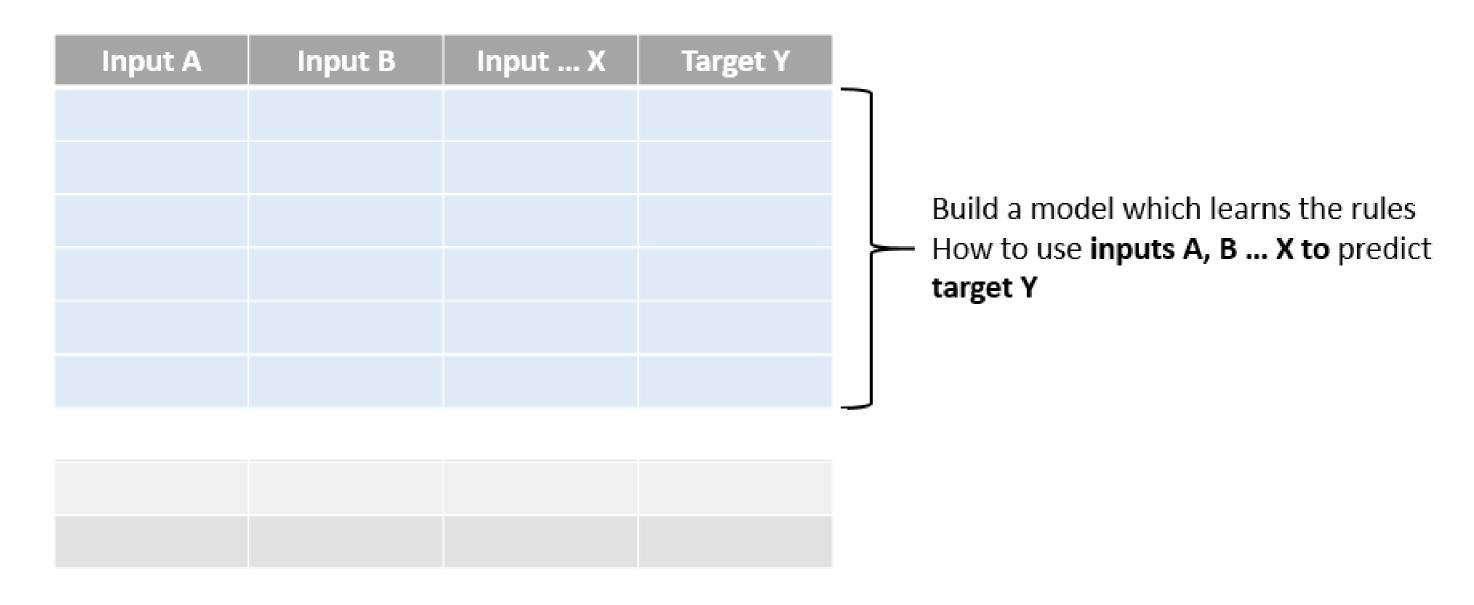


## Right model fit 2

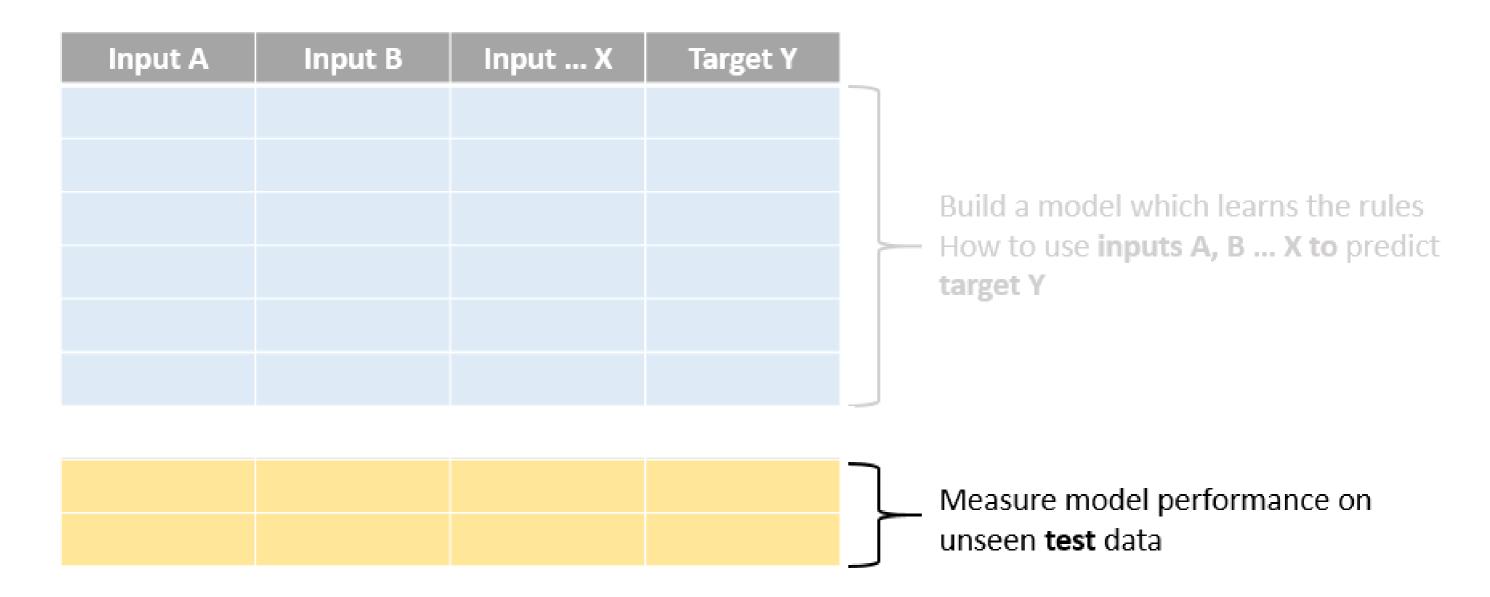




### **Model training**

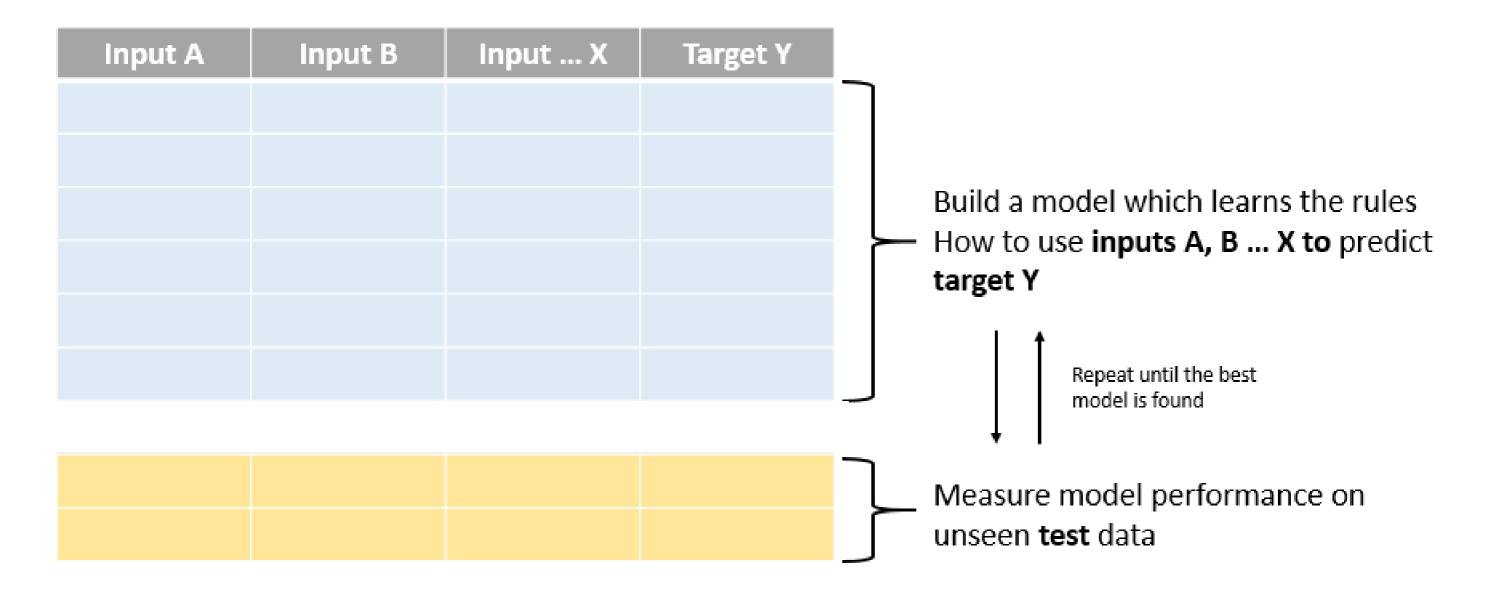


#### Assess model performance on test





#### Try a few models



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# Model performance measurement

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#### Performance measurement types

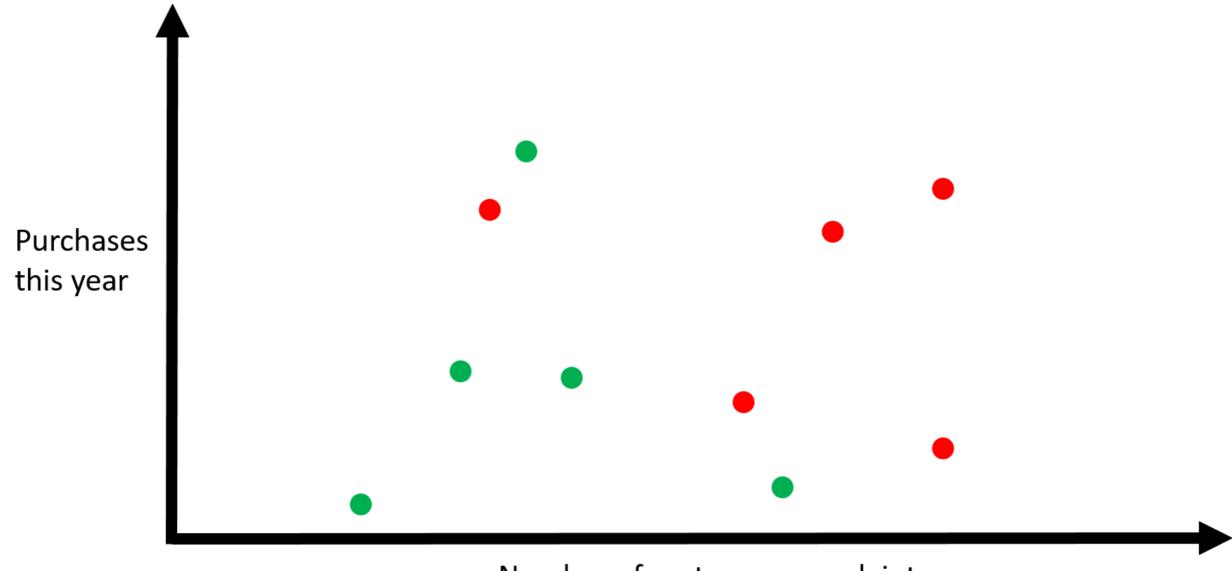
Two key supervised learning metrics:

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

#### Classification performance

- Accuracy
- Recall
- Precision

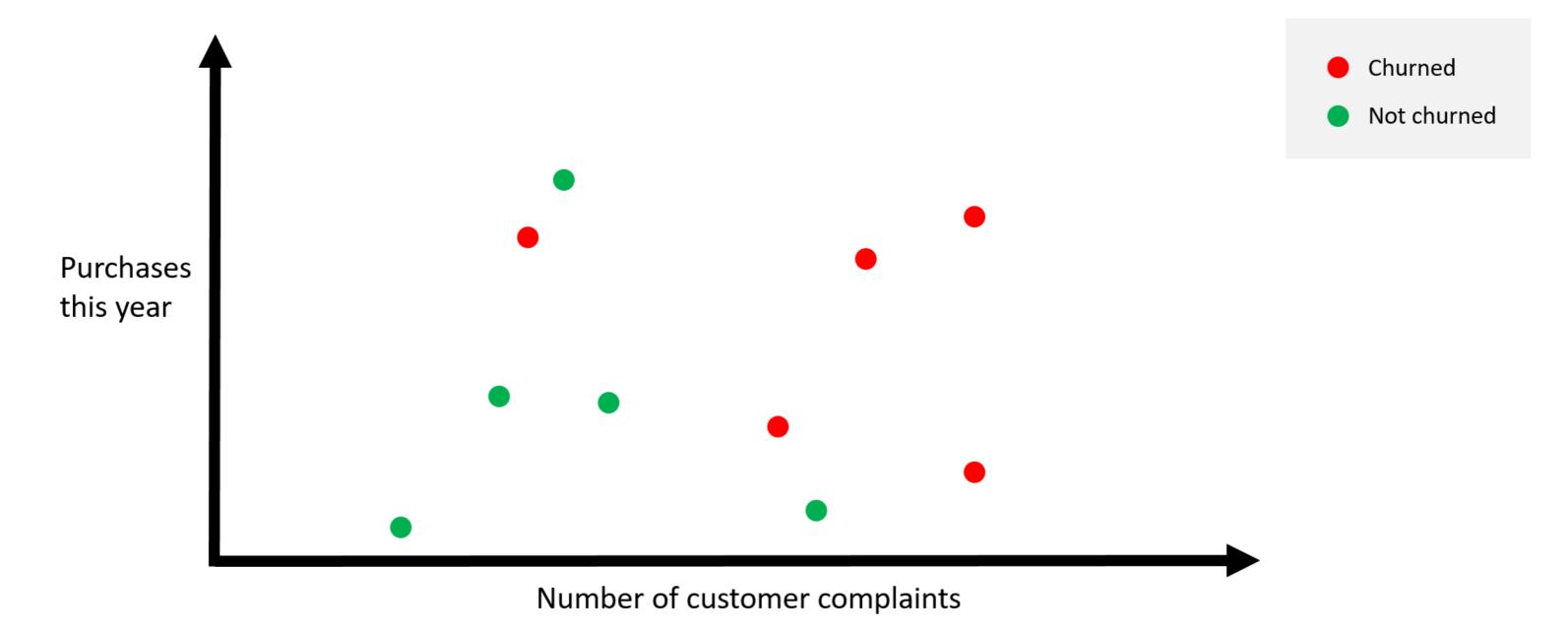
#### Churn example



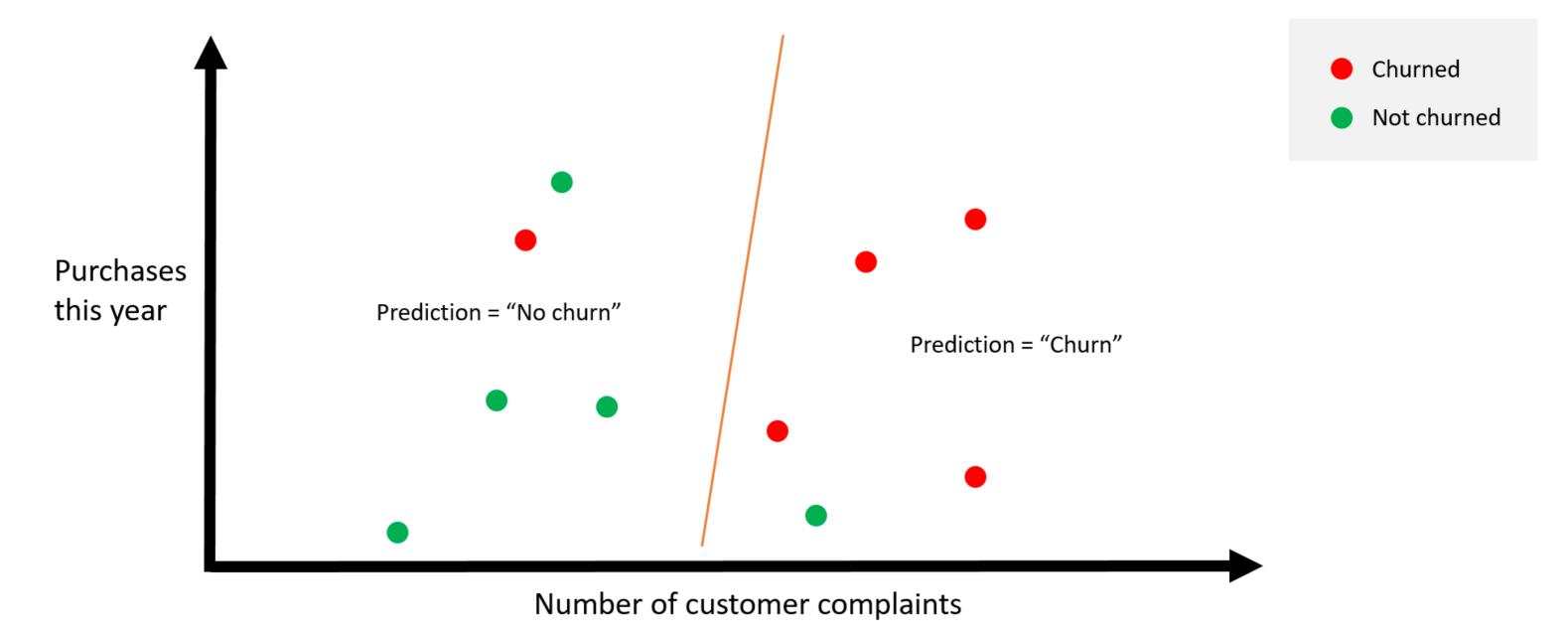
Number of customer complaints



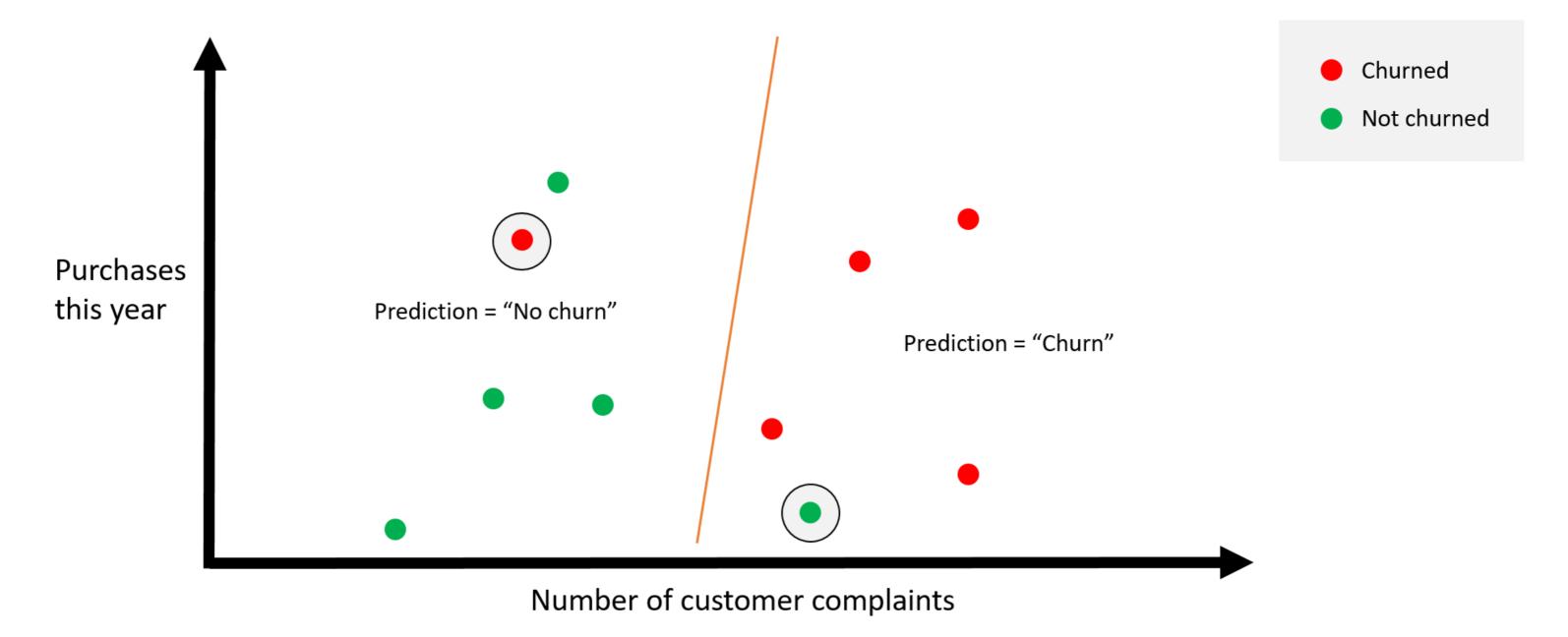
#### Churn example



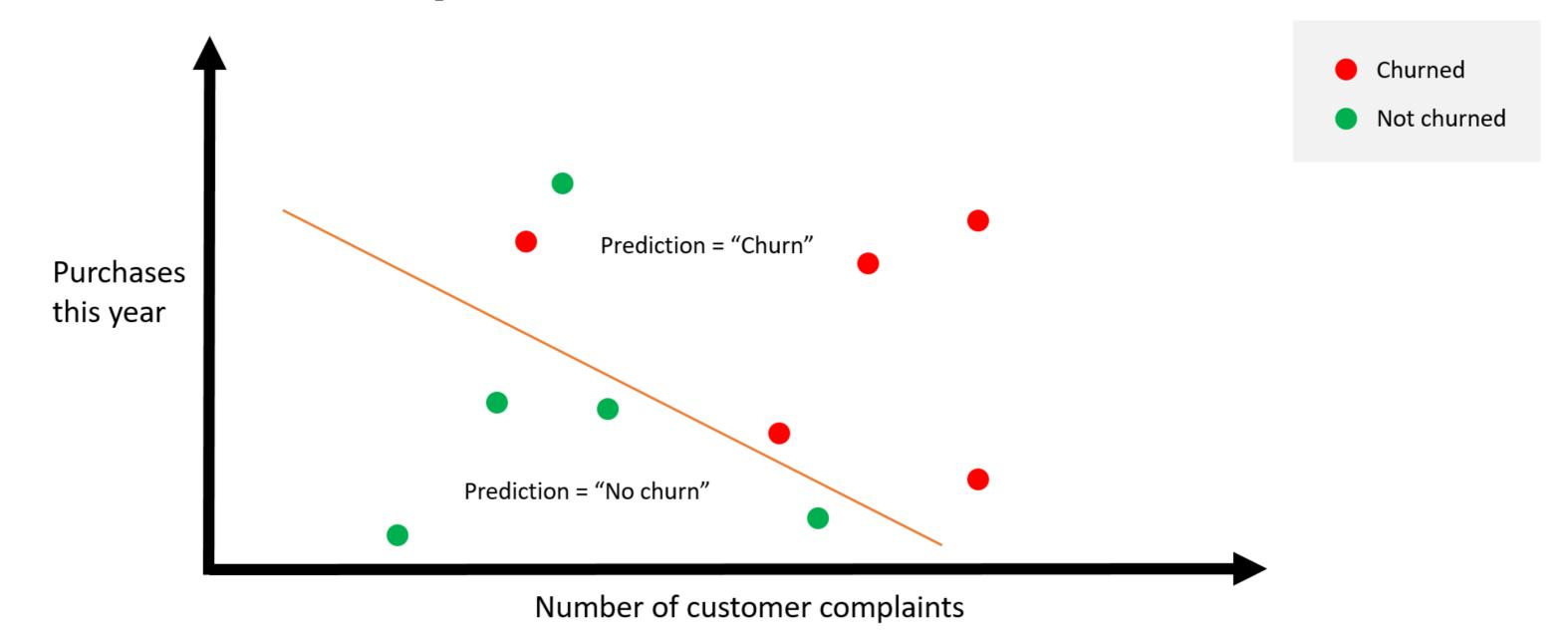
#### Churn prediction



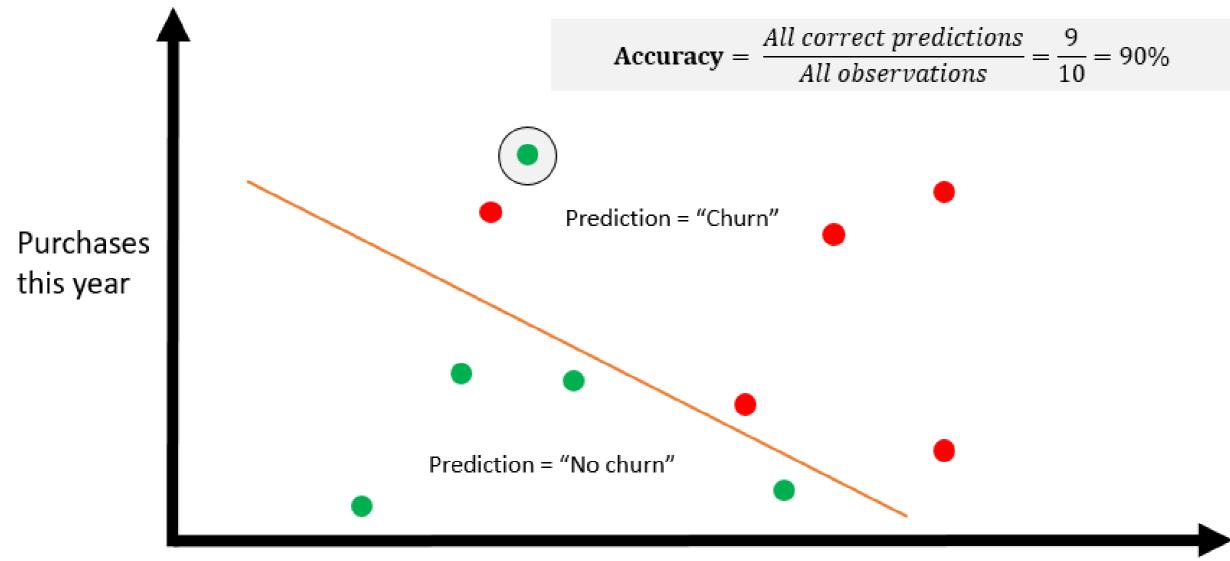
#### Mis-classified items



#### Another churn prediction



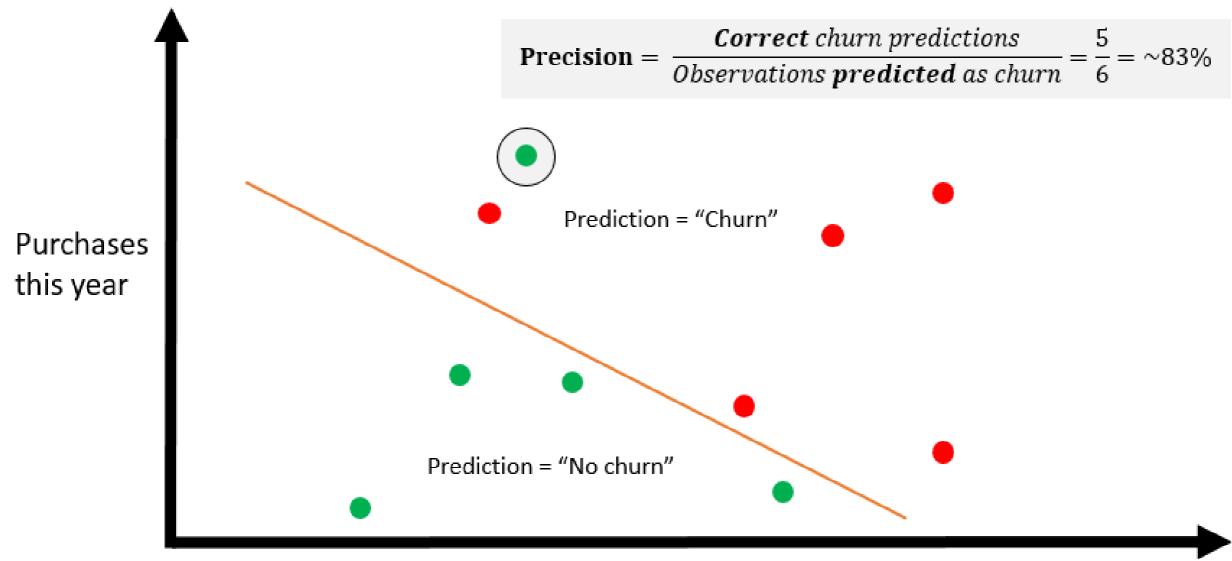
#### Accuracy



Churned

Not churned

#### Precision

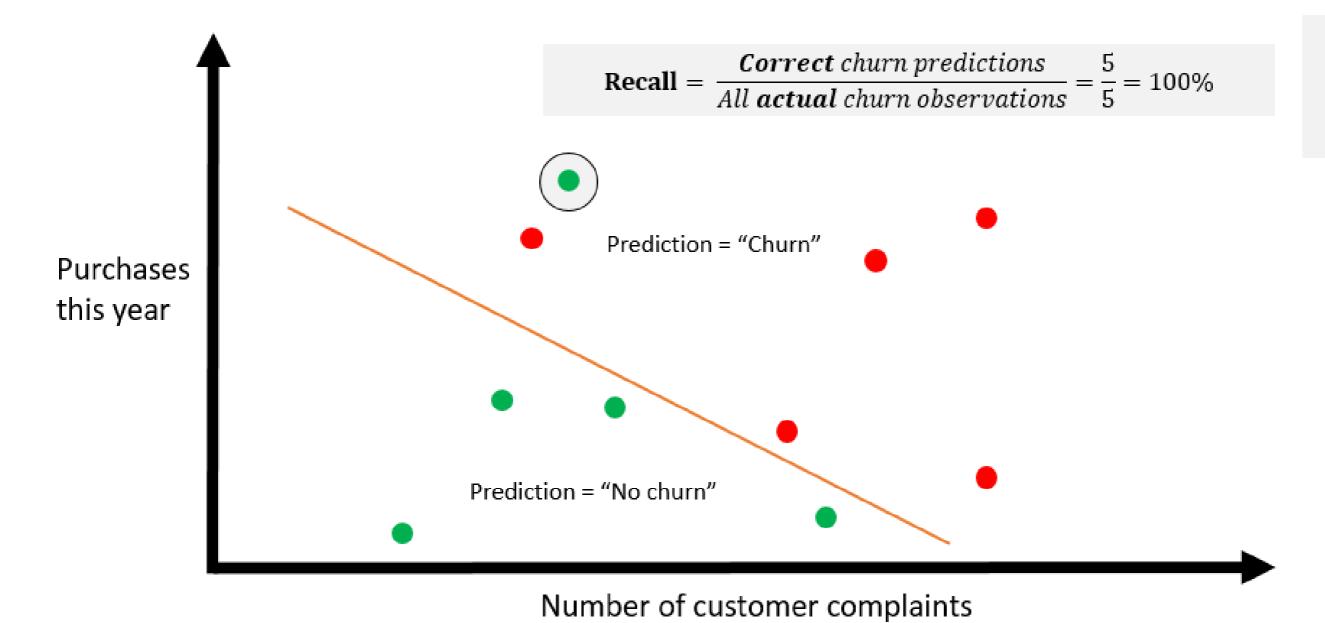


Churned

Not churned

Number of customer complaints

#### Recall



Churned

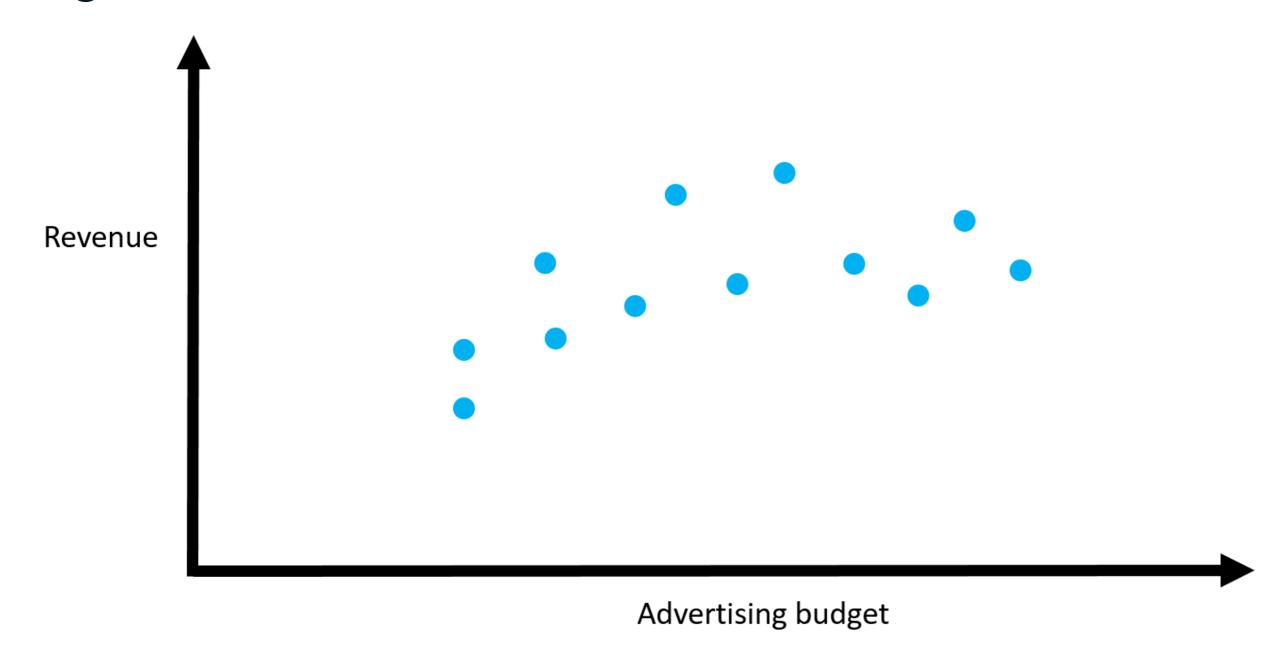
Not churned

### Regression performance

• Error

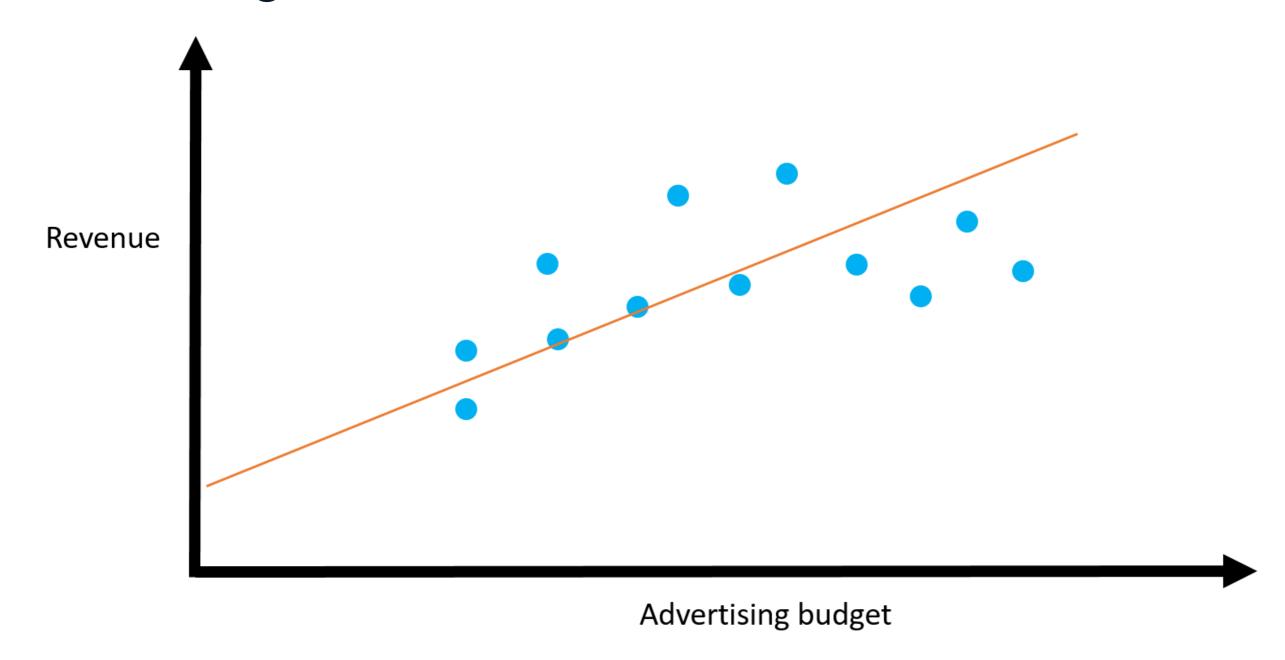


### Regression example



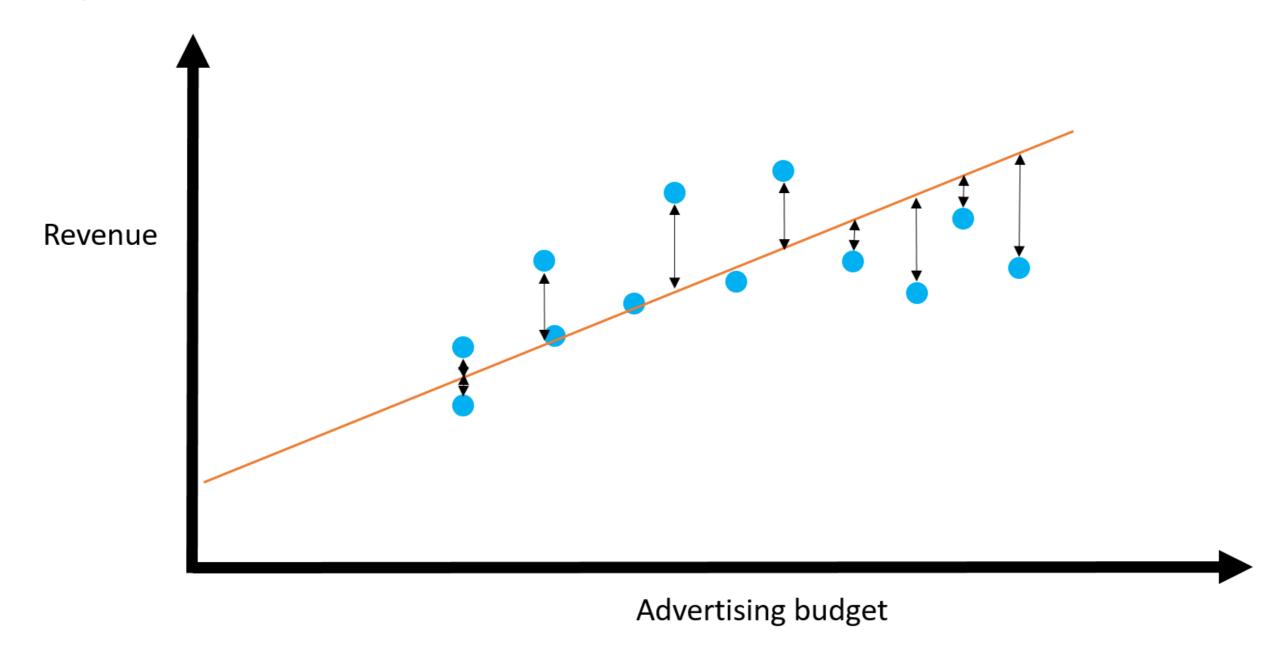


## Predicting revenue with a line

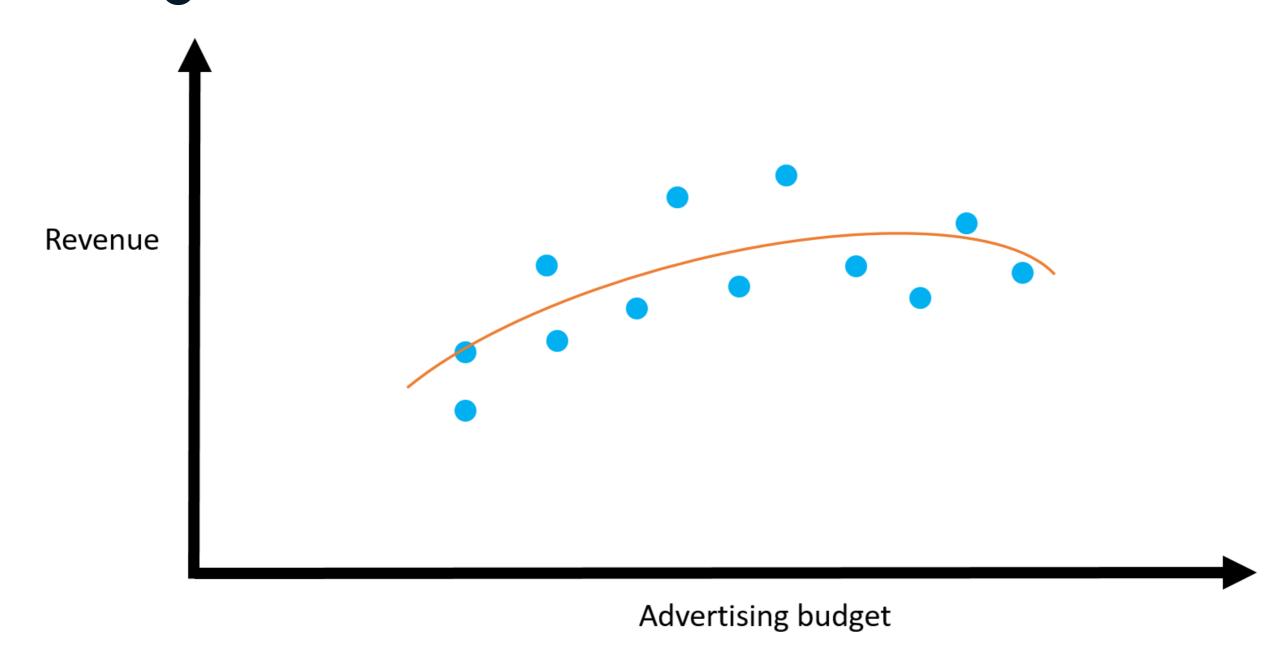




## Regression error

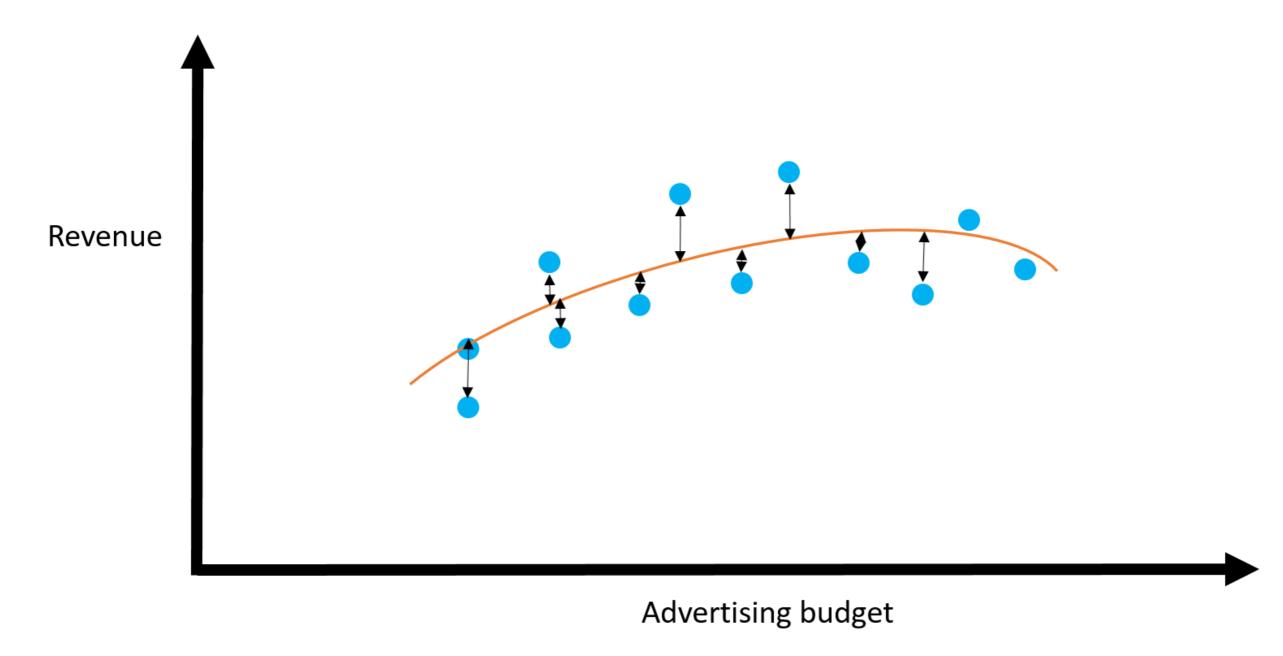


## Testing non-linear models





## **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



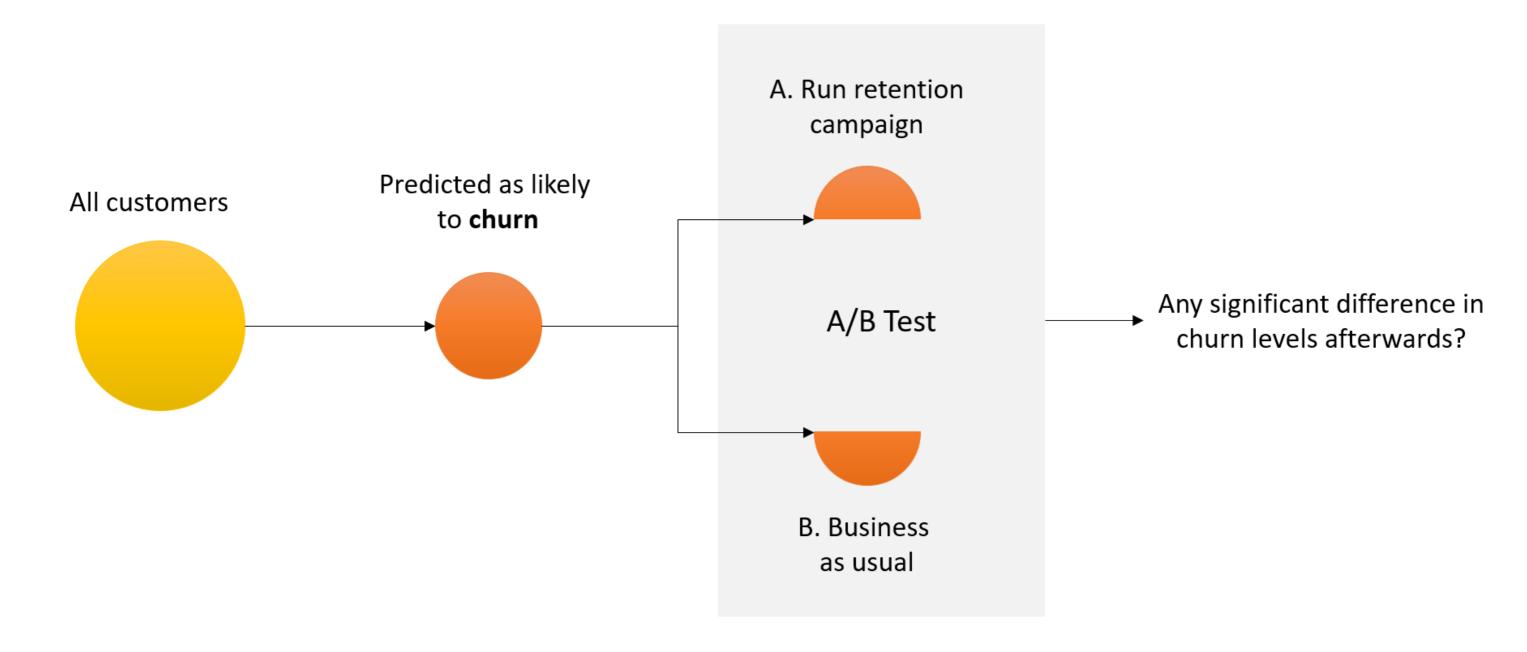
#### Non-actionable model use cases

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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