



Synthetic tabular data



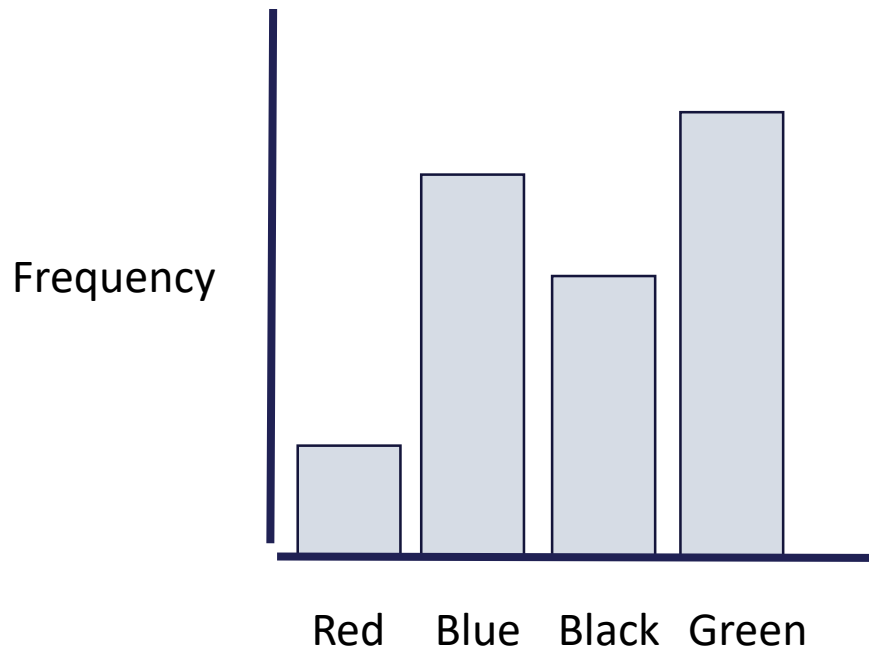


• Tabular data

Colour	Age	Income	Car make	Nr. Cards
Red	65	50000	Ford	5

• Tabular data – categorical variables

Colour	Age	Income	Car make	Nr. Cards
Red	65	50000	Ford	5



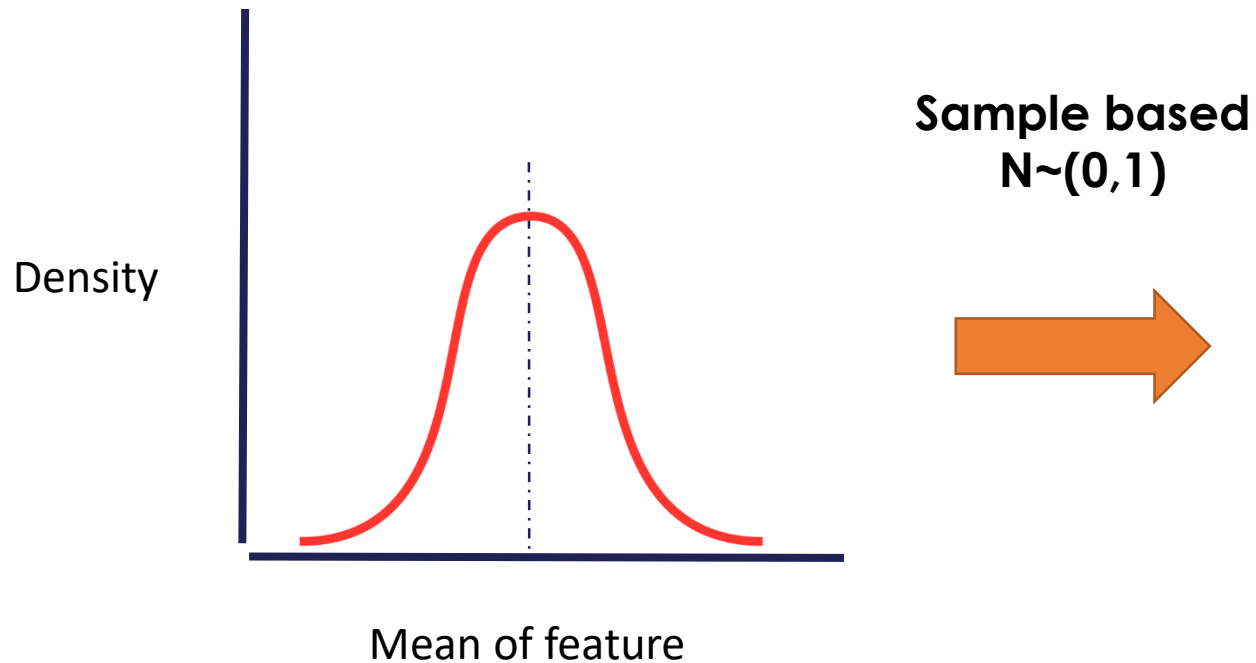
Sample based
on frequency



Colour	Age	Income	Car make	Nr. Cards
Blue			Ford	
Red			Fiat	
Black			Ford	
Green			VW	
Green			Ford	

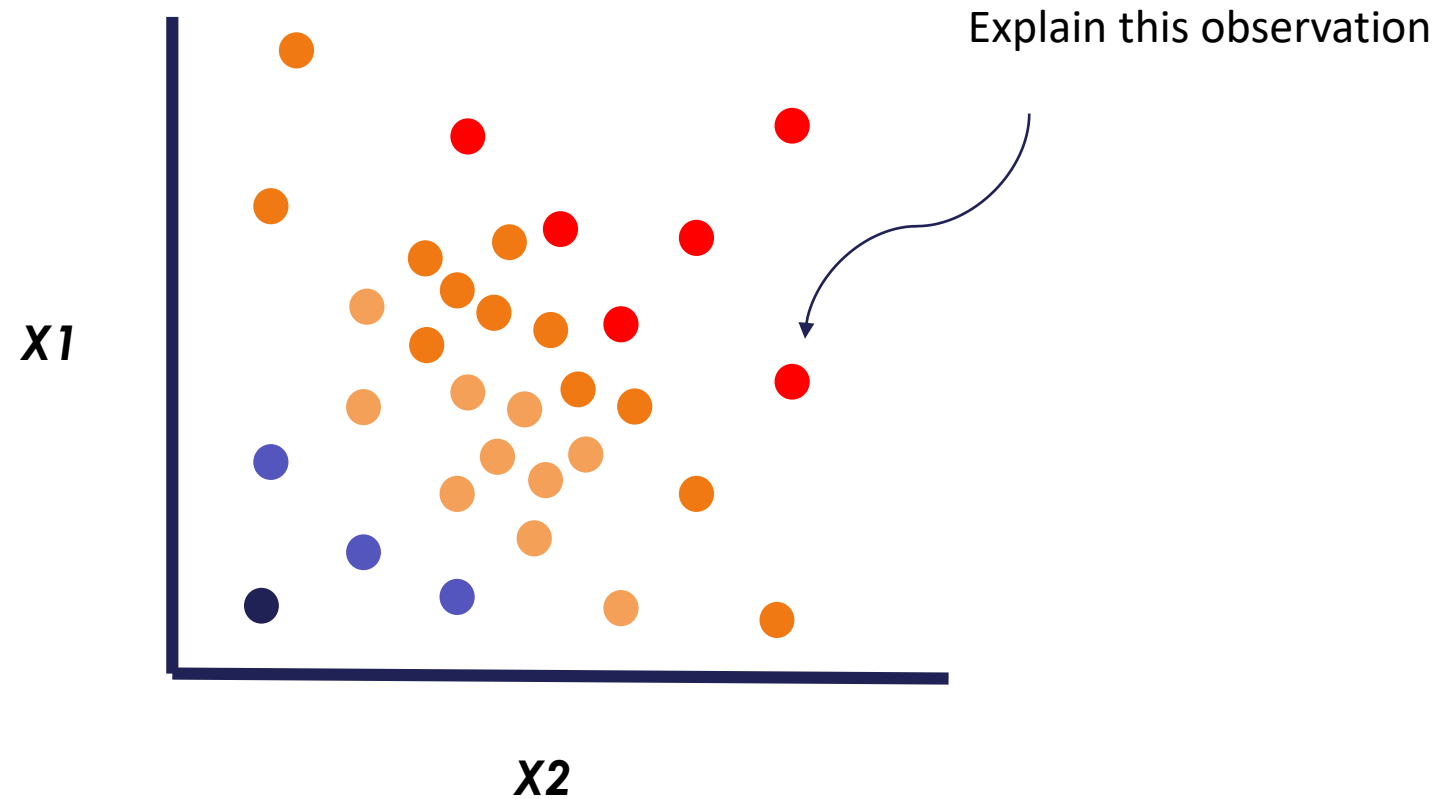
• Tabular data – numerical variables

Colour	Age	Income	Car make	Nr. Cards
Red	65	50000	Ford	5

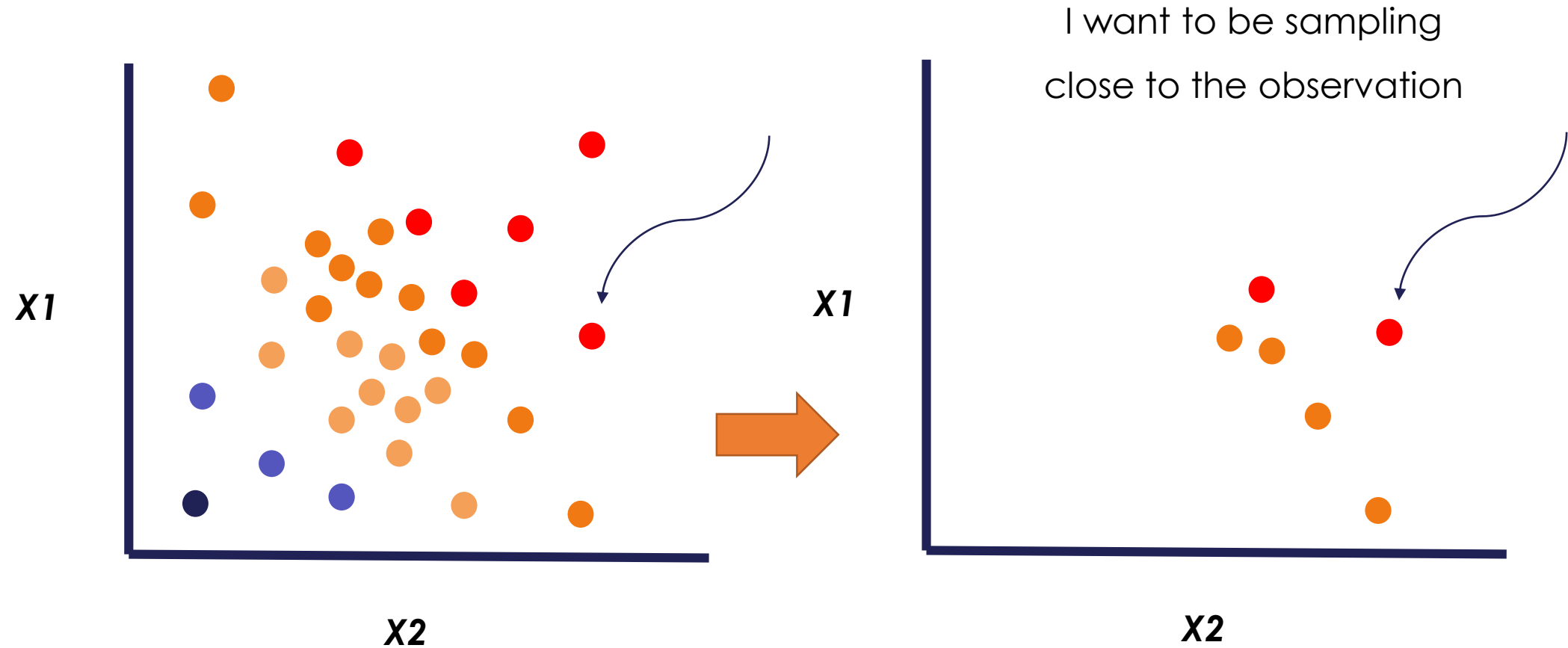


Colour	Age	Income	Car make	Nr. Cards
Blue	65	51000	Ford	5
Red	63	59000	Fiat	5.2
Black	60	55000	Ford	5.9
Green	58	45000	VW	6
Green	55	47000	Ford	4

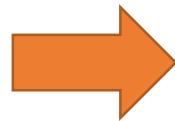
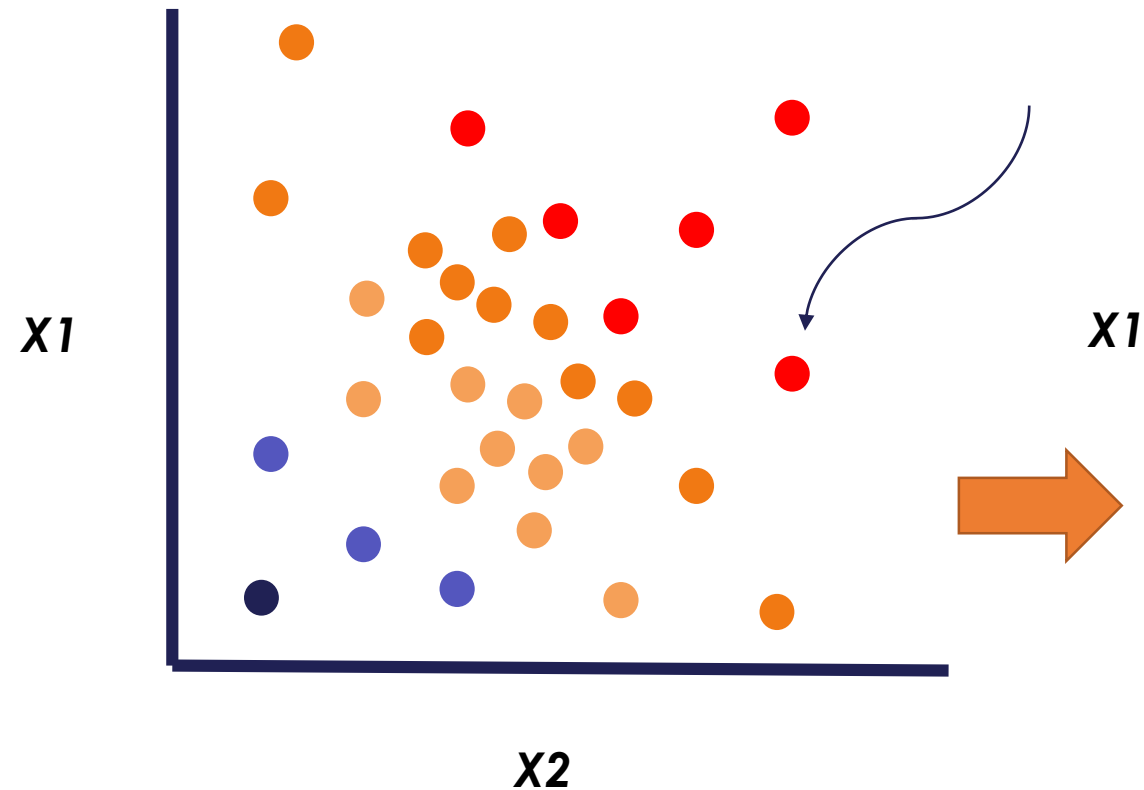
• Tabular data - problematic



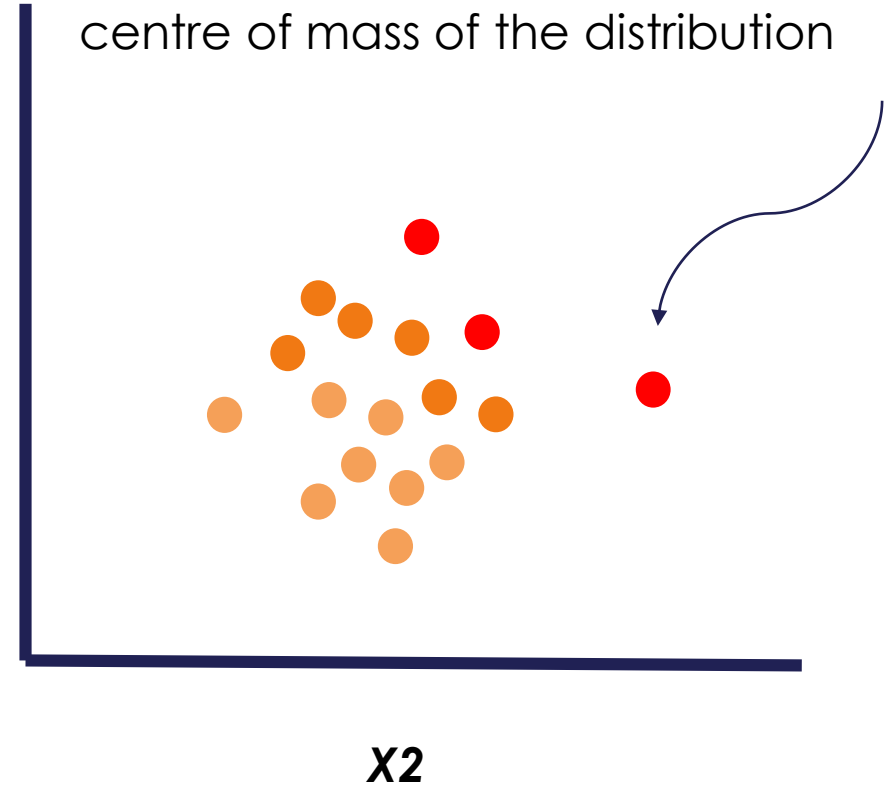
• Tabular data - problematic



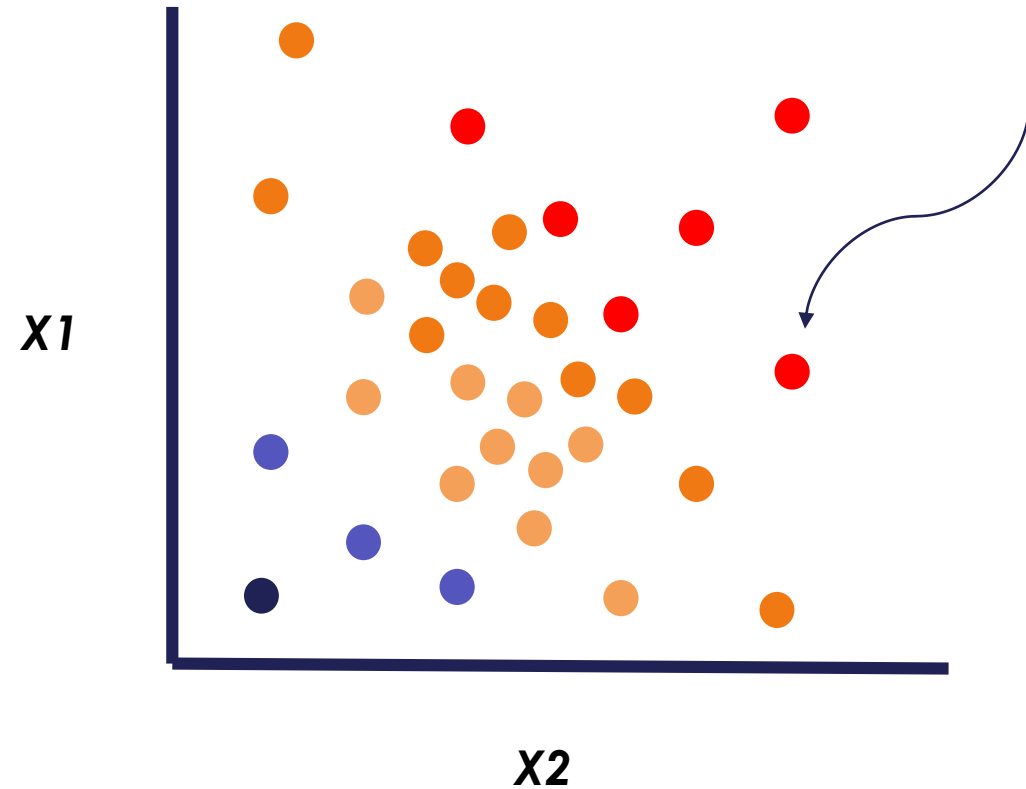
• Tabular data - problematic



Yet, I will be sampling at the
centre of mass of the distribution

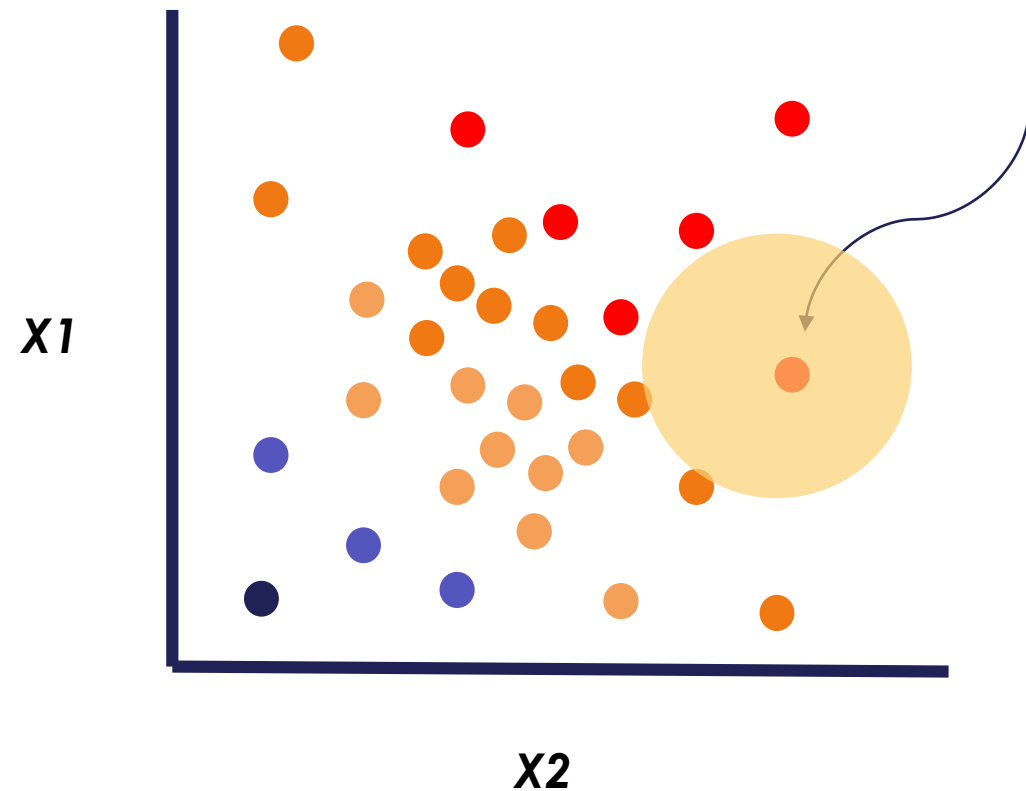


• Tabular data - problematic



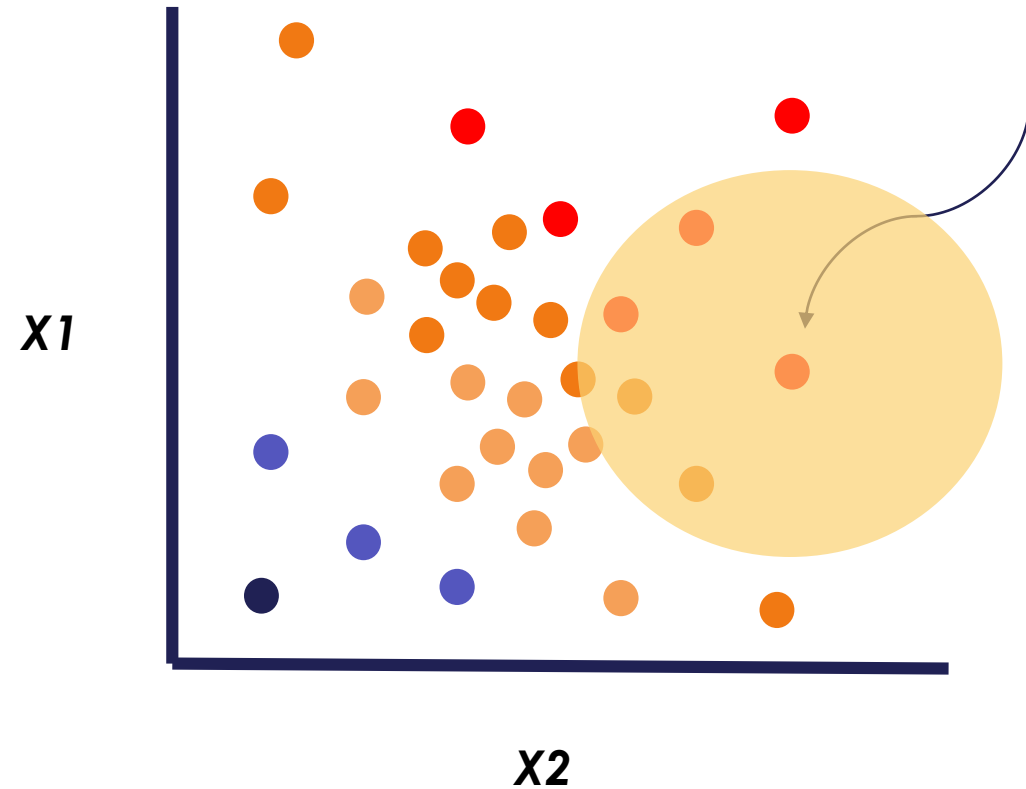
What is vicinity?

• Tabular data - problematic



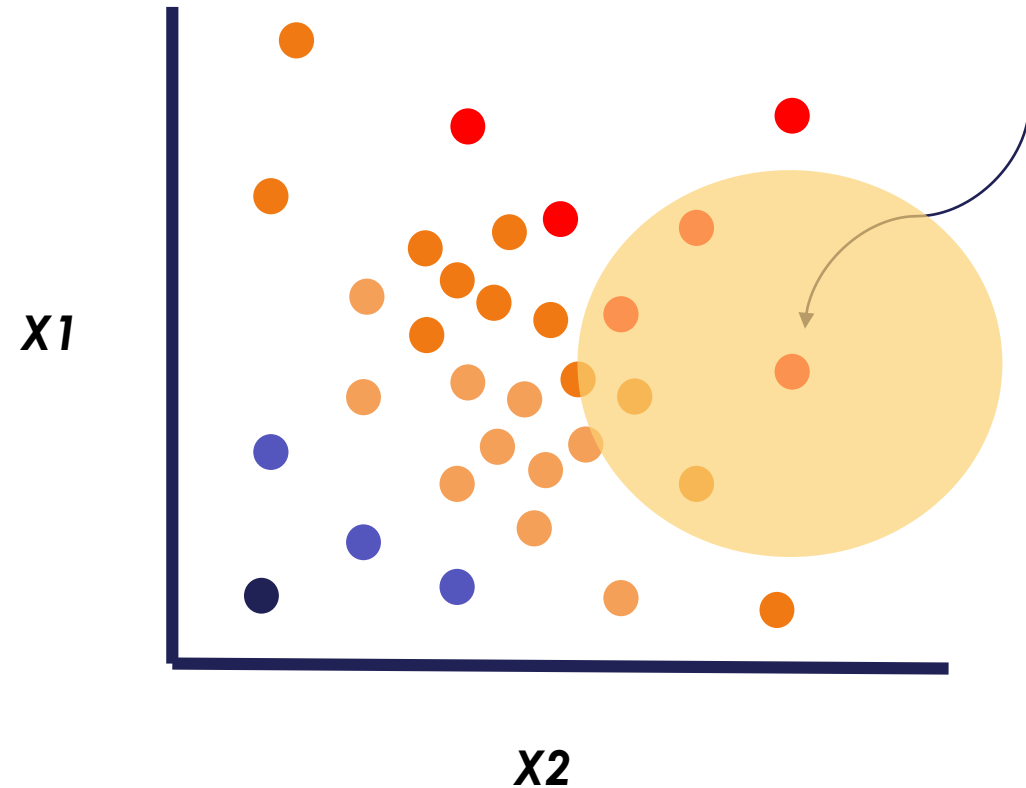
What is vicinity?

• Tabular data - problematic



What is vicinity?

• Tabular data - problematic

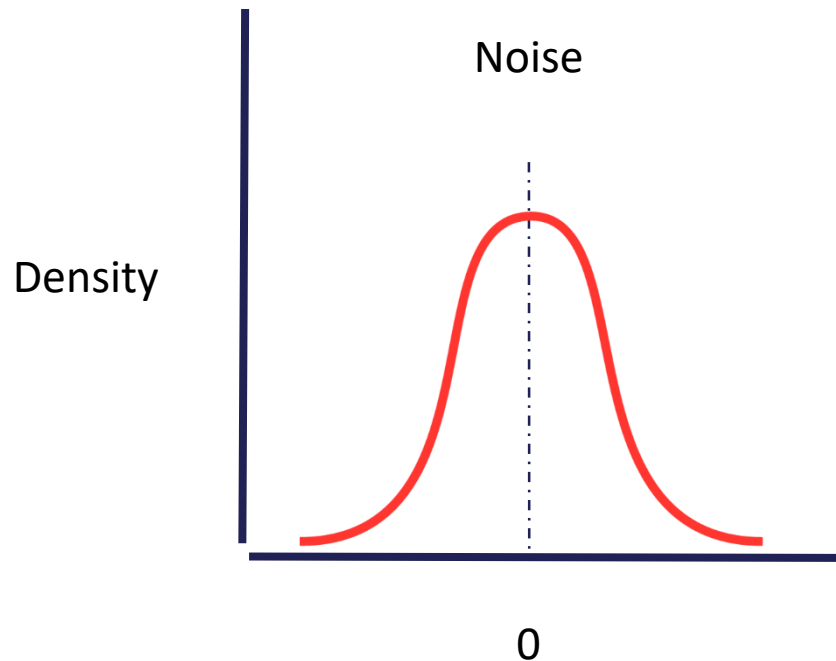


What is vicinity?

→ Kernel width (arbitrary)

• Tabular data – alternative sampling

Colour	Age	Income	Car make	Nr. Cards
Red	65	50000	Ford	5



Add noise
extracted from
 $N(0,1)$



Colour	Age	Income	Car make	Nr. Cards
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• Tabular data – alternative sampling

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Original implementation → sample at random from variable distribution.

Alternative (better) implementation → Add random noise to the observation we want to explain.

LIME with tabular data

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Trained
black box



Predictions
y1
y2
y3
y4
y5

LIME with tabular data

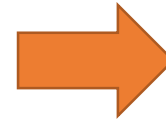
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Trained
black box

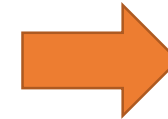


Predictions
y1
y2
y3
y4
y5

Colour	Age	Income	Car make	Nr. Cards	Predictions
0	65	51000	1	5	y1
1	63	59000	0	5.2	y2
0	60	55000	1	5.9	y3
0	58	45000	0	6	y4
0	55	47000	1	4	y5



Explainable
model



Interpretations

LIME with tabular data

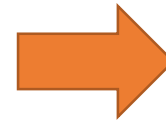
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Trained
black box

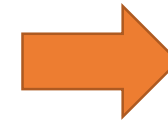


Predictions
y1
y2
y3
y4
y5

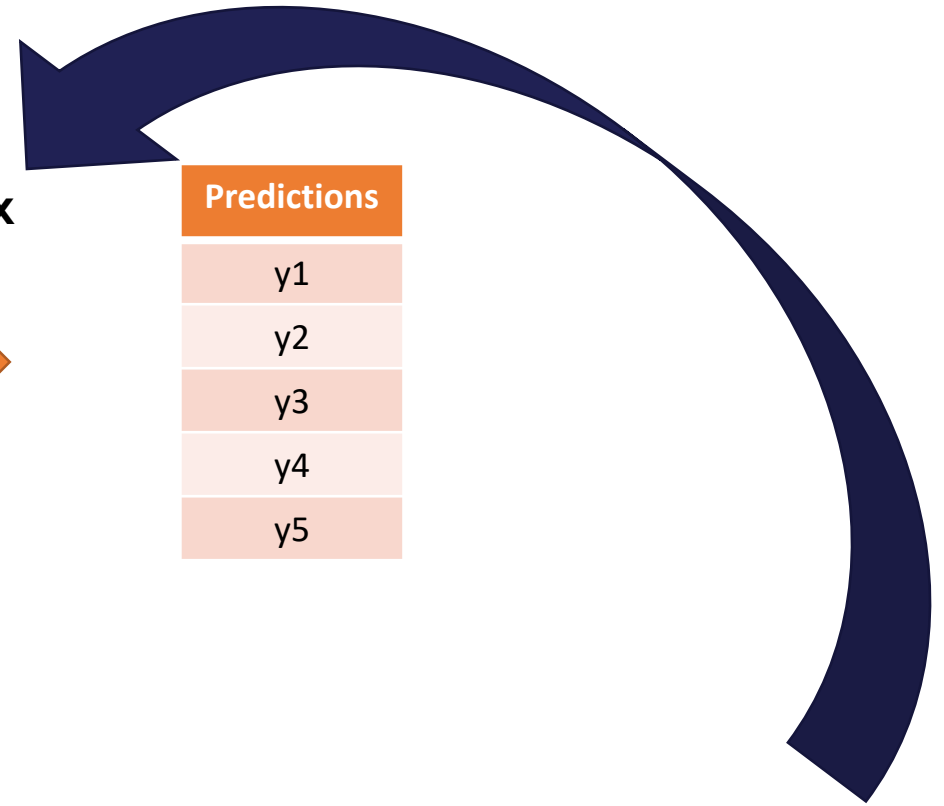
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0	58	45000	0	6	y4
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Explainable
model



Interpretations





Considerations

- The categorical variables are **one hot encoded** for the explainable model.
- Numerical variables can be sampled from a Gaussian or lhs distribution. Alternatively, add noise to the observation.
- Option to discretize continuous features and treat them as categorical.

THANK YOU

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