

### Exercise 1: Factorizing distributions

- (a) Can we factorize the joint distribution  $\mathbb{P}(x)$  as  $\mathbb{P}(x_S)\mathbb{P}(x_{-S})$ ? How can we factorize the joint distribution, such that the distribution is preserved? Formally prove your answer.
- (b) Let  $x_S \perp x_{-S}$ . Does the factorization preserve the joint now? Formally prove your answer.
- (c) Illustrate the two factorizations in a schematic drawing. *Hint:* You can draw a 2D scatterplot with two dependent variables. Given a fixed value for the conditioned variable, draw the range of values that conditional and marginal sampling consider.

### Exercise 2: Feature importance and extrapolation

Based on the results in Exercise 1, explain why and when the different feature importance methods extrapolate.

- (a) Over which distributions does PFI evaluate the model? Under which assumptions is the model evaluated outside the domain?
- (b) Over which distributions does CFI evaluate the model? What about SAGE? Do the methods extrapolate?
- (c) For both PFI and CFI evaluate whether/when the perturbed variables are dependent/independent of the target variable.
- (d) What does that mean for the interpretation of PFI and CFI?
- (e) Can a feature be relevant for CFI but not relevant for PFI?

### Exercise 3: In class discussion

Discuss with your neighbor. Which of the aforementioned methods is superior? PFI or the extrapolation-free alternatives?

- (a) Which method is most suitable for situations where we aim to understand the model's mechanism? If any?
- (b) Which method is most suitable for situations where we want to understand the data generating mechanism?
  - (i) In order to find features that are informative of the prediction target?
  - (ii) In order to select the smallest possible set of features, which would enable the same prediction performance?
  - (iii) In order to find variables that are causal for the prediction target?