

## Task 1 (Correlation vs MI)

a) Sample  $n = 1000$  observations from multivariate normal distribution:

$$\begin{pmatrix} X \\ Y \end{pmatrix} \sim \mathcal{N}\left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 & \rho \\ \rho & 1 \end{pmatrix}\right),$$

where  $\rho \in [0, 1]$ .

- Draw a heatmap of a distribution for  $\rho = 0, 0.5, 1$ .
  - Compute Pearson correlation of  $X$  and  $Y$ , and mutual information (after discretizing variables into 10 bins each) for various values of  $\rho$ .
  - For each  $\rho$  repeat the experiment from the previous bullet point  $N = 100$  times. Visualise the results.
  - Draw plots of  $\hat{I}(X, Y)$  as a function of  $-\log(1 - \rho^2)/2$  and  $-\log(1 - \hat{\rho}^2)/2$  (use  $\hat{I}(X, Y)$  and  $\hat{\rho}$  from the previous bullet point). What is the relationship between  $x$  and  $y$  in the plots?
- b) Sample  $n = 1000$  observations from a normal distribution  $X \sim \mathcal{N}(0, 1)$  and  $Y = X^2 + \varepsilon$ , where  $\varepsilon \sim \mathcal{N}(0, \sigma^2)$ .
- Draw a heatmap of a distribution for  $\sigma^2 = 0, 0.5, 2$ .
  - Compute Pearson correlation of  $X$  and  $Y$ , and mutual information (after discretizing variables into 10 bins each) for various values of  $\sigma^2$ .
  - For each  $\sigma^2$  repeat the experiment  $N = 100$  times. Visualise the results.

## Task 2 (Tests of independence)

**Asymptotic tests of independence** might be based on mutual information or Pearson's  $\chi^2$  statistic. We want to test the following hypothesis

$$H_0 : X \perp\!\!\!\perp Y.$$

In this approach we compute  $\widehat{MI} = \hat{I}(X, Y)$  or  $\chi^2$  for  $X$  and  $Y$  and then, using the fact that

$$\text{if } X \perp\!\!\!\perp Y, \text{ then } 2n\widehat{MI} \rightarrow \chi^2_{(|\mathcal{X}|-1)(|\mathcal{Y}|-1)}$$

and

$$\text{if } X \perp\!\!\!\perp Y, \text{ then } \chi^2 \rightarrow \chi^2_{(|\mathcal{X}|-1)(|\mathcal{Y}|-1)},$$

we compute p-values using values of the cumulative distribution function of  $\chi^2_{(|\mathcal{X}|-1)(|\mathcal{Y}|-1)}$ .

In **permutation tests** we compute  $B$  times the value of the test statistics for a sample, in which  $X$  is permuted ( $\widehat{MI}_b$  for  $b = 1, 2, \dots, B$ ) and then we compare  $\widehat{MI}$  with  $\widehat{MI}_b$  using the following formula for p-value:

$$\frac{1 + \sum_{b=1}^B \mathbb{I}(\widehat{MI} \leq \widehat{MI}_b^*)}{1 + B}.$$

- a) Write a function which runs asymptotic independence tests.  
 Input:  $X, Y$ , stats (one of 'mi' - mutual information, 'chi2' - Pearson's statistic)  
 Output: test statistic value, p-value
- b) Write a function performing independence test based on permutations.  
 Input:  $X, Y, B$  - number of permutations  
 Output: test statistic value, p-value  
 In a function, mutual information  $\widehat{MI}_b^*$  should be computed for resampled samples  $(X_b^*, Y)$  for  $B = 1, 2, \dots, B$ , where  $X_b^*$  is a random permutation of  $X$ .  
 Useful functions: `np.random.permutation`
- c) Draw a sample from a distribution, in which  $X$  and  $Y$  are dependent and a sample, in which  $X$  and  $Y$  are independent (you may use the example from task 1 a)). Next, test for independence using:
- asymptotic test based on  $2n\hat{I}(X, Y)$ ,
  - Pearson's chi-squared test,
  - permutation test.