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### Practical 3

#### Demand monitoring, part II

In this third practical, we expand the retail sales dataset of Practical 2 to include multiple products. Data for Product 1 (from Practical 2) and an additional 9 products are given in two forms: `sales_10.csv` and `sales_10_adjusted.csv`, with the same format. As the name indicates, sales volumes in the latter file have been adjusted for trend and seasonality (removing growth and cyclical patterns).

- (a) Explain the challenges involved when analysing the risk of stock-out across several products. What is the danger of considering each series independently?
- (b) Propose one graphical and one numerical method of detecting dependence of extreme values of the demand across several products. Apply your chosen methods to the **adjusted** sales data and identify groups of related products (if any exist).
- (c) In view of your answer to (b), would you apply a Gaussian copula model to these data?
- (d) (Hard) Fit a multivariate model to the adjusted sales data and estimate the 95% Demand at Risk for the sum,  $S = X^{(1)} + X^{(2)} + \dots + X^{(10)}$ . You can do so with the following steps:
  - i. Transform the sales for each product to a uniform scale, that is, compute  $\hat{F}_i(X^{(i)})$  for  $X^{(i)}$ ,  $i = 1, \dots, n$  being each of the product sales and  $\hat{F}_i$  being the estimated Extreme Value distributions.
  - ii. Fit a multivariate copula of your choice using the `copula::fitCopula` function.
  - iii. Simulate from your fitted copula and transform the values back to their original scales (i.e. undo the transformation in i.)
  - iv. Compute the simulated values of  $S$  and their 95% Value at Risk.