# Bonus Task Energy Market Optimisation

Group 13

 $10^{th}$  October 2021

## 1 Vector Auto-Regression

#### 1.1 Decision

Consider the for-ex trading, where we do not know the future prices at a given time t. We ideally want to know the best price at the next timet+1, often it is challenging to get the result before a given time t+1. Since our energy market price is updated every 30 minutes, by using Vector Auto-Regression, we are able to predict the price before t+1 hits. Thus, Vector Auto-Regression is computationally less expensive than ARMA. To improve this model, we could take into account external data such as Covid-19 (exogenous variable) or the interaction between Weather and Bushfire data (endogenous variable). As a result, our model might be less accurate due to the prediction errors. On the other hand, we could also try using other forecasting algorithm such as Long Short-Term Memory.

### 1.2 Concept

The equation for Vector Auto-Regression with 2 predictors and 195 lags are:

$$\begin{aligned} \mathbf{Y}_{1,t} &= \alpha_1 + \beta_{1,1} \times Y_{1,t-1} + \ldots + \beta_{1,195} \times Y_{2,t-1} + \epsilon_{1,t} \\ \mathbf{Y}_{2,t} &= \alpha_1 + \beta_{2,1} \times Y_{1,t-1} + \ldots + \beta_{2,195} \times Y_{2,t-1} + \epsilon_{2,t} \end{aligned}$$

where  $Y_{1,t}$  is the matrix of predicted price and  $Y_{2,t}$  is the matrix of predicted Demand. Here, we are taking the predicted price for the bonus task. For the prediction results, we use the training period to obtain  $\beta$  and test it on  $1^{th}$  July 2021 until  $15^{th}$  of August 2021. To calculate the revenue, we simulate the predicted price on the mandatory task algorithm. Finally, we obtained a revenue of \$4.184 millions. This is a pretty good estimate to the mandatory task's revenue, which is \$6 millions.

#### 1.3 Justification

Prior to fitting in the Vector Auto-regression model, we consider:

- Granger Causality Test: it is found that the optimal columns to use for Vector Auto-regression model in  $market\_data.xlxs$  are Victoria Trading Price and Demand.
- $\bullet$  Dickey-Fuller Test (Stationary Test): since all the p-values are significant, we reject H0 that the data is non-stationary. Thus, all the columns are stationary.
- Log-transformation: To reduce the variance of Victoria Trading Price and Demand, we consider log transformation with the formula: log(X) + min(X) + 1. We chose to add min(X) + 1 to deal with negative numbers. Since our data is already stationary, we did not use log-difference.
- Hannah-Quinn Information Criterion (Optimal Lag): we use 195 as the optimal lag.