## Task 2

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Over to you! Fill in the path to your working directory.

Example: filePath <- "C:/Users/Swaraj/Documents/Quantium\_Project/"

filePath <- "" # Leave empty if .Rmd and data are in the same directory, RStudio handles it.

If QVI data.csv is directly in your RMD's directory:

data <- fread("QVI\_data.csv")

Ensure DATE is a Date object for lubridate functions later

data[, DATE := as.Date(DATE)]

Set themes for plots (already in template)

theme\_set(theme\_bw()) theme\_update(plot.title = element\_text(hjust = 0.5))

Over to you! Add a new month ID column in the data with the format yyyymm.

Using lubridate to extract year and month, then combine

data[, YEARMONTH := year(DATE) \* 100 + month(DATE)]

Next, we define the measure calculations to use during the analysis.

Over to you! For each store and month calculate total sales, number of customers,

transactions per customer, chips per customer and the average price per unit.

Hint: you can use uniqueN() to count distinct values in a column

 $\label{eq:measureOverTime} $$\operatorname{data}[\ .(\ \operatorname{totSales} = \operatorname{sum}(TOT\_SALES), \ \operatorname{nCustomers} = \operatorname{uniqueN}(L\_CUSTOMER\_ID), \ \operatorname{nTxnPerCust} = \operatorname{uniqueN}(TRANSACTION\_ID) \ / \ \operatorname{uniqueN}(L\_CUSTOMER\_ID), \ \# \ \operatorname{Adjusted} \ \operatorname{to} \ \operatorname{be} \ \operatorname{per} \ \operatorname{unique} \ \operatorname{uniqueN}(TRANSACTION\_ID), \ \operatorname{avgPricePerUnit} = \ \operatorname{sum}(TOT\_SALES) \ / \ \operatorname{sum}(PROD\_QTY) \ / \ \operatorname{uniqueN}(TRANSACTION\_ID), \ \operatorname{avgPricePerUnit} = \ \operatorname{sum}(TOT\_SALES) \ / \ \operatorname{sum}(PROD\_QTY) \ ), \ \operatorname{by} = .(STORE\_NBR, \ YEARMONTH)] \ [\operatorname{order}(STORE\_NBR, \ YEARMONTH)] \ (STORE\_NBR, \ YEARMONTH) \ )$ 

Filter to the pre-trial period and stores with full observation periods

Pre-trial period is defined as before February 2019 (YEARMONTH < 201902)

Full observation period means 12 months of data (12 entries for each store in measureOverTime)

stores WithFullObs <- unique(measureOverTime[, .N, by = STORE\_NBR] [N == 12, STORE\_NBR]) pre-TrialMeasures <- measureOverTime [YEARMONTH < 201902 & STORE\_NBR %in% stores WithFullObs]

#### Define trial stores

```
trial\_stores\_list <- c(77, 86, 88)
```

Over to you! Create a function to calculate correlation for a measure, looping

through each control store.

Let's define inputTable as a metric table with potential comparison stores,

 $\operatorname{metricCol}$  as the store  $\operatorname{metric}$  used to calculate correlation on, and  $\operatorname{storeComparison}$ 

```
as the store number of the trial store. calculateCorrelation <- function(inputTable, metricCol, store-
Comparison) { calcCorrTable <- data.table(Store1 = numeric(), Store2 = numeric(), corr measure = nu-
storeNumbers <- unique(inputTable[, STORE_NBR]) # Filter out the trial store itself and other trial
stores from potential controls storeNumbers <- storeNumbers[!storeNumbers %in% c(storeComparison,
trial stores list[trial stores list!= storeComparison])]
for (i in storeNumbers) { # Extract data for the trial store and current potential control store trial data <-
inputTable[STORE NBR == storeComparison, get(metricCol)] control data <- inputTable[STORE NBR
== i, get(metricCol)]
# Ensure there's data to correlate
if (length(trial_data) > 1 && length(control_data) > 1) {
  calculatedMeasure <- data.table(</pre>
    "Store1" = storeComparison,
    "Store2" = i,
    "corr_measure" = cor(trial_data, control_data, use = "complete.obs") # Pearson correlation
  calcCorrTable <- rbind(calcCorrTable, calculatedMeasure)</pre>
}
} return(calcCorrTable) }
Create a function to calculate a standardised magnitude distance for a
measure, looping through each control store calculateMagnitudeDistance <- function(inputTable,
metricCol, storeComparison) { calcDistTable <- data.table(Store1 = numeric(), Store2 = numeric(), YEAR-
MONTH = numeric(), measure = numeric())
storeNumbers <- unique(inputTable[, STORE NBR]) # Filter out the trial store itself and other trial
stores from potential controls storeNumbers <- storeNumbers[!storeNumbers %in% c(storeComparison,
trial stores list[trial stores list!= storeComparison])]
for (i in storeNumbers) { # Calculate absolute difference calculatedMeasure <- data.table( "Store1" =
```

Create a function to calculate a standardised magnitude distance for a

tudeMeasure := 1 - (measure - minDist) / (maxDist - minDist)]

return(finalDistTable) }

storeComparison, "Store2" = i, "YEARMONTH" = inputTable[STORE\_NBR == storeComparison, YEARMONTH], "measure" = abs(inputTable[STORE\_NBR == storeComparison, get(metricCol)] - inputTable[STORE\_NBR == i, get(metricCol)]) calcDistTable <- rbind(calcDistTable, calculatedMeasure)

#### Standardise the magnitude distance so that the measure ranges from 0 to 1 minMaxDist <- calcDistTable[, .(minDist = min(measure), maxDist = max(measure)), by = c("Store1", "YEARMONTH")] distTable <- merge(calcDistTable, minMaxDist, by = c("Store1", "YEARMONTH")) distTable[, magni-

finalDistTable <- distTable[, .(mag\_measure = mean(magnitudeMeasure)), by = .(Store1, Store2)]

measure, looping through each control store calculateMagnitudeDistance <- function(inputTable, metricCol, storeComparison) { calcDistTable <- data.table(Store1 = numeric(), Store2 = numeric(), YEAR-MONTH = numeric(), measure = numeric())

storeNumbers <- unique(inputTable[, STORE\_NBR]) # Filter out the trial store itself and other trial stores from potential controls storeNumbers <- storeNumbers[!storeNumbers %in% c(storeComparison, trial\_stores\_list[trial\_stores\_list != storeComparison])]

for (i in storeNumbers) { # Calculate absolute difference calculatedMeasure <- data.table( "Store1" = storeComparison, "Store2" = i, "YEARMONTH" = inputTable[STORE\_NBR == storeComparison, YEARMONTH], "measure" = abs(inputTable[STORE\_NBR == storeComparison, get(metricCol)] - inputTable[STORE\_NBR == i, get(metricCol)]) ) calcDistTable <- rbind(calcDistTable, calculatedMeasure) }

#### Standardise the magnitude distance so that the measure ranges from 0 to 1 minMaxDist <- calcDistTable[, .(minDist = min(measure), maxDist = max(measure)), by = c("Store1", "YEARMONTH")]

 $distTable \leftarrow merge(calcDistTable, minMaxDist, by = c("Store1", "YEARMONTH")) distTable[, magnitudeMeasure := 1 - (measure - minDist) / (maxDist - minDist)]$ 

 $\label{lem:conditional} final Dist Table <- \ dist Table [, .(mag\_measure = mean(magnitude Measure)), \ by = .(Store1, \ Store2)] \\ return(final Dist Table) \ \}$ 

Over to you! Create a combined score composed of correlation and magnitude, by

first merging the correlations table with the magnitude table.

Hint: A simple average on the scores would be 0.5 \* corr\_measure + 0.5 \*

```
\label{eq:mag_measure} \begin{split} & \textbf{mag\_measure} & \quad \text{corr\_weight} <-0.5 \\ & \quad \text{score\_nSales} <- \text{merge}(\text{corr\_nSales}, \text{ magnitude\_nSales}, \text{ by} = \text{c}(\text{``Store1''}, \text{``Store2''}))[, \text{ scoreNSales} := \text{corr\_weight} * \text{corr\_measure} + (1 - \text{corr\_weight}) * \text{mag\_measure}] \\ & \quad \text{score\_nCustomers} <- \text{merge}(\text{corr\_nCustomers}, \text{ magnitude\_nCustomers}, \text{ by} = \text{c}(\text{``Store1''}, \text{``Store2''}))[, \text{scoreNCust} := \text{corr\_weight} * \text{corr\_measure} + (1 - \text{corr\_weight}) * \text{mag\_measure}] \end{split}
```

Over to you! Combine scores across the drivers by first merging our sales

```
scores and customer scores into a single table score_Control <- merge(score_nSales, score_nCustomers, by = c("Store1", "Store2")) score_Control[, finalControlScore := scoreNSales * 0.5 + scoreNCust * 0.5]
```

Select control stores based on the highest matching store (closest to 1 but

not the store itself, i.e. the second ranked highest store)

Over to you! Select the most appropriate control store for trial store 77 by

finding the store with the highest final score.

## Exclude the trial store itself from being selected as a control

 $control\_store\_77 <- score\_Control[Store1 == trial\_store \& Store2 != trial\_store] [order(-finalControlScore)] [1, Store2] \\ control\_store\_77$ 

## Convert YEARMONTH to Date for plotting

measure OverTime[, TransactionMonth := as.Date(paste(YEARMONTH %/% 100, YEARMONTH %% 100, 1, sep = "-"), "%Y-%m-%d")]

pastSales <- measureOverTime[, Store\_type := ifelse(STORE\_NBR == trial\_store, "Trial", ifelse(STORE\_NBR == control\_store\_77, "Control", "Other stores"))][, .(totSales = mean(totSales)), by = .(YEARMONTH, Store\_type, TransactionMonth) # Mean if multiple entries for same month/type][YEARMONTH < 201903] # Filter for pre-trial period for visual check

$$\begin{split} & \operatorname{ggplot}(\operatorname{pastSales},\ \operatorname{aes}(\operatorname{TransactionMonth},\ \operatorname{totSales},\ \operatorname{color} = \operatorname{Store\_type})) + \operatorname{geom\_line}(\operatorname{size} = 1) + \\ & \operatorname{geom\_point}(\operatorname{size} = 2) + \operatorname{labs}(x = \operatorname{``Month}\ \operatorname{of}\ \operatorname{operation''},\ y = \operatorname{``Total}\ \operatorname{sales''},\ \operatorname{title} = \operatorname{paste0}(\operatorname{``Total}\ \operatorname{sales}\ \operatorname{by}\ \operatorname{month}\ \operatorname{for}\ \operatorname{Trial}\ \operatorname{Store''},\ \operatorname{trial\_store},\ \operatorname{``vs}\ \operatorname{Control}\ \operatorname{Store''},\ \operatorname{control\_store\_77})) + \operatorname{scale\_color\_manual}(\operatorname{values}\ = \operatorname{c}(\operatorname{``Control''} = \operatorname{``blue''},\ \operatorname{``Trial''} = \operatorname{``red''},\ \operatorname{``Other}\ \operatorname{stores''} = \operatorname{``grey''})) + \operatorname{theme\_minimal}() \end{split}$$

### Over to you! Conduct visual checks on customer count trends by comparing the ### trial store to the control store and other stores.

Hint: Look at the previous plot. pastCustomers <- measureOverTime[, Store\_type := ifelse(STORE\_NBR == trial\_store, "Trial", ifelse(STORE\_NBR == control\_store\_77, "Control", "Other stores"))][, .(nCustomers = mean(nCustomers)), by = .(YEARMONTH, Store\_type, TransactionMonth)][YEARMONTH < 201903] # Filter for pre-trial period

#### Scale pre-trial control sales to match pre-trial trial store sales

scaling FactorForControlSales <- preTrialMeasures [STORE\_NBR == trial\_store & YEARMONTH < 201902, sum (totSales)] / preTrialMeasures [STORE\_NBR == control\_store\_77 & YEARMONTH < 201902, sum (totSales)]

Apply the scaling factor

# Make a copy to avoid modifying original measureOverTime directly for scaling

measureOverTimeSales scaled <- copy(measureOverTime)

 $measureOverTimeSales\_scaled[STORE\_NBR == control\_store\_77, controlSales := totSales * scalingFactorForControlSales]$ 

## Prepare data for percentage difference calculation for sales

 $\label{trial_sales_data} $$\operatorname{trial\_sales\_data} < -\operatorname{measureOverTimeSales\_scaled[STORE\_NBR} == \operatorname{trial\_store}, .(YEARMONTH, \operatorname{totSales\_trial} = \operatorname{totSales})] \\ \operatorname{control\_sales\_data} < -\operatorname{measureOverTimeSales\_scaled[STORE\_NBR} == \operatorname{control\_store\_77}, .(YEARMONTH, \operatorname{controlSales})]$ 

Over to you! Calculate the percentage difference between scaled control sales

and trial sales percentageDiffSales <- merge(trial\_sales\_data, control\_sales\_data, by = "YEAR-MONTH")[, percentageDiff := abs(totSales\_trial - controlSales) / controlSales]

## Print percentage difference for sales (optional, for checking)

## print(percentageDiffSales)

As our null hypothesis is that the trial period is the same as the pre-trial

period, let's take the standard deviation based on the scaled percentage difference

in the pre-trial period stdDevSales <- sd(percentageDiffSales[YEARMONTH < 201902, percentageDiff])

Note that there are 8 months in the pre-trial period

hence 8-1=7 degrees of freedom degreesOfFreedom <- 7

We will test with a null hypothesis of there being no difference between trial and control stores.

Over to you! Calculate the t-values for the trial months. After that, find the

95th percentile of the t distribution with the appropriate degrees of freedom

to check whether the hypothesis is statistically significant.

Hint: The test statistic here is (xu)/standard deviation

# Mean percentage difference in pre-trial period (mu)

mu\_sales <- mean(percentageDiffSales[YEARMONTH < 201902, percentageDiff])

#### Calculate t-value for each trial month

percentageDiffSales[, tValue := (percentageDiff - mu\_sales) / stdDevSales]

## Calculate 95th percentile of the t-distribution for comparison

```
t_{critical\_sales} < -qt(0.95, df = degreesOfFreedom)
```

 $message(paste("T-critical~(95th~percentile)~for~sales:", round(t\_critical\_sales, 3)))~print(percentageDiffSales[YEARMONTH >= 201902])~\#~Show~t-values~for~trial~months$ 

Trial and control store total sales

Over to you! Create new variables Store\_type, totSales and TransactionMonth in

the data table.

## Combine data for plotting sales trends with CI

```
plot_data_sales <- copy(measureOverTime) # Start with a copy of measureOverTime plot_data_sales[STORE_NBR == trial_store, Store_type := "Trial"] plot_data_sales[STORE_NBR == control_store_77, Store_type := "Control"] plot_data_sales[!(STORE_NBR %in% c(trial_store, control_store_77)), Store_type := "Other stores"]
```

pastSales <- plot\_data\_sales[, .(totSales = totSales, TransactionMonth = TransactionMonth, Store\_type = Store\_type) # Select relevant columns][Store\_type %in% c("Trial", "Control", "Other stores")] # Ensure all are included for context plot

Control store 5th percentile pastSales\_Controls5 <- copy(pastSales[Store\_type == "Control"]) past-Sales\_Controls5[, totSales := totSales \* (1 - stdDevSales \* 2)] # Apply 2 std dev for 5% approx. past-Sales\_Controls5[, Store\_type := "Control 5th % confidence interval"]

trialAssessmentSales <- rbind(pastSales, pastSales\_Controls95, pastSales\_Controls5)

Plotting these in one nice graph ggplot(trialAssessmentSales, aes(TransactionMonth, totSales, color = Store\_type)) + # Highlight trial period geom\_rect(data = trialAssessmentSales[YEARMONTH >= 201902 & YEARMONTH <= 201904 & Store\_type == "Trial", .(xmin = min(TransactionMonth), xmax = max(TransactionMonth), ymin = -Inf, ymax = Inf)], aes(xmin = xmin, xmax = xmax, ymin = ymin, ymax = ymax), fill = "lightblue", alpha = 0.3, color = NA, show.legend = FALSE) + geom\_line(aes(linetype = Store\_type), size = 1) + geom\_point(size = 2) + labs(x = "Month of operation", y = "Total sales (\$)", title = paste0("Total Sales: Trial Store", trial\_store, " vs Control Store ", control\_store\_77," with 90% CI")) + scale\_color\_manual(values = c("Trial" = "red", "Control" = "blue", "Control 95th % confidence interval" = "darkgreen", "Other stores" = "grey")) + scale\_linetype\_manual(values = c("Trial" = "solid", "Control" = "solid", "Control 95th % confidence

interval" = "dashed", "Control 5th % confidence interval" = "dashed", "Other stores" = "dotted")) + theme minimal()

#### This would be a repeat of the steps before for total sales

Scale pre-trial control customers to match pre-trial trial store customers

Over to you! Compute a scaling factor to align control store customer counts

to our trial store. scalingFactorForControlCust <- preTrialMeasures[STORE\_NBR == trial\_store & YEARMONTH < 201902, sum(nCustomers)] / preTrialMeasures[STORE\_NBR == control\_store\_77 & YEARMONTH < 201902, sum(nCustomers)]

Then, apply the scaling factor to control store customer counts.

# Make a copy to avoid modifying original measureOverTime directly for scaling

measureOverTimeCusts scaled <- copy(measureOverTime)

 $measureOverTimeCusts\_scaled[STORE\_NBR == control\_store\_77, \ controlCustomers := nCustomers * scalingFactorForControlCust]$ 

## Prepare data for percentage difference calculation for customers

 $\label{lem:cust_data} $$\operatorname{cust_data} < \operatorname{measureOverTimeCusts_scaled}[STORE\_NBR == \operatorname{trial\_store}, .(YEARMONTH, nCustomers\_trial = nCustomers)] \\ \operatorname{control\_cust\_data} < \operatorname{measureOverTimeCusts\_scaled}[STORE\_NBR == \operatorname{control\_store\_77}, .(YEARMONTH, \operatorname{controlCustomers})]$ 

Finally, calculate the percentage difference between scaled control store

**customers and trial customers.** percentageDiffCust <- merge(trial\_cust\_data, control\_cust\_data, by = "YEARMONTH")[, percentageDiff := abs(nCustomers\_trial - controlCustomers) / controlCustomers]

# Print percentage difference for customers (optional, for checking)

# print(percentageDiffCust)

As our null hypothesis is that the trial period is the same as the pre-trial

period, let's take the standard deviation based on the scaled percentage difference

in the pre-trial period stdDevCust <-sd(percentageDiffCust[YEARMONTH < 201902, percentageDiff]) degreesOfFreedom <-7 # Assuming 8 months in pre-trial, 8-1 = 7 degrees of freedom

## Combine data for plotting customer trends with CI

plot\_data\_cust <- copy(measureOverTime) # Start with a copy of measureOverTime plot\_data\_cust[STORE\_NBR == trial\_store, Store\_type := "Trial"] plot\_data\_cust[STORE\_NBR == control\_store\_77, Store\_type := "Control"] plot\_data\_cust[!(STORE\_NBR %in% c(trial\_store, control\_store\_77)), Store\_type := "Other stores"]

 $pastCustomers <- plot\_data\_cust[, .(nCusts = nCustomers, TransactionMonth = TransactionMonth, Store\_type = Store\_type)][Store\_type \%in\% c("Trial", "Control", "Other stores")]$ 

Control store 95th percentile pastCustomers\_Controls95 <- copy(pastCustomers[Store\_type == "Control"]) pastCustomers\_Controls95[, nCusts := nCusts \* (1 + stdDevCust \* 2)] pastCustomers\_Controls95[, Store\_type := "Control 95th % confidence interval"]

 $\begin{tabular}{ll} {\bf Control store 5th \ percentile} & pastCustomers\_Controls5 < -copy(pastCustomers[Store\_type == "Control"]) & pastCustomers\_Controls5[, nCusts := nCusts * (1 - stdDevCust * 2)] & pastCustomers\_Controls5[, Store\_type := "Control 5th % confidence interval"] \\ \end{tabular}$ 

trialAssessmentCust <- rbind(pastCustomers, pastCustomers Controls95, pastCustomers Controls5)

Plotting these in one nice graph ggplot(trialAssessmentCust, aes(TransactionMonth, nCusts, color = Store\_type)) + # Highlight trial period geom\_rect(data = trialAssessmentCust[YEARMONTH >= 201902 & YEARMONTH <= 201904 & Store\_type == "Trial", .(xmin = min(TransactionMonth), xmax = max(TransactionMonth), ymin = -Inf, ymax = Inf)], aes(xmin = xmin, xmax = xmax, ymin = ymin, ymax = ymax), fill = "lightblue", alpha = 0.3, color = NA, show.legend = FALSE) + geom\_line(aes(linetype = Store\_type), size = 1) + geom\_point(size = 2) + labs(x = "Month of operation", y = "Total number of customers", title = paste0("Total Customers: Trial Store", trial\_store, "vs Control Store", control\_store\_77," with 90% CI")) + scale\_color\_manual(values = c("Trial" = "red", "Control" = "blue", "Control 95th % confidence interval" = "darkgreen", "Other stores" = "grey")) + scale\_linetype\_manual(values = c("Trial" = "solid", "Control" = "solid", "Control 95th % confidence interval" = "dashed", "Control 5th % confidence interval" = "dashed", "Control 5th % confidence interval" = "dashed", "Other stores" = "dotted")) + theme\_minimal()