



ASIGNMENT 1

ISSS602 – Data Analytics Lab

2020 TOKYO OLYMPICS
Insights from Swimming Reaction Time

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1. OVERVIEW

With technological progress, data has become easier to collect, shared and analyzed. As a result, data analytics has been used in many sports to carry out in-depth studies on athletes' performance metrics, thereby providing their team with insights to improve future results (Morgulev, 2018).

2. OBJECTIVES

The main objective of this study is to understand how swimmers' reaction time vary with different factors, and whether it provides any advantage for better final performance.

The reaction time (R.T) is defined as the time taken for the feet to leave the starting position after the start signal.

JMP Pro 16.0 is used throughout the processes of data import, preparation, and analysis, using various wrangling, interactive exploration of trends and relationships, as well as hypothesis testing to validate these observations.

3. DATA

3.1 Data Used

Omega, the official time partner of the Olympics, provided the original data source in PDF form, separated by event (stroke styles), round, and gender. Data was then extracted into a single .csv file named “swimming”, which will be imported into JMP Pro and serve as the main source file for the subsequent data processing and analysis.

3.2 Data Quality Issues and Wrangling Solutions

Issue 1:

From the R.T column, using column view (*Figure 1*), one outlier identified for row 1347 with reaction time of 33.19s – not realistic being generally sub-second. Tracing back to the data table and the original pdf file, the data was not imported correctly for this row, as the R.T of 0.66s was read as “Team”, and the split 50 timing was read as R.T.

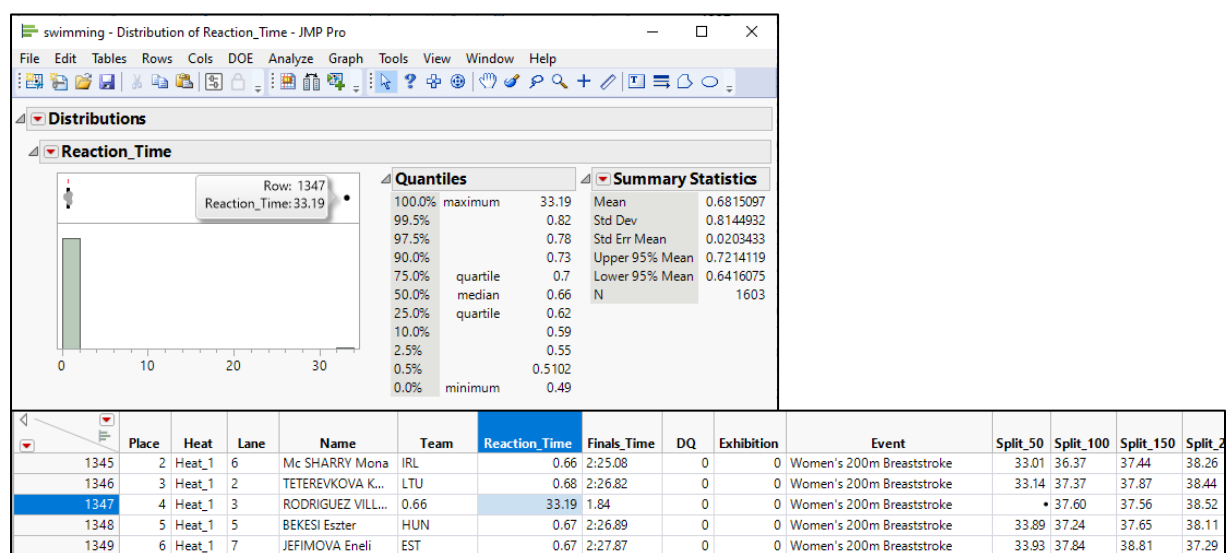


Figure 1: Abnormal R.T

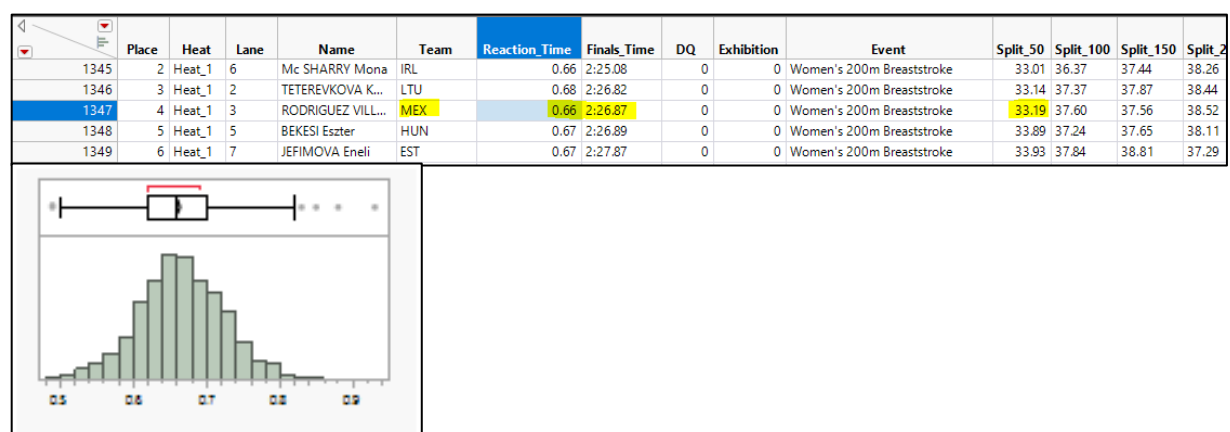


Figure 2: Corrected R.T based on Source PDF and Resulted Overall Distribution

Resolution: As only 1 row is erroneous, editing on the worksheet is acceptable and achieve a reasonable-looking distribution (*Figure 2*)

Issue 2: There are 193 missing values for Reaction time (10.7%) (Figure 3).

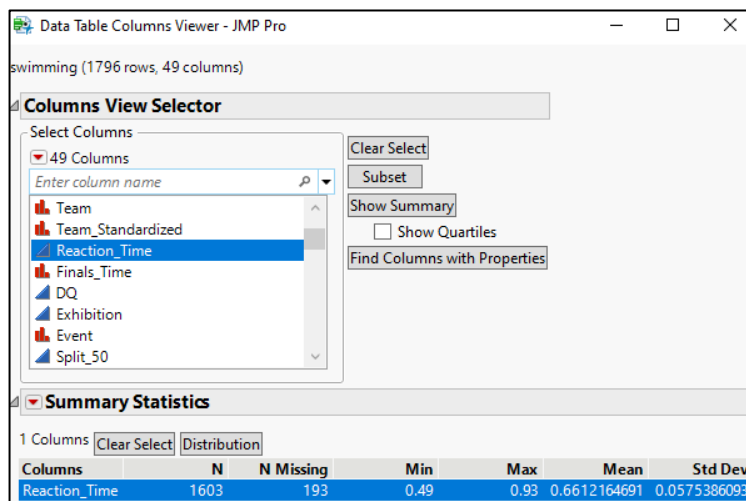


Figure 3: Missing R.T Values

Resolution: By selecting a sample R.T cell with missing value, using Rows > Row Selection > Select Matching cells..., all missing entries can be identified on Row pane. Using Data View on the selected rows, it can be noted that all 7 relay events face this issue (167 out of 193 rows). All these 193 rows will be excluded and hidden from the subsequent analysis.

This is acceptable because this study is focusing on the R.T from the starting signal (the horn sound). In relays, only the first swimmers start with that signal, everyone afterwards starts when the preceding swimmer ends his/her turn. Hence, most of R.T in relays are not relevant to our study. The other 26 data points count for a very small portion, the remaining 1603 data points constitute significant size for statistical conclusions.

Issue 3: Some split timing columns are in categorical data type.

Resolution: formatting them into continuous data type via Standardized Column Attributes (Figure 4).

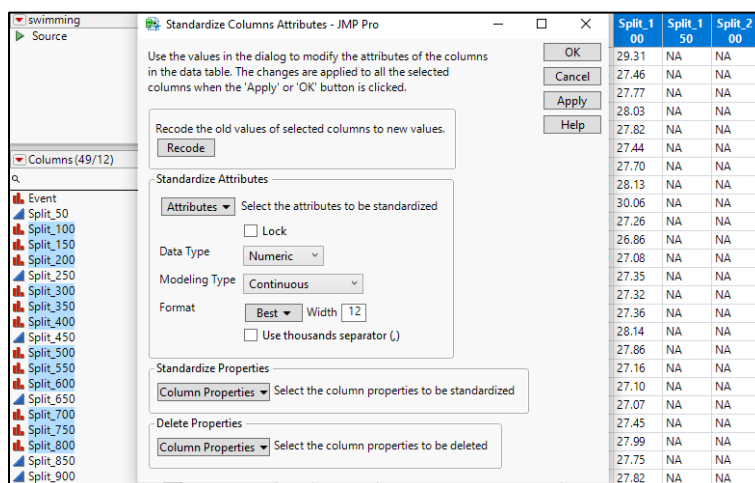


Figure 4: Correction on Split Timing Data Type

Issue 4: Team names are not consistent. E.g., New Zealand, having both “NZL” and “NZL – New Zealand”.

Resolution: Recoding this column to “Team_Standardized” column to correct all repeated and inconsistent values (Figure 5).

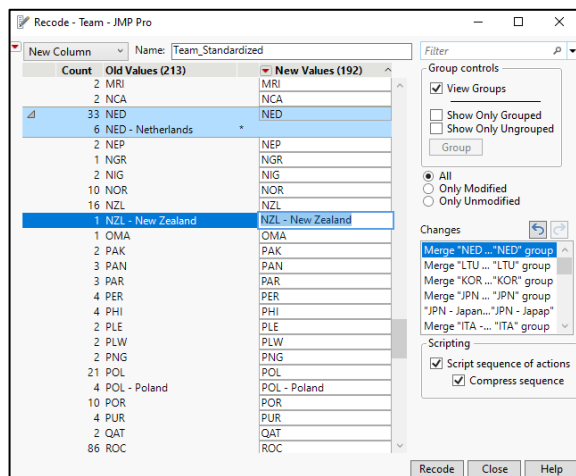


Figure 5: Correcting Inconsistent Team Data via Recoding

Issue 5: there is no gender column for separating the population in later analysis.

Resolution: a new “Gender” column is created by recoding the “Event” column, grouping all Men's events into “Male”, all Women's into “Female” (Figure 6).

	ed	Reaction_Time	Finals_Time	DQ	Exhibition	Event	Gender
946		0.76	7:49.14	0	0	Men's 800m Freestyle	Male
947		0.74	7:53.00	0	0	Men's 800m Freestyle	Male
948		0.66	7:54.98	0	0	Men's 800m Freestyle	Male
949		0.75	7:45.73	0	0	Men's 800m Freestyle	Male
950		0.68	7:46.09	0	0	Men's 800m Freestyle	Male
951		0.69	7:46.94	0	0	Men's 800m Freestyle	Male
952		0.79	7:47.73	0	0	Men's 800m Freestyle	Male
953		0.76	7:49.47	0	0	Men's 800m Freestyle	Male
954		0.71	7:49.97	0	0	Men's 800m Freestyle	Male
955		0.72	7:59.04	0	0	Men's 800m Freestyle	Male
956		0.73	8:00.16	0	0	Men's 800m Freestyle	Male
957		0.56	57.47	0	0	Women's 100m Backstroke	Female
958		0.53	57.72	0	0	Women's 100m Backstroke	Female
959		0.59	58.05	0	0	Women's 100m Backstroke	Female
960		0.61	58.43	0	0	Women's 100m Backstroke	Female

Figure 6: New Gender Column from Event Column

Issue 6: As Heats are preliminary rounds to select the top athletes to semis and finals, they contain swimmers from countries traditionally less developed in the sport and hence the R.T may differ. All the round information is not separated, and combined under the column “Heat”

Resolution: a new column Round Type is recoded from the original column Heat (Figure 7).

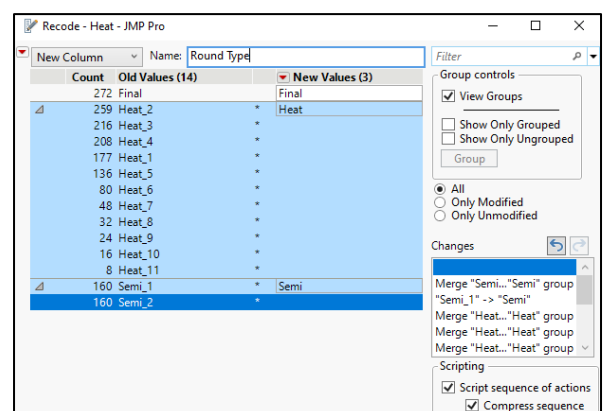


Figure 7: New Round Type Column from Heat Column via Recoding

Issue 7: There is no separate column for swimming styles of event (e.g., backstroke, freestyle) as a potential factor affecting R.T. Also, backstroke style has a different start position (inside the pool) from the rest of the styles (on the blocks). Hence it is worthwhile to analyze separately by separating the styles into two groups.

Resolution: Using recoding, two new columns “Stroke” and “Start_Position” are created (Figure 8).

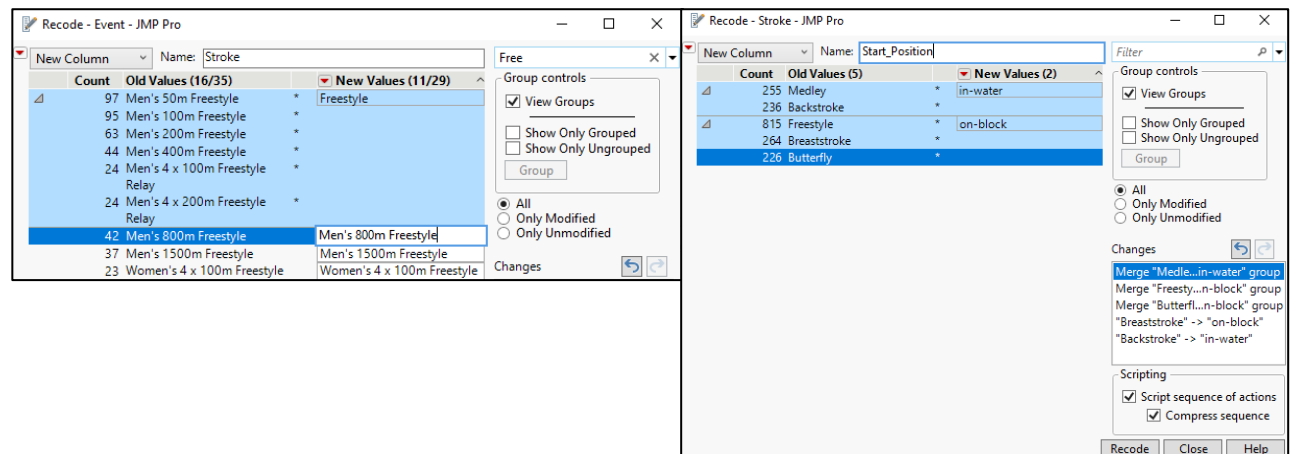


Figure 8: New Column Stroke and Start_Position Created via Recoding

Issue 8: Besides swimming style, swing distance may affect how the athletes train - sprinter vs distance swimmers - and hence could affect their R.T.

Resolution: A new column “Distance” is created from Event column using the race length information (e.g., 50m, 100m...). Afterwards, another column “Sprint_Distance” which groups all 50m events into “Sprint” and the rest into “Distance” is created.

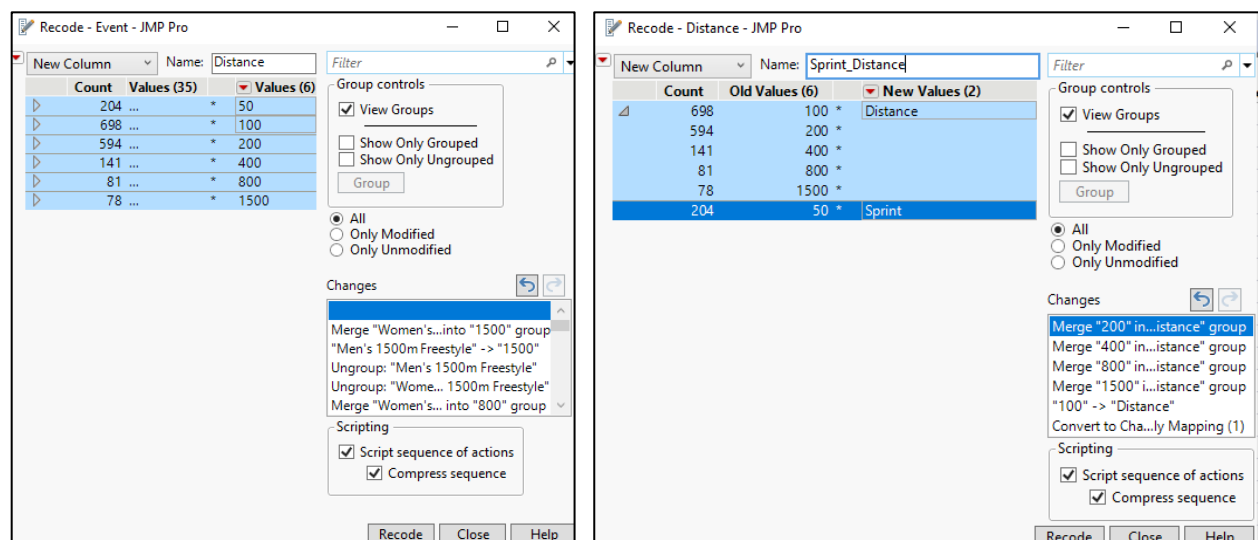


Figure 9: New Column Distance and Sprint_Distance Created via Recoding

The final JMP table is named “Swim_Final”.

4. DATA ANALYSIS

4.1 Reaction Time with Gender and Start Position

4.1.1 Subdividing the Populations and Normality Tests

Besides gender being the potential determining factor for differences in R.T (*Figure 10a*) with male swimmers having lower mean R.T than females, the starting positions seem to be another factor. Backstroke starting position is different from inside the pool with their feet against the wall, instead of from the blocks above the pool in the rest of the strokes. Hence backstroke R.T form a separate population from the rest.

Separating all the R.T by the above 2 factors, we can make the following observations based on *Figure 10b*.

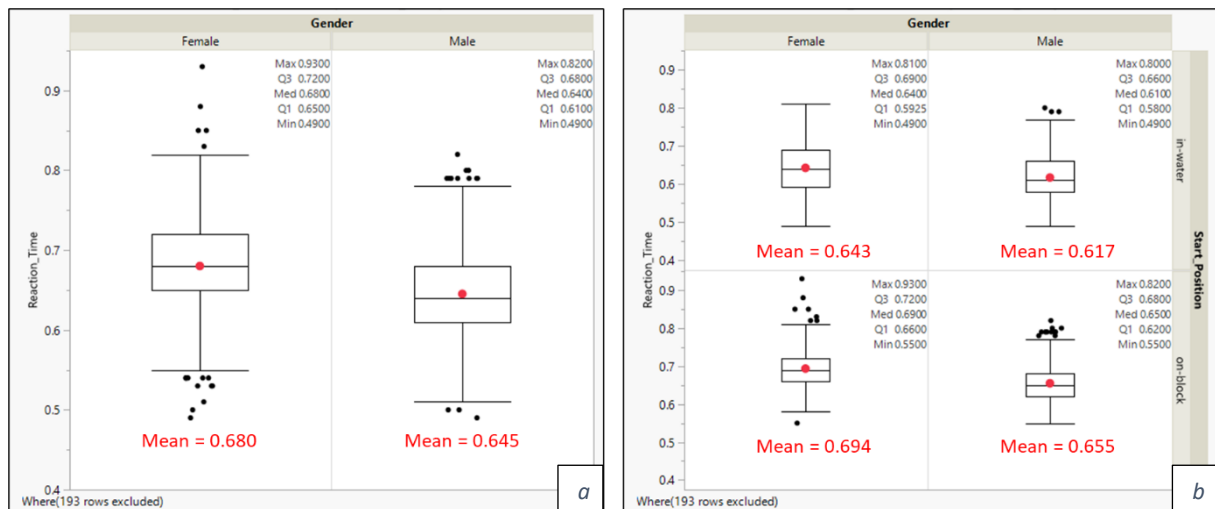


Figure 10: Male vs Female R.T, Dissected by Start_Position

Observation 1: Female in-water starting position has lower mean RT than on-block starting position

Observation 2: Male in-water starting position has lower mean RT than on-block starting position

Observation 3: Male have lower mean R.T than Female for in-water starting position

Observation 4: Male have lower mean R.T than Female for on-block starting position

Group	Gender	Start Position
1	Female	In-water
2	Male	In-water
3	Female	On-block
4	Male	On-block

Table 1: Groupings for R.T vs Gender/Start Position Study

Before selecting appropriate confirmatory analysis tests, it is necessary to perform the normality check on the above groups, with 99% confidence level selected. By the distribution and goodness-of-fit test, *Figure 11* is obtained and summarized in *Table 2* to test the following null and alternative hypotheses.

Null hypothesis H_0 : The group R.T is normally distributed.

Alternative hypothesis H_a : The group R.T is not normally distributed.

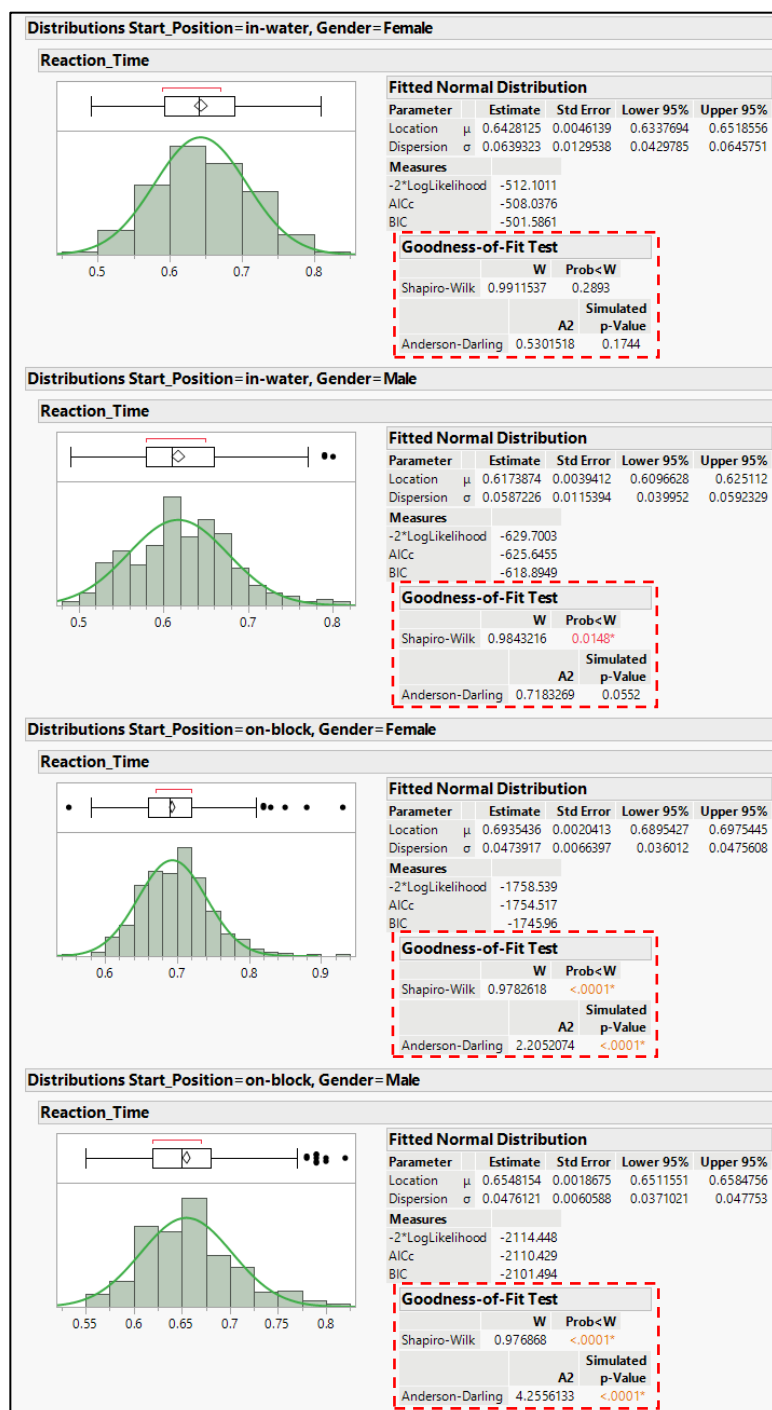


Figure 11: Distribution of Group 1-4 RT

Group	Gender	Start Position	Prob<W (Shapiro-Wilk)	p-value (Anderson-Darling)	Conclusion on H ₀
1	Female	In-water	0.2893	0.1744	Not rejected
2	Male	In-water	0.0148	0.0552	Not rejected
3	Female	On-block	<0.0001	<0.0001	Rejected
4	Male	On-block	<0.0001	<0.0001	Rejected

Table 2: Normality Test Results for Group 1-4

From Table 2, based on both Shapiro-Wilk and Anderson-Darling goodness-of-fit tests, we can infer that for in-water starting position, the R.T follow the normal distribution, while for other strokes the R.T distribution is not normal. Hence, the type of hypothesis tests with the null hypotheses are:

Observation No.	Comparison Pairs between Group:			Hypothesis No.	Type of Hypothesis Test	Null Hypothesis H ₀
1	1	vs	3	A	Nonparametric	The mean R.T of female is the same for in-water and on-block starting positions.
2	2	vs	4	B	Nonparametric	The mean R.T of male is the same for in-water and on-block starting positions.
3	1	vs	2	C	Parametric	The mean R.T of female and male is the same for in-water starting positions.
4	3	vs	4	D	Nonparametric	The mean R.T of female and male is the same for on-block starting positions.

Table 3: Hypothesis Tests Applied to Group 1-4 Comparison

4.1.2 Starting Position and Reaction Time

Firstly, the difference in mean R.T between the two starting positions is studied for each gender.

From *Figure 12*, for the nonparametric tests on hypothesis A and B, with the confidence level of 99%, the p-values are lower than the alpha value of 0.01. Therefore, we reject the null hypotheses A and B, and infer that for both male and female swimmers, the mean R.T in for in-water starting position is lower than on-block starting position.

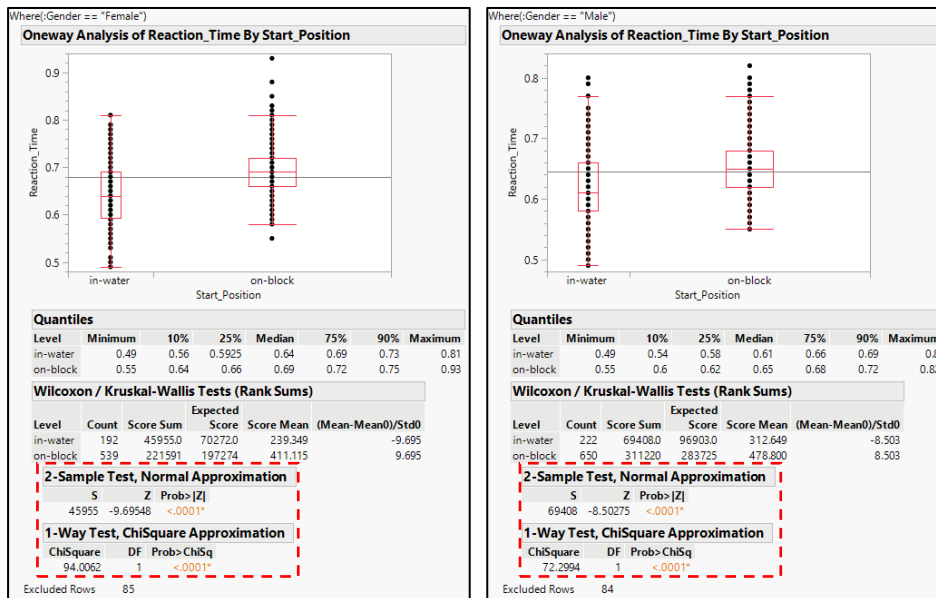


Figure 12: Hypothesis Test A and B Results for Mean R.T vs Start_Position

This is a reasonable inference as swimmers may need to exert more force when starting on the blocks to push propel their bodyweight both vertically and horizontally into air before entering the pool. In the case of backstroke, the athletes exert force mostly in the horizontal direction. Although the drag force under water is higher than air, for the short distance of R.T, this appears to be a less significant factor.

4.1.3 Gender and Reaction Time

For hypothesis C, we are checking if the mean R.T of male and female athletes are the same for in-water starting position. By conducting both the parametric Pooled t Test and the Welch's Test, the result in *Figure 13* is obtained.

Under the Pooled t Test (equal variances assumption), from the two-tail test result (Prob > |t|: <0.0001), the observed difference in mean R.T is statistically significant at 99% confidence level. Hence, we reject the null hypothesis that the 2 means are the same. From the left-tail test result (Prob < t: <0.0001), the observation of male R.T being "less than" female R.T is statistically significant, and we can infer that male athletes' mean R.T is lower than that of female athletes.

Under the Equal Variances test, as the p-values across the tests are all higher than the alpha value of 0.01, we cannot reject the null hypothesis that the variances of these 2 populations are equal. As these variances are inferred to be the same, we can maintain our conclusion from the previous Pooled t Test.

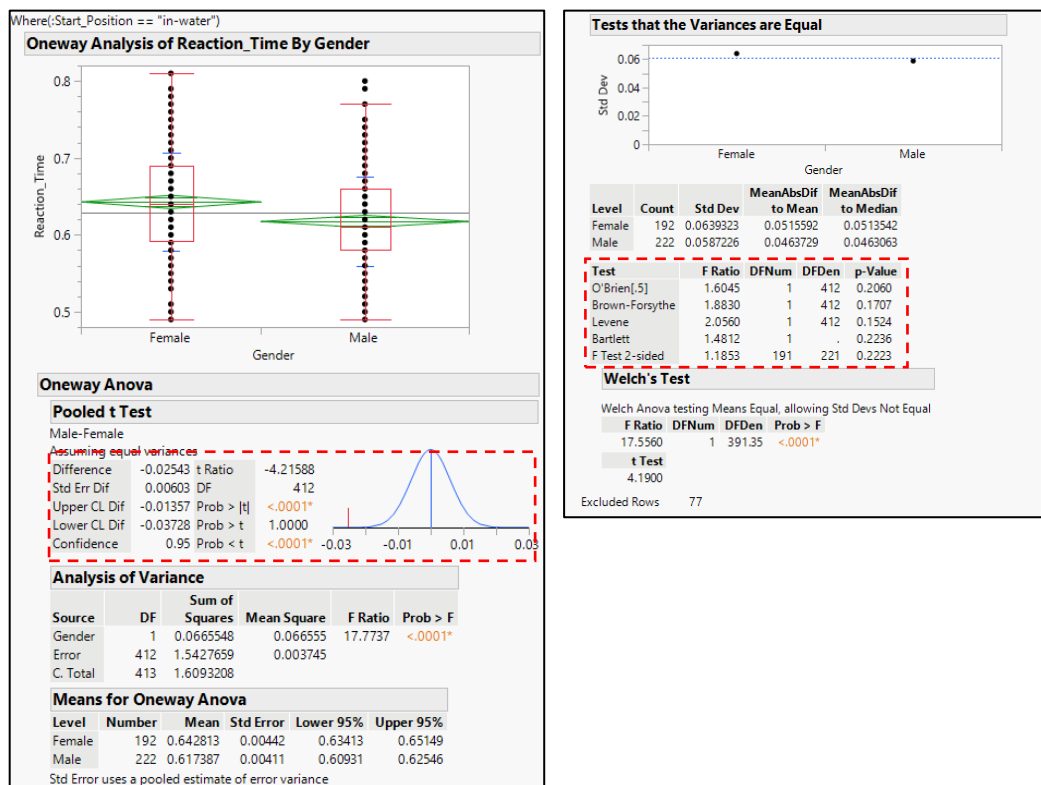


Figure 13: Hypothesis Test C Results for Mean R.T vs Gender (in-water)

For hypothesis D, we are repeating the hypothesis C for the on-block starting position. Applying the nonparametric test for these 2 populations of female and male mean R.T, we obtain the result in Figure 14. From both tests, the p-value is lower than the specified alpha level of 0.01. Therefore the null hypothesis that these two mean R.T values are the same is rejected, and it can be inferred that the observed lower mean R.T from male swimmers is statistically significant. This is not a surprising result as male athletes with generally stronger physique, hence higher ability to generate force and to have better R.T performance.

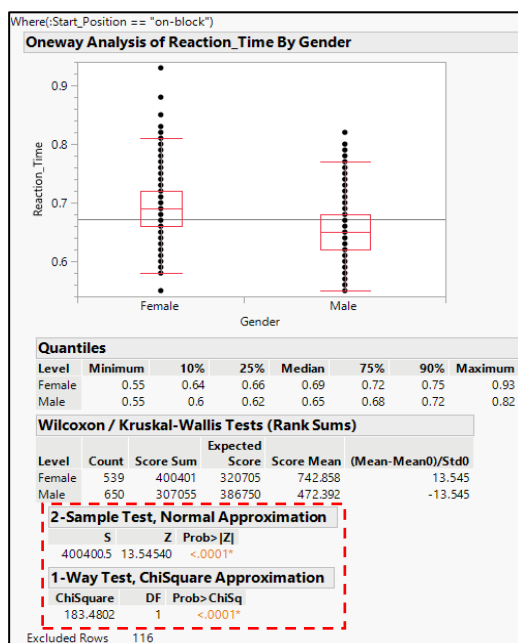


Figure 14: Hypothesis Test D Results for Mean R.T vs Gender (on-block)

4.2 Swimming Events and Reaction Time

4.2.1 Reaction Time Between Sprinter and Long-Distance Swimmers

In this section, the difference in mean R.T between sprint swimmers and distance swimmers is explored. In this data set, 50m distance is defined as “sprint” and further distances are defined as “distance”. As 50m event is only applicable to freestyle stroke in this Olympics, our analysis will be limited to freestyle events for a fair comparison of swimmer population. A separate JMP table with only this stroke is created (“Swimming_FreestyleOnly”) from the table “Swimming_Final” using the Row > Select Matching Cells function.

Figure 15 shows that:

Observation 5: R.T of female sprinter is better than female distance swimmer

Observation 6: R.T of male sprinter is better than female distance swimmer

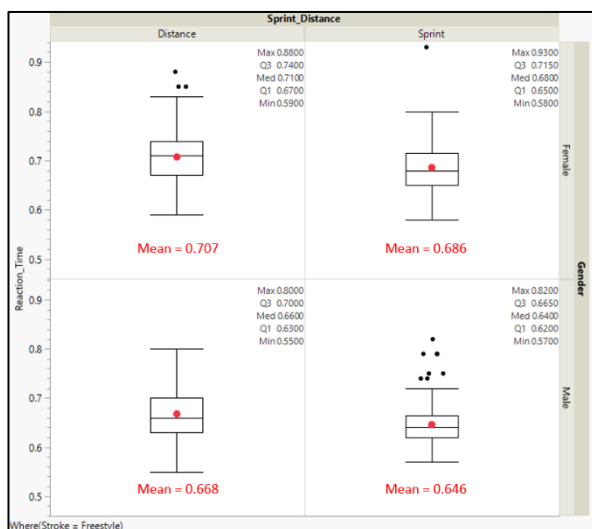


Figure 15: R.T vs Sprint and Distance Events

Similar to previous analyses, normal distribution check is done followed by confirmatory hypothesis testing to check if the above observation is statistically significant. Similar to Hypothesis 0, the following normality test aims at determining the distribution types of sprinter and distance swimmers' RT.

Hypothesis E: sprinter and distance swimmers' RT is normally distributed.

By the distribution and goodness-of-fit test, the following results in Figure 16 are obtained, and summarized in Table 4, based on which the appropriate tests are listed in Table 5.

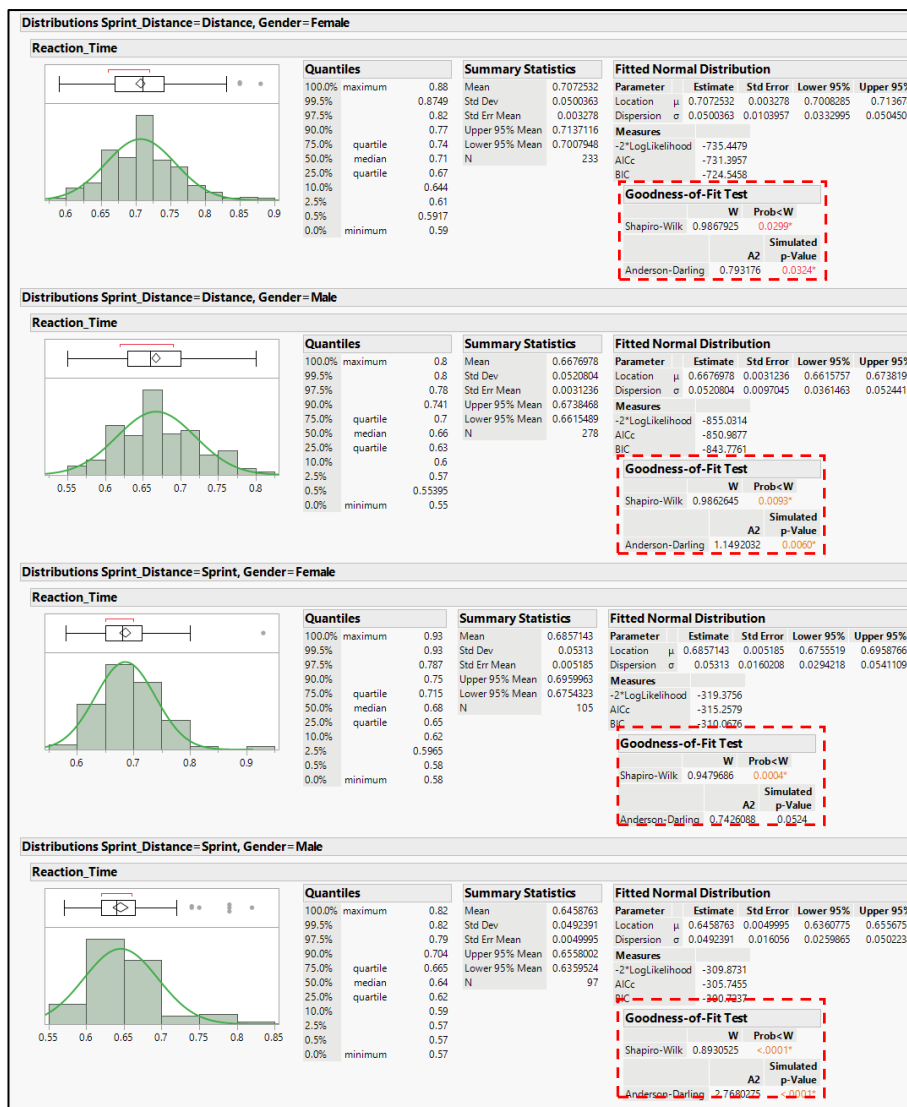


Figure 16: Distribution of R.T in Groups 5-8

Group	Gender	Sprint or Distance	Prob<W (Shapiro-Wilk)	p-value (Anderson-Darling)	Conclusion on H_0
5	Female	Distance	0.0299	0.0324	Accepted
6	Female	Sprint	0.0004*	0.0524	Accepted
7	Male	Distance	0.0093	0.0060	Rejected
8	Male	Sprint	<0.0001	<0.0001	Rejected

Table 4: Normality Test Results for Group 5-8

*Note that the Shapiro-Wilk test does not support the Null hypothesis of normal distribution for this group. Hence both parametric and non parametric tests will be performed for conservatism under hypothesis test F below (Table 5).

Observation No.	Comparison Pairs			Hypothesis Test No.	Type of Hypothesis Test	Null Hypothesis
5	5	vs	6	F	Parametric and Nonparametric	The mean R.T of sprinter and long-distance female swimmers is the same.
6	7	vs	8	G	Nonparametric	The mean R.T of sprinter and long-distance male swimmers is the same.

Table 5: Hypothesis Tests Applied to Group 5-8 Comparison

For hypothesis test F, we are interested to find out if the mean R.T of sprinter and long-distance is the same for female swimmers.

In *Figure 17* Pooled t Test (equal variances assumption), the two-tail test result ($\text{Prob} > |t|$: <0.0004) shows the difference in mean R.T is statistically significant at 99% confidence level. Hence, we reject the null hypothesis that the 2 means are the same. From the left-tail test result ($\text{Prob} < t$: <0.0002), the observation of sprinters' R.T being "less than" long-distance swimmers' R.T is statistically significant, and we can infer that sprinters' mean R.T is lower than that of long-distance swimmers.

Under the Equal Variances test, the p-values across the tests are all greater than the alpha value of 0.01, we cannot reject the null hypothesis of equal variances. As these variances are inferred to be the same, we can maintain our view from the previous pooled T-test.

Under the nonparametric test (Wilcoxon), the p-values are lower than the alpha value, hence we reject the null hypothesis and infer that they have different mean R.T, and that the observed lower R.T from the sprint swimmers is statistically supported.

The same analysis can be done on the male population using nonparametric test for hypothesis G. From the result in *Figure 18*, the p-value is again lower than the specified alpha level of 0.01, hence the null hypothesis can be rejected. Similar inference to that of female population can be drawn that sprinters have lower mean R.T than long-distance swimmers.

Note that swimmers who compete in the 50m events can also compete in the longer distance events. Hence their R.T may reduce the mean R.T among the longer distance population. Despite this, the long-distance population still have statistically longer R.T for both male and female, further reinforcing our inferences above.

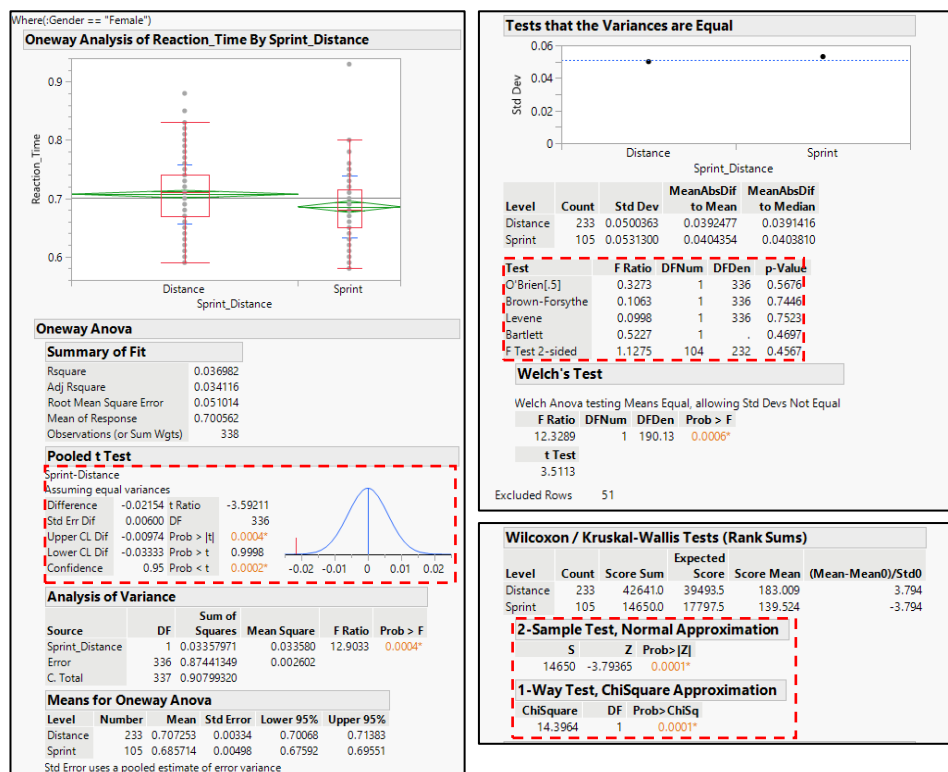


Figure 17: Hypothesis Test F Results for Mean R.T vs Sprint_Distance (Female)

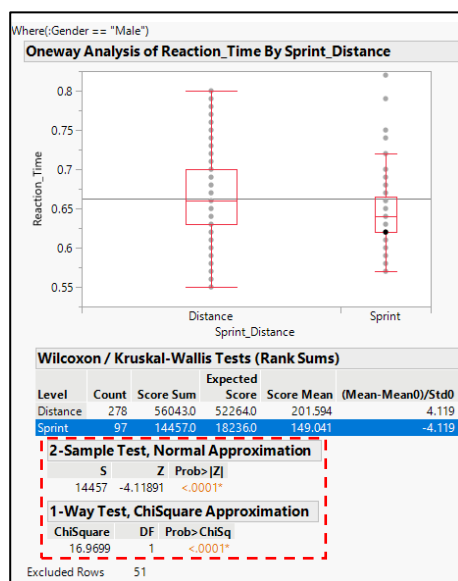


Figure 18: Hypothesis Test G Results for Mean R.T vs Sprint_Distance (Male)

4.2.2 Reaction Time Correlation with Split Performance and Final Placements

In this section, the impact of having short R.T on performance is explored. Specifically, it is reasonable to split the analysis by not only gender, but also the sprint-distance category above. This is because having a short R.T is more critical to sprinters given the lack of race distance to make up for any poor start. Additionally, the metrics for performance is chosen to be the 50m split timing as this is applicable to both groups. By plotting the mean split 50 timing against the R.T, we can form the following observation. Note that the mean (at each R.T increment on the X-axis) is chosen for better illustration, and this does not change the correlation observed.

Observation 7: R.T has a positive correlation to 50m split performance for sprint swimmers

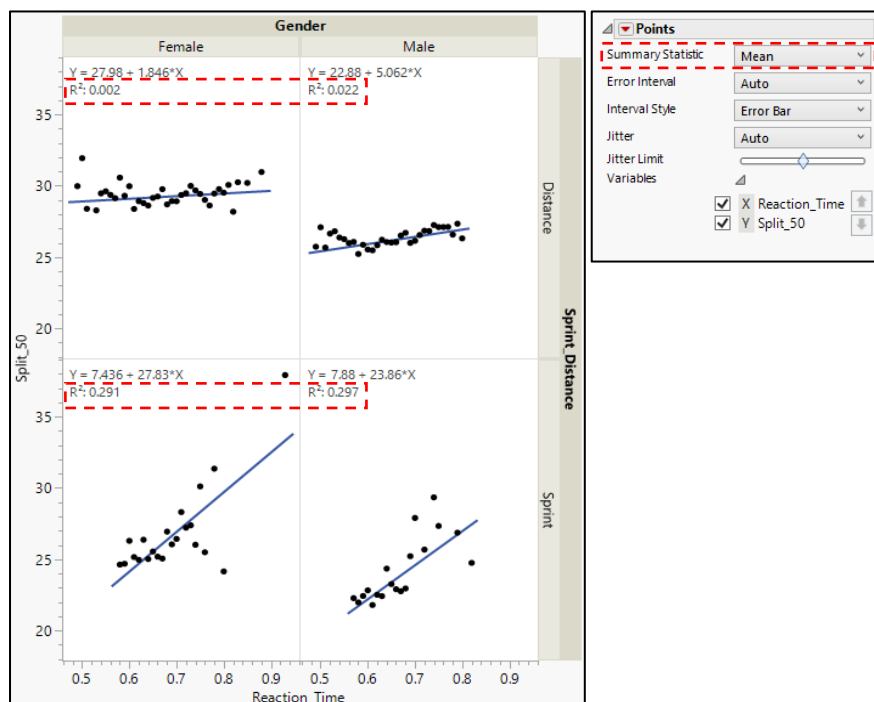


Figure 19: Correlation Between Split 50m Timing and R.T

From Figure 19, while there is a positive linear correlation between the R.T and the split_50 timing performance among male and female sprinters ($R^2 \sim 0.3$), there is almost no linear correlation in longer distance events. This is reasonable as distance swimmers may have other focus areas to improve their final placement, such as form, stamina, swimming tactics... rather than relying on good R.T.

Extending this check to the split timing beyond 50m:

Observation 8: R.T has no correlation to further split performance.

In Figure 20, there is effectively no linear correlation between R.T and split performance in further distances with R^2 being extremely small. We can infer that the early advantage in R.T is not a significant factor to swimmers' timing performance further down the race, as other factors explained above will likely to have a greater relationships to timing performance.



Figure 20: The Lack of Linear Correlation between R.T and Further Split Timings

4.3 Other Aggregate Observations

4.3.1 The Outliers

When the aggregate distribution of male and female R.T is plotted in *Figure 21*:

Observation 9: there are more outliers in the heats, followed by the semifinals, and least in the finals.

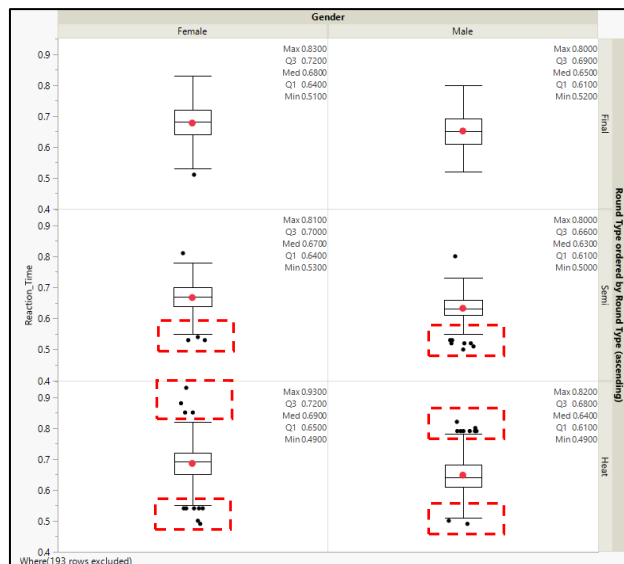


Figure 21: Frequency of Outliers Across Rounds

During heats, swimmers' ability spectrum is the widest as strong medal contenders race with others, many of whom are from countries less developed in the sport. Hence it is likely to find outliers with respect to the quartiles of the population of heat swimmers. As outliers are found in both upper and lower ends of the quartiles, their R.T performance is likely to cancel off each other on average, leading to similar mean and median.

4.3.2 The R.T of Past Good Performers within a Race

In swimming events, athletes with better timing in previous rounds are placed into the middle lanes:

Lane #	1	2	3	4	5	6	7	8
Qualifying Timing Position	7	5	3	1	2	4	6	8

Hence it is reasonable to associate these lanes to high past performers, and their R.T can be placed in comparison with the others. When plotting R.T by the lane positions (*Figure 22*):

Observation 10: past good performing swimmers based on their placement into the middle lanes do not have outstanding R.T.

Despite the previous distinction in R.T between sprinters and long-distance swimmers, when comparing with competitors on the same event, strong contenders are not observed to possess significant advantage in their R.T. This may suggest their ability for winning races do not heavily rely on this metrics.

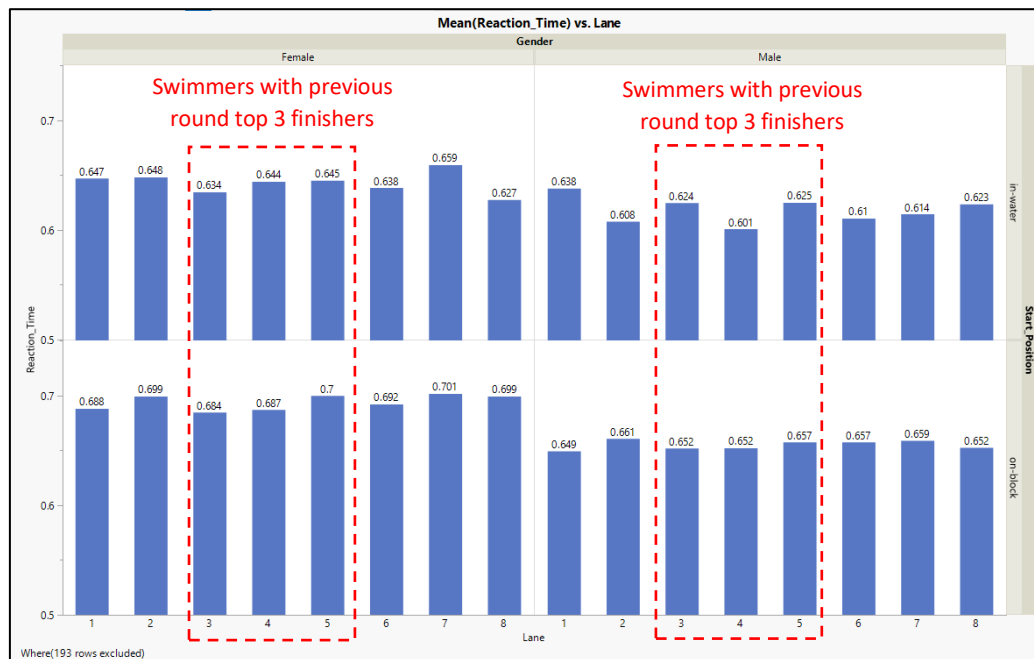


Figure 22: Past Top Performers' R.T vs The Rest

5. RECOMMENDATIONS

From the preceding analysis, based on the perspective of data, the following improvements are recommended:

1. As Tokyo 2020 did not include sprint events other than freestyle into, more sprint data on these events can be gathered from other relevant competitions to explore how R.T correlate with the split 50 timing performance across all strokes for the purpose of future Olympics preparation.
2. The age of the athletes may play an important role to the ability of achieving good R.T and overall performance. Hence, age information can be included for future studies.
3. R.T of individuals on each relay team can affect the overall team performance. Hence, data importing can be improved to cover the missing relays data.
4. As R.T is observed to have positive correlation to sprinter's timing, more in-depth studies on biomechanical factors that could affect the R.T can be explored (posture, body ratios...). This may have implications for tailored training.
5. R.T is one of several constituents of the start of a race, the others being the flight and the slip phase (Thanopoulos, 2012). Hence more data on these phases can be analyzed to discover further how the start performance can affect final timing.

Bibliography

- Morgulev, E. A. (2018). Sports Analytics and the Big-Data Era. *International Journal of Data Science and Analytics volume* , 213-222.
- Thanopoulos, V. R. (2012). Differences in the efficiency between the grab and track starts for both genders in greek young swimmers . *Journal of human kinetics*, 43-51.