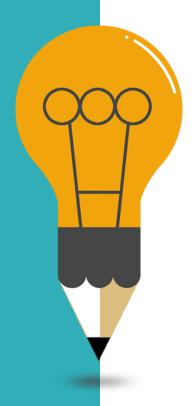


# Assessment 3: Why Not Watch? Business case

**MATH2406 Applied Analytics** 

### Content



01	Introduction
02	Data Analysis
03	Discussion
04	Conclusion & Recommendation



### **GOAL**

To find if the WNW's new algorithm is worth rolling out to all their subscribers by analyzing the results from a recent change they made in their recommendation engine. To find any bias in the data collected and present how this can be correct and to provide recommendations for future A/B Tests



To perform fundamental statistical analyses (descriptive analysis, hypothesis testing, ANOVA, correlation, and linear regression) in order to identify key findings

### Sample Data

01

#### **Date (Interval)**

- Period of observation
- 1/7 31/7

02

#### Age (Ordinal)

- Age of the customer
- Min age 18, Max age 55

03

#### **Gender (Ordinal)**

- Gender of the customer
- F for Female, M for Male

04

#### Social\_metric (Ordinal)

- Combined metric based on previous viewing habits
- 0~10

05

#### Time\_since\_signup (Interval)

- No. of months since the customer signed up
- 0 months 24 months

06

#### **Demographic (Nominal)**

- Demographic number
- 1-4

07

#### Hours\_watched (Ratio)

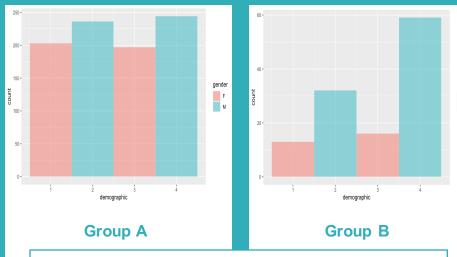
- Number of hours watched in that day
- 0.5 hours 8.3 hours / per day

# Is there any bias in the data?

Inequality in Age & Demographic ratio



- There is inequality in age ratio.
- Absence of certain age range in Group B especially for Female



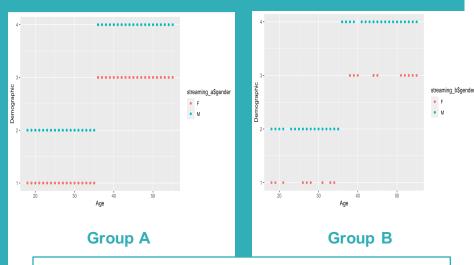
- There is inequality in demographic ratio.
- Group B subscribers from Demographic
   4 has the highest number

# Is there any bias in the data?

Inequality in Gender & Demographic ratio



- There is inequality in gender ratio.
- For group A, 480 males and 400 females
- For group B, 91 males and 29 females



- For Group A & B, certain demographic has only has certain age groups.
- Demographic 1 & 2 only has age group between 18-35 and Demographic 3 & 4 has age group between 35 over.

### How could any bias be corrected?



#### Remove the extreme values from the data

E.g., the very small data close to 0 or very maximum values

#### Sufficient sample size

- Calculate the minimal sample size before launching the test
- Assess the test results only after test reaches the minimal sample size

#### Sampling is completely randomised

- Includes all demographics (Gender, age, etc.)
- Random sampling means that any custome rs of WNW has the same probability to be chosen to see a variation of A/B test

#### Sufficient period of time

 Observation period should be long enough to ensure the sample represents true population



# DATA ANALYSIS:

A/B Testing / Regression

### Two-sample hypothesis test

01

02



Null hypothesis  $H_0: \mu_1 = \mu_2$ 

Two groups A and B have the same efficacy, i.e. that they produce an equivalent number of hours watched in that day.

Alternative hypothesis  $\,H_A:\mu_1
eq\mu_2\,$ 

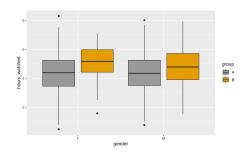
There is a difference in number of hours watched between two groups, i.e. that A and B have different efficacy.

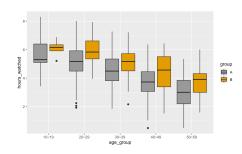
Statistical significance
The statistical significance is

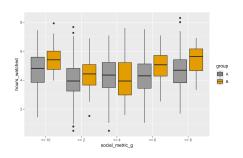
The statistical significance is measured by the p-value, i.e. the probability of observing a discrepancy between our samples at least as strong as the one that we actually observed.

Two-tailed test ha

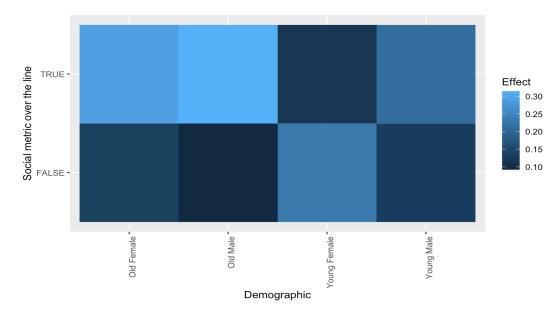
A two-tailed test has been chosen since no reason to know a priori whether the discrepancy between the results of A and B will be in favor of A or B.





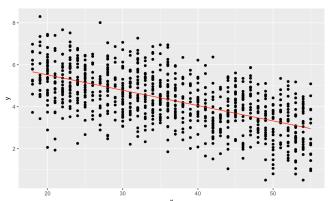


### A/B Test Result

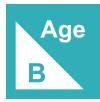


- Demographic defined by the categories: gender, above(social metrics over 5), and
   young(age below 29), old(age above 30)
- Increased hours watched in Group B
- Significant effect shows especially for those whose <u>social metric belongs 6-10</u> with an <u>older demographic</u>.

### Linear Regression

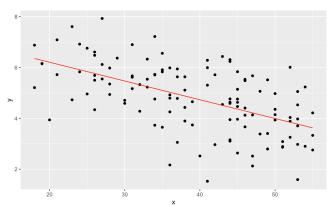






Residual standard error: 1 067 on 878 degrees of freedom
Multiple R-squared: 0.3511, Adjusted R-squared: 0.3503
F-statistic. 475 on 1 and 878 DF. p-value: < 2.2e-16

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

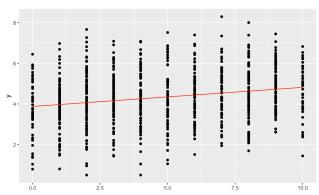


Residual standard error: 1.105 on 118 degrees of freedom

F-statistic: 54.04 on 1 and 118 DF. p-value: 2.806e-11

Multiple R-squared: 0.3141. Adjusted R squared: 0.3083

### Linear Regression







```
2-
0.0 2.5 5.0 7.5 10.0
```

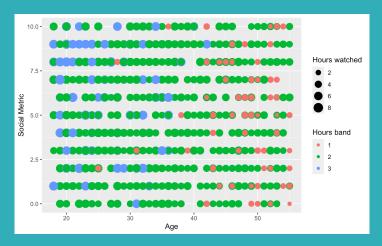
F-statistic: 42.19 on 1 and 878 DF, p-value: 1.386e-10

Residual standard error: 1.247 on 118 degrees of freedom

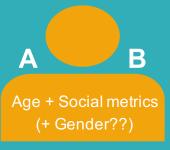
F-statistic. 17.11 on 1 and 118 Dt, p-value: 6.654e-05

Multiple R-squared: 0.1266, Adjusted R-squared: 0.1192

### **Multiple Regression**



```
lm(formula = hours_watched ~ age + social_metric. data = streamina_a)
 Residuals:
              10 Median
 -3.6244 -0.6361 -0.0271 0.6988 2.8773
 Coefficients:
                Estimate Std. Error t value Pr(>|t|)
  (Intercept)
                6.535941
                          0.137147 47.657 < 2e-16 ***
               -0.072279
                          0.003262 -22.157 < 2e-16 ***
 social_metric 0.084869
                          0.011619 7.305 6.25e-13 ***
 Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
 Residual standard error: 1.037 on 877 degrees of freedom
Multiple R-squared: 0.3883, Adjusted R-squared: 0.3869
 F-statistic: 278.3 on 2 and 877 DF, p-value: < 2.2e-16
```

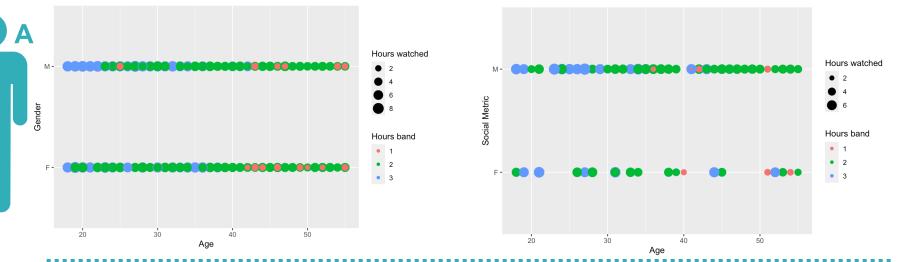




```
lm(formula = hours_watched ~ age + social_metric, data = streamina_b)
     Residuals:
                   10 Median
     -2.65282 -0.61812 0.06309 0.68267 1.80700
     Coefficients:
                   Estimate Std. Error t value Pr(>|t|)
     (Intercept)
                   6.840745
                             0.391174 17.488 < 2e-16 ***
                  -0.075783
                             0.008972 -8.446 9.47e-14 ***
     social_metric 0.176314 0.031714 5.560 1.73e-07 ***
     Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
     Residual standard error: 0.9874 on 117 degrees of freedom
Multiple R-squared: 0.4574, Adjusted R-squared: 0.4482
     F-statistic: 49.32 on 2 and 117 DF, -value: 2.92e-16
```

# Multiple Regression

Group A & B : Age + Gender ... (+ Social metrics??)



```
Analysis of Variance Table
```

```
Model 1: hours_watched ~ age + gender

Model 2: hours_watched ~ gender + age + social_metric

Res.Df RSS Df Sum of Sq F Pr(>F)

1 877 1000.1

2 876 942.9 1 57.254 53.192 6.769e-13 ***

---

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

#### Analysis of Variance Table

```
Model 1: hours_watched ~ age + gender

Model 2: hours_watched ~ age + gender + social_metric

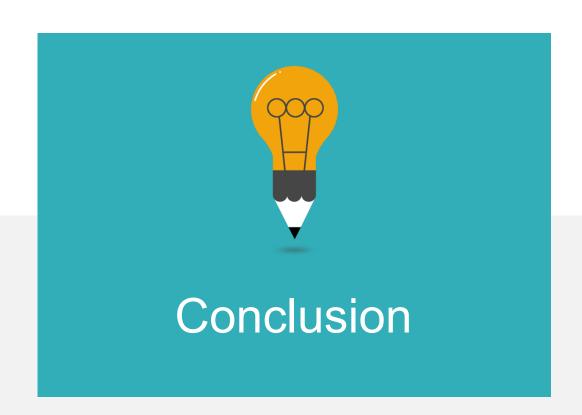
Res.Df RSS Df Sum of Sq F Pr(>F)

1 117 144.20

2 116 114.05 1 30.153 30.668 1.935e-07 ***

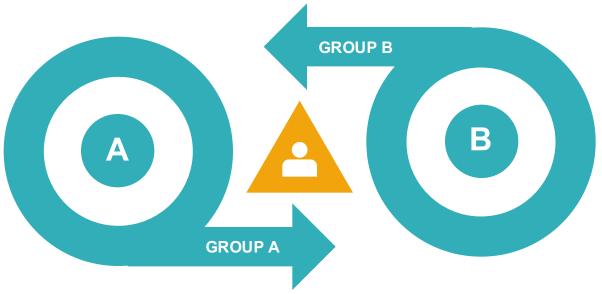
---

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```



After implementing new recommendation engine there has been an **increase** in overall number of hours watched.

Positive effects on customers over the age of 30 with high social metrics



Relationship found between age + gender + social\_metric and hours watched.

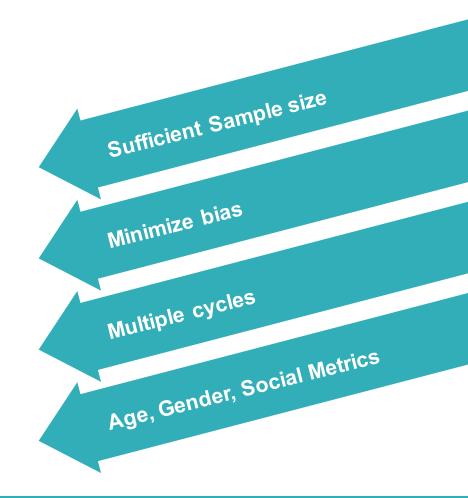
Predict hours watched using a new recommendation engine based on subscribers age, gender and their social metric.

Improvements to be made for future A/B Test:

- 1. Satisfy minimum sample size to make a confident call
- 2. Randomise sampling to gain full representative of the population
- 3. Have sufficient length of time for observation to conclude the effects of the treatment

### RECOMMENDATION

Based on the analysis of given sample data, it is worth rolling out new recommendation engine to all subscribers





Thank you