

InterStats

January 5, 2022

```
[ ]: library(tidyverse)
library(xtable)
library(ggplot2)
library(ggpubr)
library(ggrepel)
library(gridExtra)
library(rstatix)
library(survival)
library(survminer)
library(vcd)
library(finalfit)
library(gtsummary)
```

Read data

```
[2]: surveys1 <- read_csv("Datasets/effec1.quest.compil.csv", locale = locale("fr"),
  ↳ show_col_types = FALSE)
surveys2 <- read_csv("Datasets/effec2.quest.compil.csv", locale = locale("fr"),
  ↳ show_col_types = FALSE)
surveys3 <- read_csv("Datasets/effec3.quest.compil.csv", locale = locale("fr"),
  ↳ show_col_types = FALSE)
usages1 <- distinct(read_csv("Datasets/usages.effec1.csv", locale =
  ↳ locale("fr"), show_col_types = FALSE))
usages2 <- distinct(read_csv("Datasets/usages.effec2.csv", locale =
  ↳ locale("fr"), show_col_types = FALSE))
usages3 <- distinct(read_csv("Datasets/usages.effec3.csv", locale =
  ↳ locale("fr"), show_col_types = FALSE))
```

1 Observation of learners' engagement

```
[3]: # Fix Exam.bin on usages1
usages1 <- usages1 %>% mutate(Exam.bin = case_when(last.quiz == 5 & last.video
  ↳ > 30 ~ 1,
  TRUE ~ 0))

usages1 <- usages1 %>%
```

```

mutate(EngagementLevel = case_when(Exam.bin == 1 & Assignment.bin == 1
  ↪1 ~ 3, # Completers
                                     last.quiz > 0 | Assignment.bin == 1
  ↪== 1 ~ 2, # Disengaging Learners
                                     last.video / 35 > 0.1 ~ 1, #
  ↪Auditing Learners
                                     TRUE ~ 0)) # Bystanders

usages2 <- usages2 %>%
  mutate(EngagementLevel = case_when(Exam.bin == 1 & Assignment.bin == 1
  ↪1 ~ 3, # Completers
                                     last.quiz > 0 | Assignment.bin == 1
  ↪== 1 ~ 2, # Disengaging Learners
                                     last.video / 35 > 0.1 ~ 1, #
  ↪Auditing Learners
                                     TRUE ~ 0)) # Bystanders

usages3 <- usages3 %>%
  mutate(EngagementLevel = case_when(Exam.bin == 1 & Assignment.bin == 1
  ↪1 ~ 3, # Completers
                                     last.quiz > 0 | Assignment.bin == 1
  ↪== 1 ~ 2, # Disengaging Learners
                                     last.video / 35 > 0.1 ~ 1, #
  ↪Auditing Learners
                                     TRUE ~ 0)) # Bystanders

```

```

[4]: df1 <- usages1 %>% group_by(EngagementLevel) %>% tally(name = "value") %>%
  mutate(csum = rev(cumsum(rev(value))),
         pos = if_else(is.na(value/2 + lead(csum, 1)), value/2, value/2 +
  ↪lead(csum, 1)))

plot1 <- ggplot(df1, aes(x = "", y = value, fill = factor(EngagementLevel))) +
  geom_col(width = 1, color = 1) +
  coord_polar(theta = "y") +
  geom_label_repel(data = df1,
                  aes(y = pos, label = paste0(round(value / sum(value) * 100,
                  digits = 2), "%")),
                  size = 4.5, nudge_x = 1, show.legend = FALSE) +
  scale_fill_brewer(palette = "Pastel2", name = "Engagement level",
                  labels = c('Bystanders', 'Auditing Learners',
                  'Disengaging Learners', 'Completers')) +

  theme_void() + labs(title = expression('1'^st* iteration'))

df2 <- usages2 %>% group_by(EngagementLevel) %>% tally(name = "value") %>%
  mutate(csum = rev(cumsum(rev(value))),

```

```

      pos = if_else(is.na(value/2 + lead(csum, 1)), value/2, value/2 +
↪lead(csum, 1)))

plot2 <- ggplot(df2, aes(x = "" , y = value, fill = factor(EngagementLevel))) +
  geom_col(width = 1, color = 1) +
  coord_polar(theta = "y") +
  geom_label_repel(data = df2,
    aes(y = pos, label = paste0(round(value / sum(value) * 100,
                                   digits = 2), "%")),
    size = 4.5, nudge_x = 1, show.legend = FALSE) +
  scale_fill_brewer(palette = "Pastel2", name = "Engagement level",
    labels = c('Bystanders', 'Auditing Learners',
               'Disengaging Learners', 'Completers')) +

  theme_void() + labs(title = expression('2'^nd* ' iteration'))

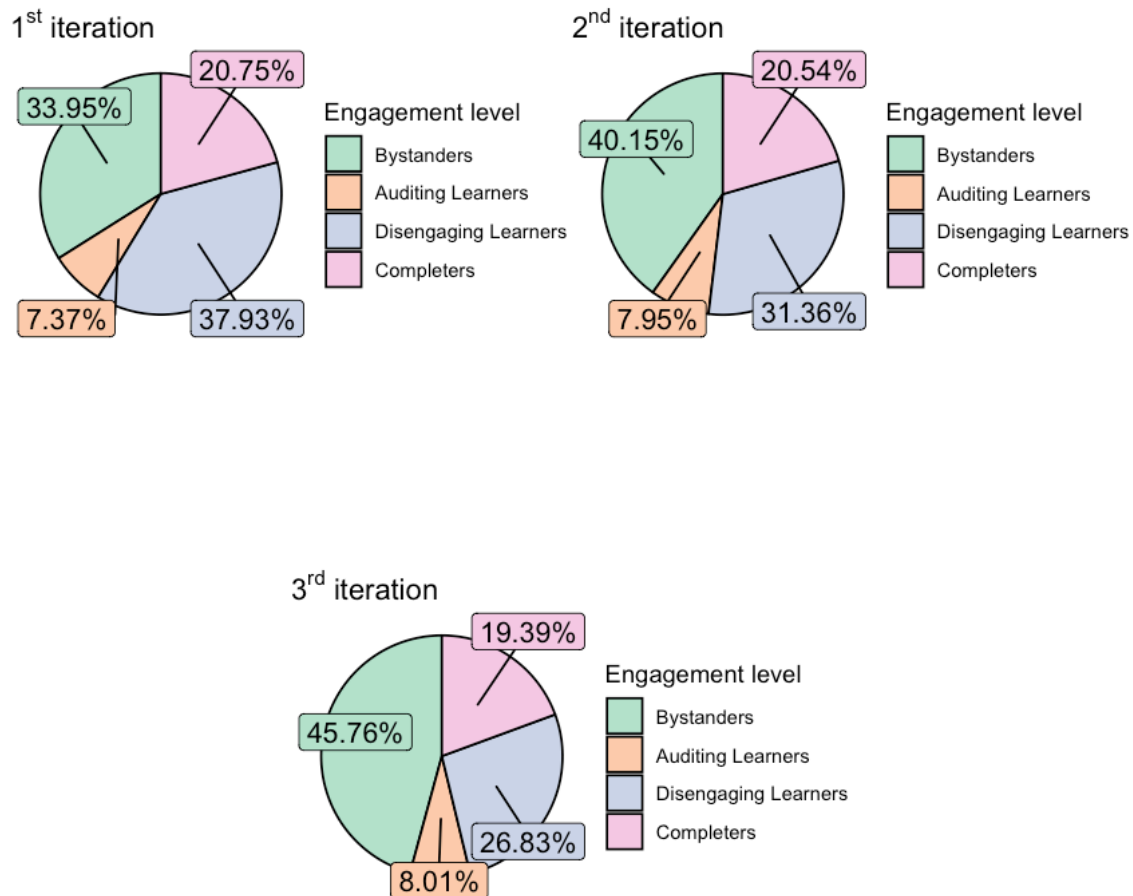
df3 <- usages3 %>% group_by(EngagementLevel) %>% tally(name = "value") %>%
  mutate(csum = rev(cumsum(rev(value))),
    pos = if_else(is.na(value/2 + lead(csum, 1)), value/2, value/2 +
↪lead(csum, 1)))

plot3 <- ggplot(df3, aes(x = "" , y = value, fill = factor(EngagementLevel))) +
  geom_col(width = 1, color = 1) +
  coord_polar(theta = "y") +
  geom_label_repel(data = df3,
    aes(y = pos, label = paste0(round(value / sum(value) * 100,
                                   digits = 2), "%")),
    size = 4.5, nudge_x = 1, show.legend = FALSE) +
  scale_fill_brewer(palette = "Pastel2", name = "Engagement level",
    labels = c('Bystanders', 'Auditing Learners',
               'Disengaging Learners', 'Completers')) +

  theme_void() + labs(title = expression('3'^rd* ' iteration'))

g <- arrangeGrob(plot1, plot2, plot3, ncol = 2, layout_matrix=rbind(c(1,1,2,2),
↪c(NA, 3, 3, NA)))
grid.arrange(plot1, plot2, plot3, ncol = 2, layout_matrix=rbind(c(1,1,2,2),
↪c(NA, 3, 3, NA)))

```



2 Inferential Statistics

Prepare data to apply hypothesis tests

```
[ ]: # Merge data
iter1 <- distinct(left_join(surveys1, usages1, by='Student_ID'))
iter2 <- distinct(left_join(surveys2, usages2, by='Student_ID'))
iter3 <- distinct(left_join(surveys3, usages3, by='Student_ID'))
iter3 <- iter3 %>% mutate_at(c('Curiosity.MOOC', 'Rencontres'), as.numeric)
df <- iter1 %>% full_join(iter2) %>% full_join(iter3)
```

```
[6]: df <- df %>% mutate(Country_HDI = case_when(Country_HDI == "TH" ~ "Very High",
                                                Country_HDI == "H" | Country_HDI == "M" ~ "Intermediate",
                                                Country_HDI == "B" ~ "Low"))
df$Country_HDI <- factor(df$Country_HDI, levels = c("Very High", "Intermediate", "Low"))

df <- df %>% mutate(Gender = case_when(Gender == "un homme" ~ "Male",
                                       Gender == "une femme" ~ "Female"))
df$Gender <- factor(df$Gender, levels = c("Male", "Female"))
df$Exam.bin <- factor(df$Exam.bin)
```

2.1 Gender and Viewed videos

```
[7]: df.t_test <- df %>% drop_na(c(Gender, last.video))
df.t_test %>%
  group_by(Gender) %>%
  get_summary_stats(last.video, type = "mean_sd")
```

A tibble: 2 × 5

Gender	variable	n	mean	sd
<fct>	<chr>	<dbl>	<dbl>	<dbl>
Male	last.video	6103	20.843	13.458
Female	last.video	2990	20.773	13.912

```
[8]: # Independent sample t-test
stat.test <- df.t_test %>%
  t_test(last.video ~ Gender)
stat.test
```

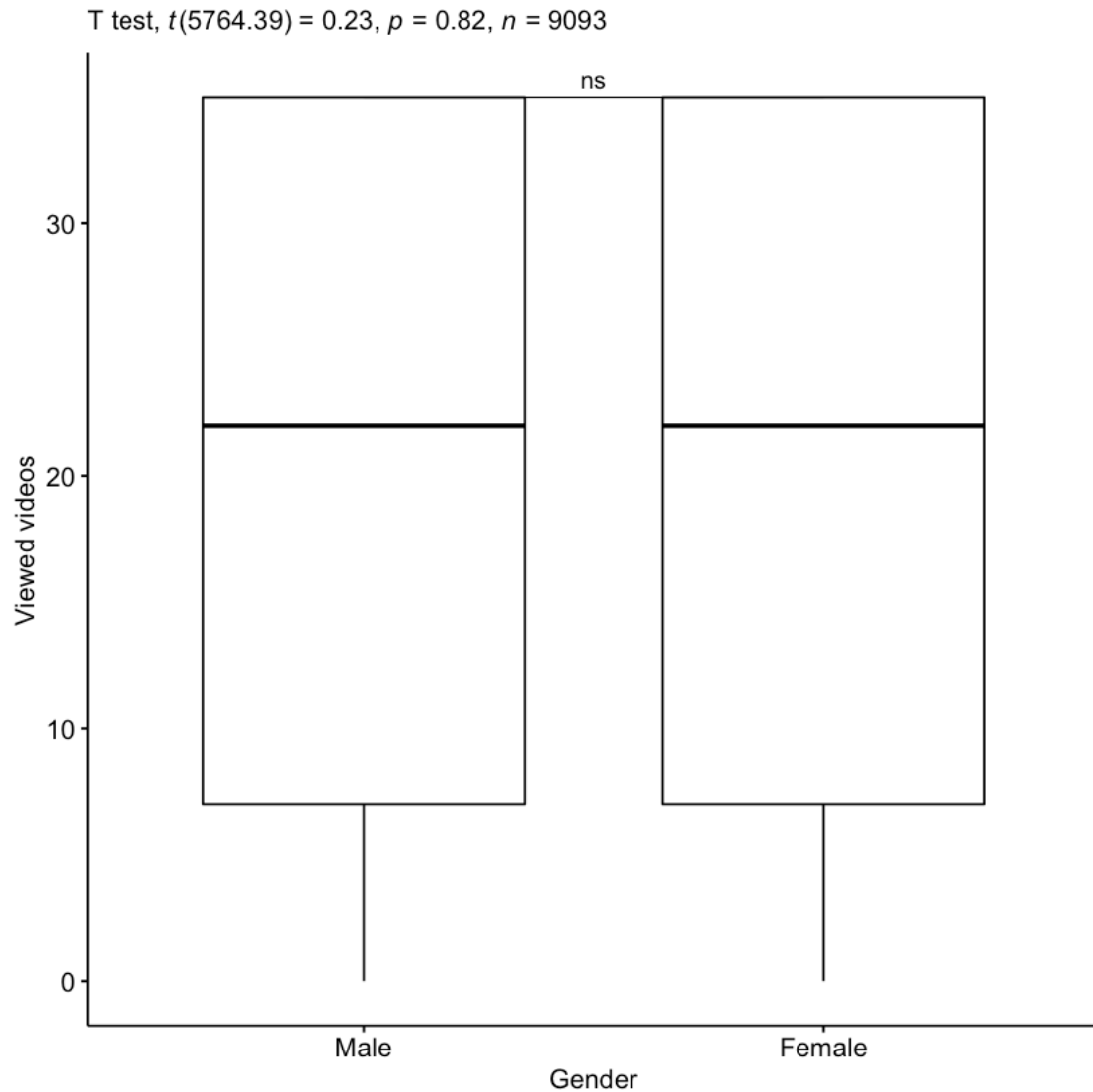
A rstatix_test: 1 × 8

	.y.	group1	group2	n1	n2	statistic	df	p
	<chr>	<chr>	<chr>	<int>	<int>	<dbl>	<dbl>	<dbl>
1	last.video	Male	Female	6103	2990	0.2292759	5764.39	0.819

```
[9]: # Create a box-plot
bxp <- ggboxplot(
  df.t_test, x = "Gender", y = "last.video",
  ylab = "Viewed videos", xlab = "Gender")

# Add p-value and significance levels
stat.test <- stat.test %>% add_xy_position(x = "Gender") %>%
  add_significance()

bxp +
  stat_pvalue_manual(stat.test, tip.length = 0) +
  labs(subtitle = get_test_label(stat.test, detailed = TRUE))
```



2.2 Country's HDI and Viewed videos

```
[10]: df.anova <- df %>% drop_na(c(Country_HDI, last.video))
df.anova %>%
  group_by(Country_HDI) %>%
  get_summary_stats(last.video, type = "mean_sd")
```

A tibble: 3 × 5

Country_HDI <fct>	variable <chr>	n <dbl>	mean <dbl>	sd <dbl>
Very High	last.video	7264	21.665	13.580
Intermediate	last.video	667	18.099	13.846
Low	last.video	1032	16.633	12.638

```
[11]: # Levene test
levene <- df.anova %>% levene_test(last.video ~ Country_HDI)
levene
```

```
A tibble: 1 × 4
```

	df1	df2	statistic	p
	<int>	<int>	<dbl>	<dbl>
	2	8960	11.85427	7.220328e-06

```
[12]: # Welch One way ANOVA test
res.aov <- df.anova %>% welch_anova_test(last.video ~ Country_HDI)
res.aov
```

```
A rstatix_test: 1 × 7
```

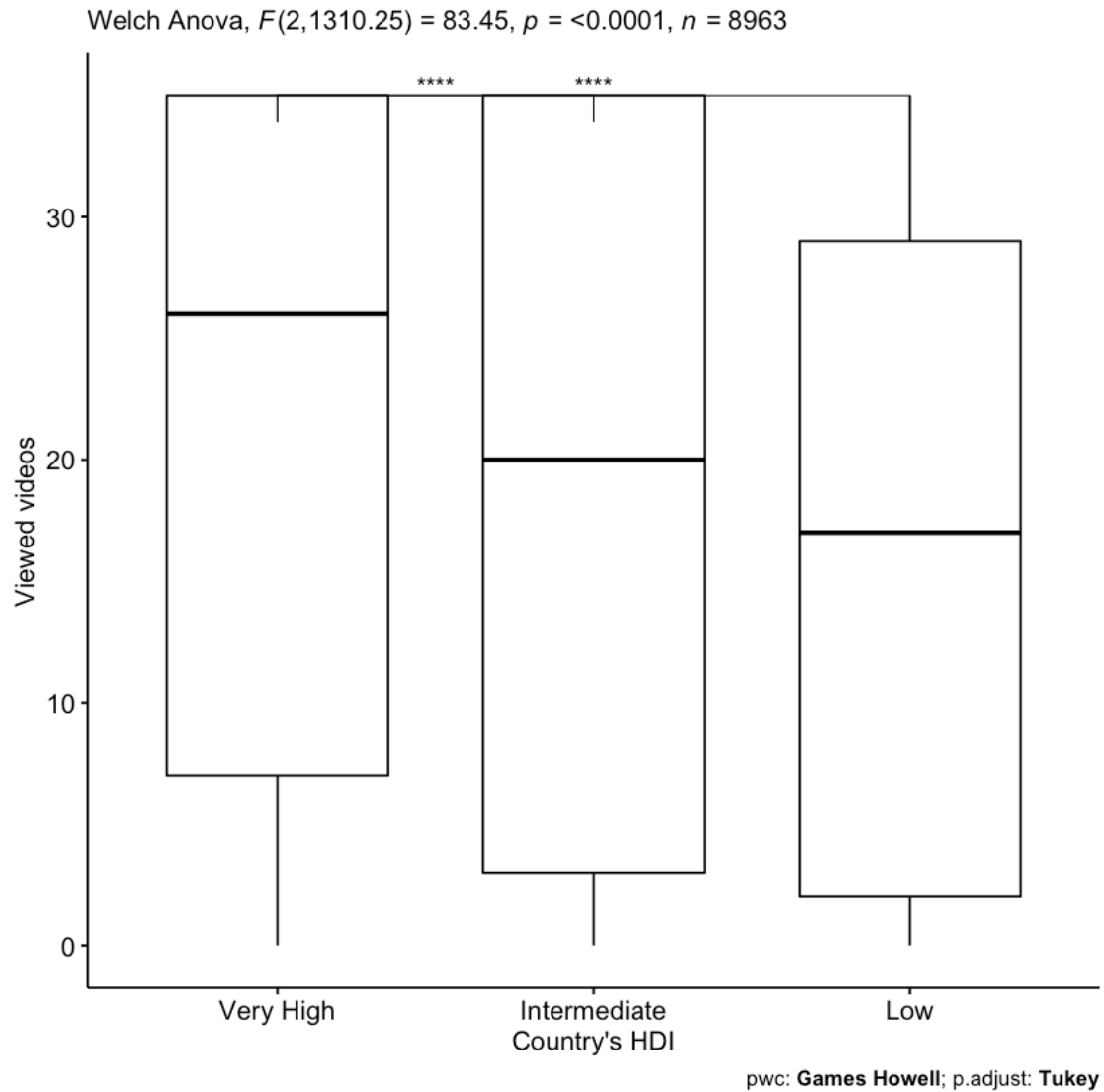
	.y.	n	statistic	DFn	DFd	p	method
	<chr>	<int>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>
1	last.video	8963	83.45	2	1310.252	7.7e-35	Welch ANOVA

```
[13]: # Pairwise comparisons (Games-Howell)
pwc <- df.anova %>% games_howell_test(last.video ~ Country_HDI)
pwc
```

```
A rstatix_test: 3 × 8
```

	.y.	group1	group2	estimate	conf.low	conf.high	p.adj
	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
1	last.video	Very High	Intermediate	-3.566385	-4.879709	-2.25306145	0.000
2	last.video	Very High	Low	-5.032584	-6.028394	-4.03677382	0.000
3	last.video	Intermediate	Low	-1.466199	-3.026445	0.09404754	0.071

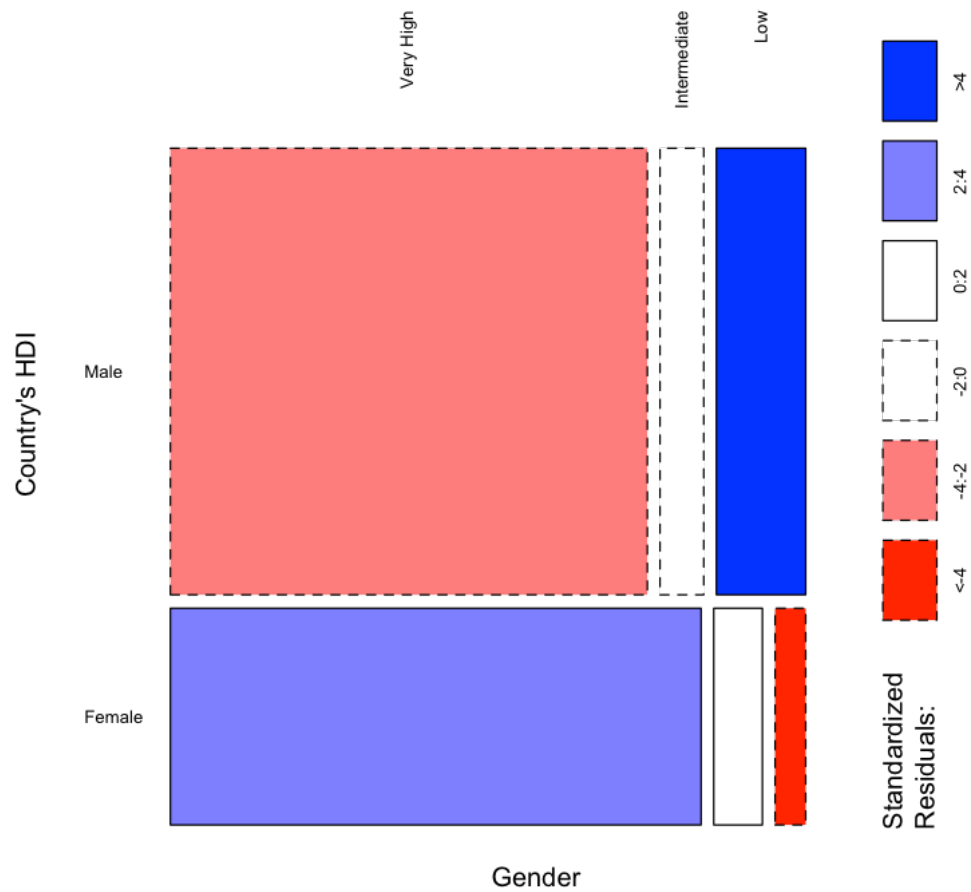
```
[14]: # Visualization: box plots with p-values
pwc <- pwc %>% add_xy_position(x = "Country_HDI")
ggboxplot(df.anova, x = "Country_HDI", y = "last.video",
  ylab = "Viewed videos", xlab = "Country's HDI") +
  stat_pvalue_manual(pwc, hide.ns = TRUE) +
  labs(
    subtitle = get_test_label(res.aov, detailed = TRUE),
    caption = get_pwc_label(pwc)
  )
```



2.3 Genders and Country's HDI groups

```
[15]: df.chisq <- df %>% drop_na(c(Country_HDI, Gender))
```

```
[16]: # Mosaic plot
mosaicplot( ~ Gender + Country_HDI, data = df.chisq,
  shade = TRUE, las=2, dir=c("h", "v"),
  main = NULL,
  xlab = "Gender", ylab = "Country's HDI")
```

```
[17]: # Chi-squared test
chisq <- chisq_test(df.chisq$Gender, df.chisq$Country_HDI)
chisq
```

A rstatix_test: 1 × 6	n	statistic	p	df	method	p.signif
	<int>	<dbl>	<dbl>	<int>	<chr>	<chr>
1	8957	179.0476	1.32e-39	2	Chi-square test	****

2.4 Course accomplishment and viewed videos

Prepare data

```
[18]: df <- df %>% mutate(status = case_when(last.video < 30 ~ 1,
                                             TRUE ~ 0))
```

2.4.1 Regression models

```
[19]: # Logistic regression
lr <- glm(Exam.bin ~ Country_HDI + Gender, data=df, family = "binomial")
summary(lr)
```

Call:

```
glm(formula = Exam.bin ~ Country_HDI + Gender, family = "binomial",
    data = df)
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.0553	-1.0553	-0.9008	1.3047	1.6997

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-0.29429	0.02882	-10.213	< 2e-16 ***
Country_HDIIntermediate	-0.38536	0.08550	-4.507	6.57e-06 ***
Country_HDILow	-0.86831	0.07755	-11.197	< 2e-16 ***
GenderFemale	-0.01281	0.04662	-0.275	0.783

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

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 12030 on 8950 degrees of freedom
 Residual deviance: 11877 on 8947 degrees of freedom
 (7969 observations deleted due to missingness)
 AIC: 11885

Number of Fisher Scoring iterations: 4

```
[20]: # OR plot
df %>%
  or_plot("Exam.bin", c("Gender", "Country_HDI"), table_text_size=4,
  ↪title_text_size=14,
  plot_opts=list(xlab("OR, 95% CI"), theme(axis.title =
  ↪element_text(size=12))))
```

Note: dependent includes missing data. These are dropped.

Waiting for profiling to be done...

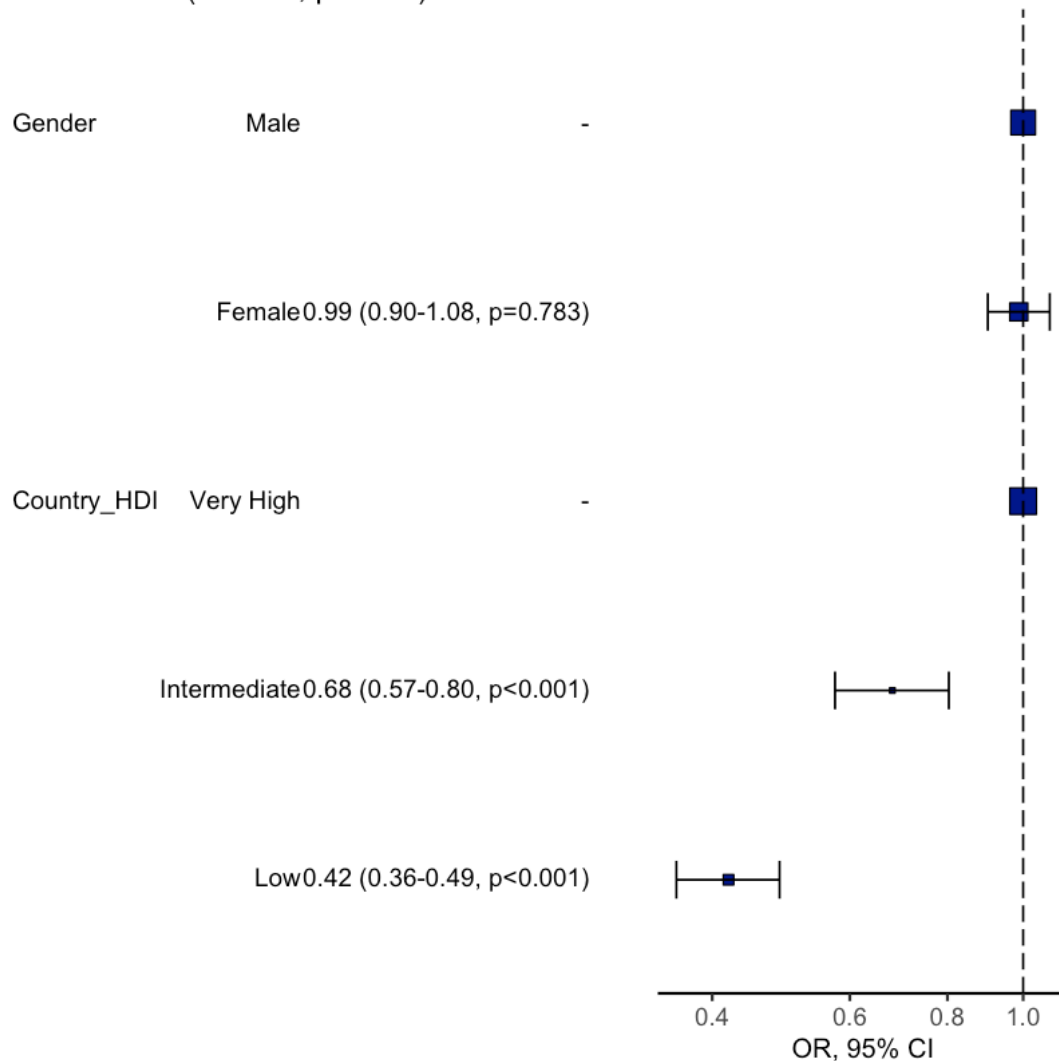
Waiting for profiling to be done...

Waiting for profiling to be done...

Warning message:

"Removed 2 rows containing missing values (geom_errorbarh)."

Exam.bin: OR (95% CI, p-value)



```
[21]: # Poisson Regression
pr <- glm(last.video ~ Country_HDI + Gender, data=df, family = "poisson")
pr
```

Call: glm(formula = last.video ~ Country_HDI + Gender, family = "poisson", data = df)

Coefficients:

(Intercept)	Country_HDIIntermediate	Country_HDILow
3.08560	-0.18080	-0.26909
GenderFemale		
-0.02596		

Degrees of Freedom: 8950 Total (i.e. Null); 8947 Residual
(7969 observations deleted due to missingness)

Null Deviance: 104100

Residual Deviance: 102600 AIC: 141100

2.4.2 Survival Analysis

```
[22]: # Reference: Very High + Male
surv = coxph(formula = Surv(last.video, status) ~ Country_HDI + Gender, data = df)
summary(surv)
```

Call:

```
coxph(formula = Surv(last.video, status) ~ Country_HDI + Gender,
      data = df)
```

n= 8951, number of events= 5338
(7969 observations deleted due to missingness)

	coef	exp(coef)	se(coef)	z	Pr(> z)
Country_HDIIntermediate	0.34175	1.40741	0.04955	6.897	5.32e-12 ***
Country_HDILow	0.56717	1.76327	0.03870	14.656	< 2e-16 ***
GenderFemale	0.01706	1.01720	0.02991	0.570	0.569

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

	exp(coef)	exp(-coef)	lower .95	upper .95
Country_HDIIntermediate	1.407	0.7105	1.2771	1.551
Country_HDILow	1.763	0.5671	1.6345	1.902
GenderFemale	1.017	0.9831	0.9593	1.079

Concordance= 0.54 (se = 0.004)

Likelihood ratio test= 220 on 3 df, p=<2e-16

Wald test = 243.6 on 3 df, p=<2e-16

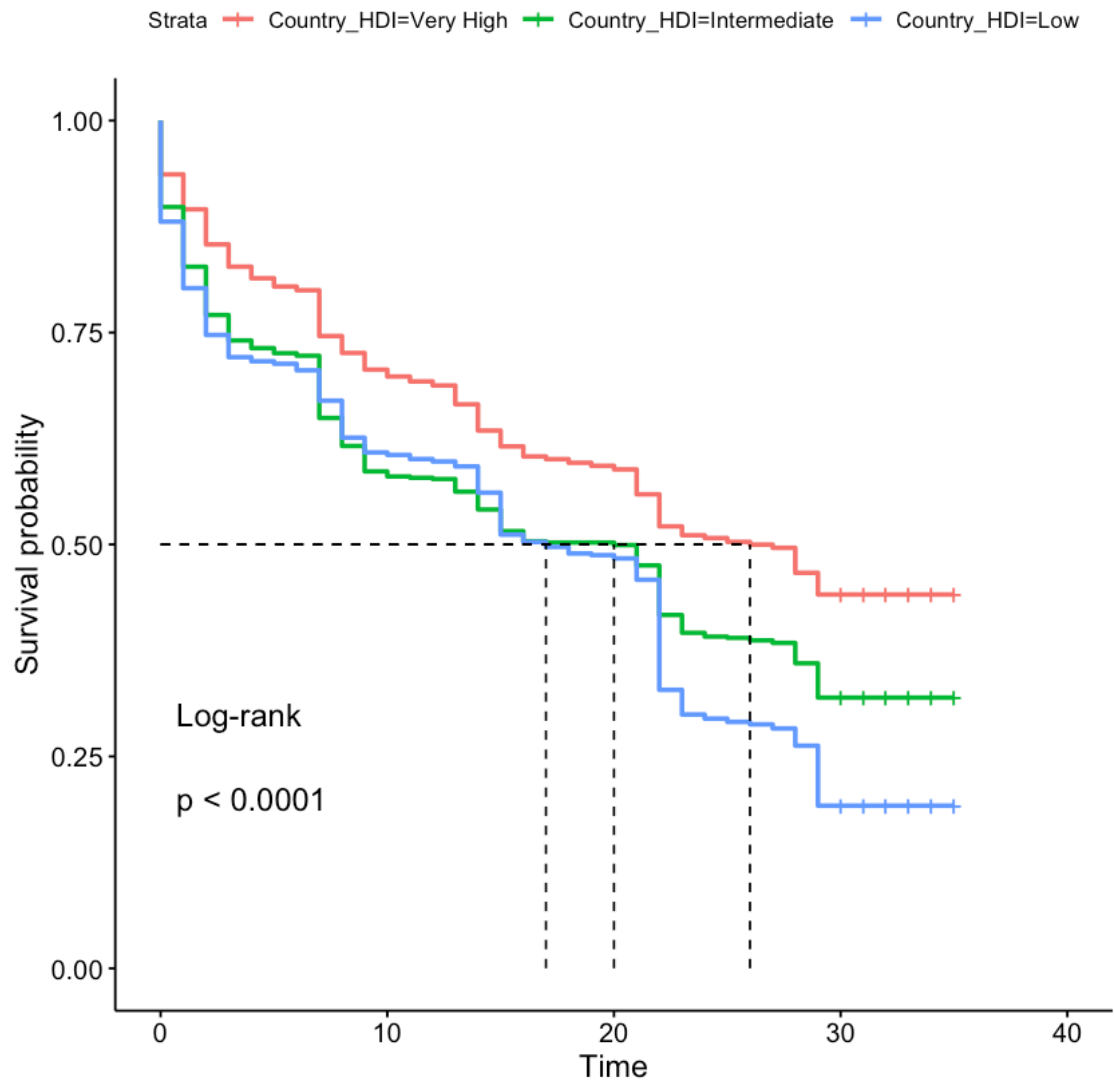
Score (logrank) test = 249.3 on 3 df, p=<2e-16

```
[23]: # Survival plot of Country's HDI
survival_HDI <- survfit(Surv(last.video, status) ~ Country_HDI, data = df)
ggsurvplot(
  survival_HDI,
```

```

conf.int = FALSE,
surv.median.line = c('hv'),
data = df,
pval = TRUE,
pval.method = TRUE,
risk.table = FALSE)

```

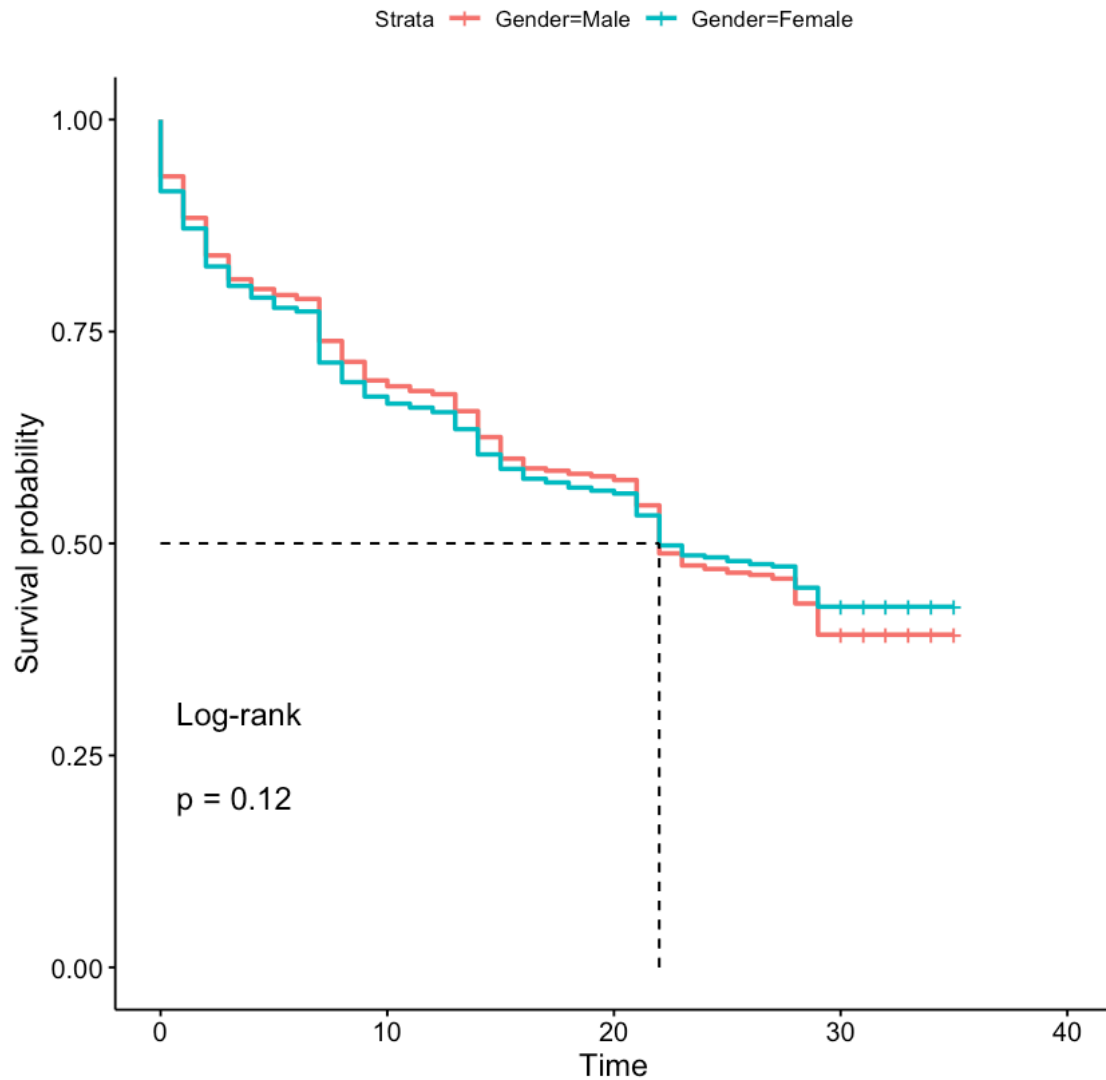


```

[24]: # Survival plot of Gender
survival_Group <- survfit(Surv(last.video, status) ~ Gender, data = df)
ggsurvplot(
  survival_Group,
  conf.int = FALSE,
  surv.median.line = c('hv'),

```

```
data = df,
pval = TRUE,
pval.method = TRUE,
risk.table = FALSE)
```



```
[25]: # Merge table
t1 <- lr %>% tbl_regression(exponentiate = TRUE)
t2 <- surv %>% tbl_regression(exponentiate = TRUE)
tbl_merge <-
  tbl_merge(
    tbls = list(t1, t2),
    tab_spanner = c("**Course completion**", "**View videos to Quit**")
  )
```

```
[26]: tbl_merge %>%  
      as_gt() %>%  
      gt::gtsave(filename = "plots/merged_table.png")
```

```
[ ]:
```