R Statistical Analysis Name: Yi 2024-07-01

```
library(ggplot2)
```

##

##

filter, lag ## ## ##

programming languages.

##

Max.

[1] 1.11312

[1] 0.7938829

"perceptions")])

:1.624

Max.

library(tidyselect) library(dplyr)

Attaching package: 'dplyr'

The following objects are masked from 'package:stats': intersect, setdiff, setequal, union

The following objects are masked from 'package:base': World Happiness Statistical Analysis The World Happiness Report consists of data on global happiness from over 156 countries released by the United Nations. The report has six indicators affecting the happiness index: GDP per capita, social support, healthy life expectancy, freedom, generosity, and corruption perception. The project aims to use filtering techniques,

statistical modeling, data clustering, and regression models to examine the data set with the LaTeX and R

import the datasets, rename the variables, and show the top ten happiness <- read.csv("world happiness.csv",sep=",")</pre>

• Load the world happiness.csv file into R and rename the variables with appropriate headers. colnames(happiness) <- c("rank", "country", "score", "income", "support", "health", "freedo</pre>

0.393

0.410

0.341

m", "generosity", "perceptions") rank Finland 7.769 1.340 1.587 0.986 0.596 0.153 Denmark 7.600 1.383 1.573 0.996 0.592 0.252

head(happiness, 10) ## ## 1 ## 2 ## 3 3 Norway 7.554 1.488 0.603 0.271 1.582 1.028 0.354 ## 4 Iceland 7.494 1.380 0.591 1.624 1.026

country score income support health freedom generosity perceptions ## 5 5 Netherlands 7.488 1.396 1.522 0.999 0.557 0.322 ## 6 6 Switzerland 7.480 1.452 0.572 1.526 1.052 0.263

0.118 0.298 0.343 ## 7 0.574 Sweden 7.343 1.387 1.487 1.009 0.267 0.373 ## 8 8 New Zealand 7.307 1.303 1.557 1.026 0.585 0.380 0.330 ## 9 Canada 7.278 9 1.365 1.505 1.039 0.584 0.285 0.308 ## 10 Austria 7.246 1.376 1.475 1.016 0.532 0.244 0.226 10 ncol(happiness) ## [1] 9

 Display the number of attributes in the World Happiness database counted by columns and rows. nrow(happiness)

[1] 156

• Generate an overview with the summary function and specify whether there are zero values.

summary(happiness)

country rank income score

Length: 156 Min. 1.00 Min. :0.0000 • Min. :2.853 ## 1st Qu.: 39.75 Class :character 1st Qu.:4.545 1st Qu.:0.6028 ## Median : 78.50 Mode :character Median :5.380 Median :0.9600 ## Mean : 78.50 :5.407 Mean Mean :0.9051 ## 3rd Qu.:117.25 3rd Qu.:6.184 3rd Qu.:1.2325

Max. :156.00 :7.769 :1.6840 Max. Max. ## health support freedom generosity ## :0.0000 Min. :0.000 Min. Min. :0.0000 Min. :0.0000 ## 1st Qu.:1.056 1st Qu.: 0.5477 1st Qu.:0.3080 1st Qu.: 0.1087 ## Median :1.272 Median :0.7890 Median : 0.4170 Median :0.1775 ## :1.209 :0.7252 Mean :0.3926 :0.1848 Mean Mean Mean ## 3rd Qu.:1.452 3rd Qu.: 0.8818 3rd Qu.: 0.5072 3rd Qu.: 0.2482

Max.

:1.1410

perceptions ## Min. :0.0000 ## 1st Qu.:0.0470 ## Median :0.0855 ## Mean :0.1106 ## 3rd Qu.: 0.1412 ## Max. :0.4530 sum(is.na(happiness)) ## [1] 0 Calculate the mean and standard deviation for the World Happiness database.

:0.6310

Max.

:0.5660

 $\sigma = \sqrt{\frac{1}{n} \sum_{i=1}^{n} (x_i - \mu)^2}$. The low sd value of 1.11312 means data points are closer to the mean compared to points with +2 or +3 deviation. mean(happiness\$score) ## [1] 5.407096 sd(happiness\$score)

The mean is the average calculated by adding up all the values and dividing by the total numbers using the

formula $\bar{x} = \frac{1}{n} \sum_{i=1}^{n} x_i$. The standard deviation (sd) is the distance from the mean with the formula

• Display the correlation between the World Happiness attributes in the datasets. The Pearson correlation coefficient measures the correlation between two values. The values closer to 1 present a strong relationship, and a correlation closer to 0 suggests a weak relationship. The example shows a strong correlation between the happiness score and GDP per capita, suggesting that more income leads to a higher level of happiness. cor(happiness\$score, happiness\$income)

Included Plots • Create a pairwise scatterplot in R that displays the relationship between the continuous variables. The scatterplot shows a positive correlation between GDP per capita and the attributes (social support, healthy life expectancy, freedom, and corruption perceptions). The linear slope (upwards or downwards) between the data points indicates a positive correlation. The points that are scattered across the plot suggest a weaker relationship.

pairs(happiness[c("rank", "score", "income", "support", "health", "freedom", "generosity",

0.0 0.3 0.6

0.0

0.3

0.0

income

0.0

rank

0.0

100

0 -

freedom 0.0 generosity perceptions

Create a scatter plot and histogram to show the relationship between the variables.

0.0

support

health

 The histogram displays the distribution of the happiness score based on frequency. The Normal Distribution shows that the data points are symmetrically distributed around the center, and the standard deviation is spread evenly. The normal distribution suggests that most countries have an average happiness score of 5 to 7. # Histogram to show the frequency of happiness ggplot(happiness, aes(x = score)) +geom_histogram(binwidth = 0.5, col='gray',) + labs(title = "Distribution of the Happiness Score", x = "Happiness Score", y = "Freque ncy") Distribution of the Happiness Score 30 -20 adnency Fre 10 -

6

Happiness Score

The regression model shows the relationship between the dependent and the independent variables. The

distribution of the residuals is near zero indicating that the data point is approximate to a linear model. The low coefficient p-value implies that the model is statistically significant. The high Multiple R-squared values

Create a regression analysis between the variables in the database.

of 0.7425 and Adjusted R-squared of 0.7374 indicate a well-fit model.

model <- lm(score ~ income + health + freedom, data = happiness)</pre>

Create a clustering model to display the relationship between the variables.

happiness\$cluster <- as.factor(clusters\$cluster)</pre>

Clustering of Health vs Happiness

0.3

set.seed(123)

7 -

6 -

4 -

3 -

2 -

Noway

|celand

Hong Kong, and Japan followed by Europe countries.

summarise(mean_happiness = mean(health))

Denmark -

country happiness <- happiness %>%

top regions <- country happiness %>% arrange(desc(mean_happiness)) %>%

group by(country) %>%

head(10)

0.3 -

0.0 -

freedom, and safety.

0.0

score

geom point() +

attributes like social support, and healthy life expectancy also increase.

ggplot(happiness, aes(x = health, y = score, color = cluster)) +

summary(model) ## ## Call: ## lm(formula = score ~ income + health + freedom, data = happiness) ## ## Residuals: ## Min 1Q Median 3Q Max ## -1.94349 -0.35976 0.07486 0.42184 ## ## Coefficients: ## Estimate Std. Error t value Pr(>|t|)## (Intercept) 2.4201 0.1667 14.519 < 2e-16 *** ## income 1.1781 0.2104 5.599 9.84e-08 *** ## health 1.4578 0.3480 4.189 4.73e-05 *** 6.298 3.09e-09 *** ## freedom 2.1993 0.3492 ## 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 ## Signif. codes: ## ## Residual standard error: 0.5704 on 152 degrees of freedom ## Multiple R-squared: 0.7425, Adjusted R-squared: 0.7374 ## F-statistic: 146.1 on 3 and 152 DF, p-value: < 2.2e-16 model ## ## Call: ## lm(formula = score ~ income + health + freedom, data = happiness) ## ## Coefficients: (Intercept) health freedom ## income ## 2.420 1.178 1.458 2.199

• The K-means clustering model aims to group similar cluster points in the World Happiness database. The

expectancy are scattered in a linear direction. This suggests that as the GDP Per Capita increases, other

0.9

cluster

3

7.3

1.14

1.12

1.10

1.08

1.06

1.04

Austria_

Canada

New Zealand

The plot shows that the places with the highest health scores are regions in East Asia including Singapore,

ggplot(top regions, aes(x = reorder(country, -mean happiness), y = mean happiness, fill

data points for the four attributes: happiness score, GDP per capita, social support, and healthy life

clusters <- kmeans(happiness[, c("score", "income", "support", "health")], centers = 3)</pre>

labs(title = "Clustering of Health vs Happiness", x = "income", y = "score")

0.6

The plot shows that the places with the highest happiness score are regions in Europe including Finland,

income

Create a comparative analysis to find the top countries with the highest happiness score.

Denmark, Norway, etc. followed by other economically developed countries.

country_happiness <- happiness %>% group_by(country) %>% summarise(mean_happiness = mean(score)) top_regions <- country_happiness %>% arrange(desc(mean_happiness)) %>% head(10)ggplot(top_regions, aes(x = reorder(country, -mean_happiness), y = mean_happiness, fill= mean_happiness)) + geom_bar(stat = "identity") + labs(title = "Average Happiness By Country", x = "country", y = "mean score") + theme(axis.text.x = element_text(angle = 100, hjust = 1)) Average Happiness By Country 8 -6 mean_happiness mean score 7.7 7.6 7.5 7.4

Switzerland -

country

Create a comparative analysis to find the countries with the highest health score.

= mean happiness)) + geom bar(stat = "identity") + labs(title = "Average Happiness By Country", x = "country", y = "mean score") + theme(axis.text.x = element_text(angle = 100, hjust = 1)) Average Happiness By Country 0.9 mean happiness mean score

Hong Kong Switzerland -Spain_ Italy -Northern Cyprus country In conclusion, the result shows that the attribute income (GDP per capita) positively influenced the happiness

score in the country. This means that improving the GDP per capita for each country will also increase the

country's happiness index, and permit individuals to access more resources. Economic resources in the country

improve the quality of life for most people and influence other criteria including life expectancy, social support,

Conclusion