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# Abstract

# Introductaion

## Background Information

It is well known that sport is essential for individual’s physical health. With sport, we can train our muscles and respiratory, as well as enhancing immunity (Ornulf Seippel. 2006). In accordance with Buraimo, Jones and Millward (2011), approximately half of the people did not participate in any kinds of sports based on several reasons, including job character, change of entertainment way and long working period. In this way, the risk of obesity, depression, high blood pressures, as well as a series of cardiovascular diseases increased rapidly, being harmful to public health. For solving health issues, augmented reality (AR) videos games are in consideration. Augmented reality is an extended version of VR. It combines virtual elements with real world, through the assist of visual devices, including eyeglasses, monitor, as well as smart devices (Tim Fisher. 2021). Overlapping on and tracking in real world objects, AR objects seems occupying the same space. Besides visualization, AR system can also contain sound and tactile, for providing an new form of world. AR technology can be applied in various types of applications, such as map and games. In this study, we focus on Pokemon Go, a popular AR mobile games.

Pokemon Go is a famous mobile game developed by Niantic Inc, as well as released at 2016 on both Apple Store and Google Play (Luke Reilly. 2017). Pokemon Go players using GPS signals to locate, catch (also obtain ingredients for training Pokemon), hatch (players walk around 2 to 10 km to obtain a Pokemon) and train virtual creatures, Pokemon. Those Pokemon can be used for battling and Gym controlling (Andrew Webster. 2015). Pokemon Go use a map and camera to display the virtual spots, like Pokestop, Gym and activity location points. (Smith. 2017) Pokemon Go is a celebrated application, with 632 millions times of downloading and 147 millions of monthly active players. Due to the popularity, Pokemon Go plausibly facilitating an obvious behavior change in public health (Dillet. 2016).

According to multiple researches, Pokemon Go obviously and positively affect the amount of physical activities. The amount of physical activities increased approximately 25%, comparing with previous activity level. Pokemon Go, additionally, rising the physical activity level across gender, ages, as well as weight status. The physical activityk level of player, inactive originally, increase sharply in general (Gunther Eysenbach. 2016). Some studies, however, suggested that Pokemon Go cannot directly advance the public physical health. Despite the best effects in first period, players’ physical activity level drop sharply, meaning that the positive effects mentioned is not sustainable (Allana LeBlanc et al., 2016). This effect can be related to the motivation of players. There are three kinds of motivation for players, including health, social and immersion. Although players with health motivation presented the significant increase of the amount of physical activities, the effects caused by social and immersion motivation is limited. Merely the time spending outdoor increased rapidly (Lukas Dominik et al., 2017). In this situation, an analysis based on several aspects was conducted.

## Research Objectives

It is commonly know that Pokemon Go was not designed for public health improvement. That is the reason of the confusion about the relation between Pokemon Go and physical activity. For truth discovering, the entire study was conducted based on four aspects. The relation between frequency of app usage and amount of app usage, firstly, is a main focus. Theoretically, the more the app usage, the higher the opportunity for players doing physical activity (e.g. catching Pokemon, or turning Pokestop). Following the application usage aspect, we also discuss the problem in players’ characteristic. Since some researches proved that Pokemon Go players tend to join game-related physical activity, instead of physical activity in general (Alessandro Gabbiadini, 2017). The aim is discovering the existence of relations between Pokemon Go players and amount of general physical activity. In fact, the level of physical activity can be affected by various factors, including motivation, education level, and gender. We, thus, want to locate variables associating with the amount of physical activities. Last but not least, we want to examine the effects of the attitude towards physical activity caused by gender or educational level?

# Data Description and Processing

The data was obtained from a study, carried out following the code of ethics of the world medical association (Declaration of Helsinki) for studies using human as data. Amazon Mechanical Turk (MTurk), an internet-based platform offering online participant pool, was applied for data collection (Buhrmester, Kwang, & Gosling, 2011; Paolacci & Chandler, 2014). The original data contains 999 records, described by 31 variables. Prior to processing data, we examine the number of missing values (which is 0), as well as filtering out records in accordance with the variable “ATTENTION\_filter1”.

“ATTENTION\_filter1” is a variable for filtering out non-focus participants as Mechanical Turk experiment was applied. In spite of the convenience as well as limitless of time and location, Mechanical Turk experiment cannot guarantee that participants are paying attention as the survey was completed online (Jennifer Jacquet, 2011). Ensuring only data from focusing-on-survey participants were collected, the item symboled by “ATTENTION\_filter1” was used. If failed choosing “Disagree” in this question, the records will be removed due to being classified as non-focus records. The remaining records inside the dataset is 981. After primary data cleaning, we transform all columns into integer score, according to the scales above-listed. This subjective assigning method is plausible for applying interval scale and the concept of distance (Chaowei Yang, 2014). Both id, submitdate, ipaddr and ATTENTION\_filter are not in consideration due to irreverent of objectives. Thus, Twenty-eight variables, for further grouping and analysis, will be discussed below.

Despite Age and gender, discrete number and nominal data respectively, all the others are ordinal variables. Both “Frequency of App Usage” and “How often sharing on social media” are ordinal data, anchored with the scale from 1 = “never” to 7 = “very often”. Former accessing the extent of players using Pokemon Go per month, while the latter demonstrate the frequency players sharing their achievements on social media. Those variables mentioned above were treated as independent variables, while the remaining variables will be grouped by row mean, in accordance with the result of Cronbach’s alpha. Cronbach’s alpha, also known as alpha reliability, is a measure for assessing strength of internal consistency, of several items or variables. The alpha score was calculated by correlating the score for every items with the total score for related observations, following the comparison of the variance of individual item scores (Cronbach Lee, 1951). The formula is:

Where is number of scale items, is to the variance associated with item , and denoted as the variance associated with the observed total scores (Chelsea Goforth, 2015). In accordance with Rule of Thumb, if the alpha score is between 0.7 and 0.8, the grouping process is plausible (Stephanie Glen, 2021). Proving by alpha score, We group variables having strong internal consistency by mean of each instances. In supplement, group-by-mean method (Underhill L.G, 1998) was applied as we prefer grouping variables, without missing much information(like median)(Akhihesh Ganti, 2021) or altering the scale(Daniel McNeish & Melissa Gordon Wolf, 2020). The details were mentioned below.

There are 12 variables, grouped as , measuring participants’ attitude towards physical activities (). (Scale of all questions were from 1 = “completely disagree” to 7 = “completely agree”). Players’ physically behavior was assessed in two aspects, recency and frequency (). The first three items for measuring recency of participants’ physical activity were “When was the last time you had (1) a walk for more than 30 min/(2) had a run/(3) had a bike ride to get some exercise?”. (The scale for those questions is 1=“more than one month ago”, 2=“about four weeks ago”, 3=“about three weeks ago”, 4=“about two weeks ago”, 5=“about one week ago”, 6=“during the last week” and 7=“yesterday”.) For measuring frequency, the following three questions were adopted: “How many times have you had (1) a walk for more than 30 min/ (2) had a run/ (3) had a bike ride to get some exercise during the last month?, with the scale from 1=”never" to 7=“every day”. Both former and latter were transformed as one variable, . The remaining three variables, names starting with “PokemonPastBehaviour”, were used for assessing participants’ physical behavior relating to Pokemon Go (). Questions represented by these three variables were “How many times have you walked more than 30 min/ had a run/ had a bike ride with the intent of searching for Pokemon Go during the last month?” ( anchored with 1=“never”, 2=“two times”,3=“from three to five times”,4=“from six to eight times”, 5=“from nine to eleven times”, 6=“from twelve to fourteen times”and 7=“every day”). Those variables were grouped as . Back of grouping, the new data set contains eight variables and 981 records. Prior to model selection, we look at the summary:

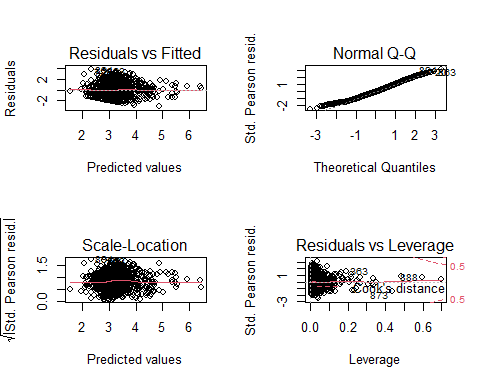
## age education Gender Attitude   
## Min. :18.00 Min. : 1.000 Min. :1.000 Min. :3.000   
## 1st Qu.:25.00 1st Qu.: 5.000 1st Qu.:1.000 1st Qu.:5.167   
## Median :30.00 Median : 8.000 Median :1.000 Median :5.417   
## Mean :32.55 Mean : 6.889 Mean :1.378 Mean :5.362   
## 3rd Qu.:37.00 3rd Qu.: 8.000 3rd Qu.:2.000 3rd Qu.:5.667   
## Max. :74.00 Max. :11.000 Max. :2.000 Max. :6.500   
## PhysicalActivity PokemonGo\_AppUsage social\_sharing PokemonGo\_Relate.Behaviour  
## Min. :1.000 Min. :1.00 Min. :1.000 Min. :1.000   
## 1st Qu.:2.333 1st Qu.:1.00 1st Qu.:1.000 1st Qu.:1.000   
## Median :3.000 Median :1.00 Median :1.000 Median :1.000   
## Mean :3.244 Mean :2.45 Mean :1.611 Mean :1.492   
## 3rd Qu.:4.167 3rd Qu.:4.00 3rd Qu.:1.000 3rd Qu.:1.667   
## Max. :7.000 Max. :7.00 Max. :7.000 Max. :7.000

According to the previous summary and plot 1, the response variable is continuous, and the data do not suggest evidence for applying a specific distribution. polynomial model, therefore, was applied. The details were discussed on the next session.

# Model Selection

Polynomial linear model is a model based on simple linear model.

For confirming the plausibility, four assumption plots were used:



After developing full linear model, stepwise selection, using AIC as criteria, was applied to select the best model. Stepwise regression is a step-by-step iterative and automatic model selection approach (Adam Hayes, 2021), based on backward regression and combining with forward. With this approach, we can re-examine the importance of variables, as well as correct the misleading caused by backward selection. For instance, it is plausible that a variable, removed in backward selection, is included in first stage of forward selection method. Applying stepwise selection method, we can include that variable again, for obtaining the best model (R. R. Hocking, 1967). McElreath (2016) revealed that Akaike Information Criteria (AIC) is a well-known information criteria, for evaluating the data-fitting performance of a model. The model with smallest AIC is the best as describing greatest amount of information with smallest amount amount of variables (Bevans, 2021). Prior to result observation, we were suggested checking the assumptions.

According to figure 2, there is no obvious positive or negative relations between residuals and fitted values. The final model is:

# Analysis Result

# Conclusion and Discussion

In accordance with the model assumptions plots, the Gamma model used can have improvement. In the future, we can try neural network model.

# Reference