### **A picture containing drawing Description automatically generated**

### **Data Science Fundamentals**

**(CS 890ES)**

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Executives User Guide

on

**"Games Description Analysis & Keyword Suggestions"**

**Submitted by:**

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**INTRODUCTION:**

The present-day technology evolution has made many changes and provided everything at fingertips. This evolution has helped in the extraction of new fundamentals and terminologies effecting and operating the world today in every possible way. Websites and applications are some of those terminologies that have evolved and encouraged us to use them and procreate advanced technologies to help ourselves out. The change in generations and the way of thought processing and imaginations of mankind allowed users to just focus on making everything available in their hands itself. Games and video games have been upgraded from television connected version to online and application-oriented mediums. We now have a game available on many platforms and mediums at users' convenience.

Digital games that elongate to Xbox versions, PlayStation accessibility, Online versions, Mobile application versions and many more means of accessibility have been designed and developed which has features that correspond to its growth and sustainability in the market. The gaming industry is a huge platform for the developers and Google Play store is the biggest marketplace for the games. Statistics show that there are 10 million people playing games every minute and new developers count increase at a rate of 1 million every year. There is a need for a survey or analysis needed by the self-employed game developers to design a game. As of now, there is no automated application or website that shows the updated information about these data insights for the developers. In this project, we would like to apply our focus on games available on the mobile medium and the appropriate game description written for it. As we can observe we have different games each categorized into different genres like adventure, action, racing, puzzle, etc.

Each game constitutes having a special description that describes the basic features and functionality of the game. This description is considered unstructured data because it cannot be easily computed to derive any insights out of it. In order to make it useful for the new game developers, this data can be used. Text extraction techniques are identifying the keywords using the NLP methodologies would help in the resultant outcome derivation. So, the prediction of these keywords is possible when there is a combination of these NLP techniques on the text with the machine learning algorithm. Data science life cycle is an approach to organize this data and produce the results out of this text content from different games. All the steps within this life cycle are explained and detailed with the self-created games dataset for this project.

**PROBLEM STATEMENT:**

In our project, the main aim is to analyze the description available for an individual game and explore the text in ways, that can help in giving the users some inputs on how description can be amplified. The problem we are trying to analyze focus on suggesting certain keywords or phrases that help elongate the description of a game in a more read full way. To provide certain inputs for the description to be attraction and help the users to have a better viewpoint on the game. So, extracting and suggesting important phrases that can be embedded in the game description to help the developers with an intellectual and sophisticated game intro is the main concern.

**SOLUTION OVERVIEW:**

In a detailed study, it is identified that the game description provides in the home page of a particular app in the play store plays a key role in its success and download rate of the game. To attain these success results, it is necessary to study the data and know about different text analysis and classification methods, as the problem statement itself contributes to the factor that this project aims at text data. So, in achieving the above task or operating the problem statement, it is very much an essential factor that we have access to the data the gives information about unique games and its game description as well. To support this, we created a dataset by web scrapping the play store website to identify unique games and contributing factors to achieve the task. And once the dataset is generated, we applied different text analysis and visualization techniques to clean the dataset and make it operative enough to feed it to the model. TF-IDF (Text Frequency-Inverse document frequency) vectorizer is used to divide and explore the text and obtain the word matrix for each game description. The obtained unique features were passed to the machine learning model, for it to get trained and form clusters with appropriate words that can be suggested to the user when provided a game description. Providing more useful and content related phrases that can improve the intro or description of a game making it reachable to most of the users is the main concern. And in the process of achieving it, many other operations have been performed on the data available which will be clearly discussed further sections.

**BUSINESS VALUE:**

There is always a growing streak in the rate of game developers in any platform. Stats show that every year there is a rise of these developers count by 5 million in all platform resources altogether. The gaming industry generated almost $135 billion in 2018. The prediction is that in 2019 it will increase to generate 152.1 billion. In the entire world, the major proportion of working people belong to the USA, which comprises 2457 companies that support up to 220thousand jobs in its country alone [8]. Also, it is expected that by 2020 the revenue generation from the gaming industry would be approximated to 160+ billions in USD. Which means the growth rate of jobs and self-employed developers would rise to around 25 million. As their growth rate is constantly increasing both in terms of investment and development, the demand for skilful developers with new ideologies are highly welcomed. This project helps those kind of developers in achieving their initial survey and analysis about the current trends through keywords extractions and suggestions. Clearly from these figures it can be stated that the graph will keep a constantly rising curve of revenue generations that are yielded at regular intervals.

**PROJECT BENEFITS:**

There would always be a puzzling situation a new game developer. Sometimes they would not be clear about the idea and type of features, functionalities expected in a game. In order to provide some research insight, this is the project that guides them what to do. Apart from just game developers, there are also ways for knowledge for the people who are interesting in knowing the facts about the google play store trends. As this is solution developed based on text extraction combined with machine learning model implementations, this project could be an example for data engineers who wants to resolve a problem of similar kind. This kind of problematic approach could be said as one of its kind and it can help the one’s looking to solve it. By rendering accurate results to the audience would increase billions of revenue each year to the gaming industry and can keep a constant streak of success for the vendors or the stakeholders. Out of all these benefits, the main objective of this project, i.e., to provide assistance for the amateur developers who are ready to design a game can be the most beneficial people.

**TOOLS:**

These are the tools utilized for the construction of this project:

Python

Google Colab

Google Drive

Google play scrapper, Beautiful Soup for scrapping

Numpy, pandas, string, re for computations

NLTK library for text extraction

SKLearn for vectorization and modelling

Matplot, Seaborn for visualization

**ROLE OF EXECUTIVES IN DATA ANALYTICS LIFE CYCLE:**

Data Analytics life cycle consists of six phases. Executives are the users that needs to use the end product. So executives are people who travels with the project team from the beginning till the end of a project development process. Executives are also the part of stakeholders group who provide the team with costs of development in each phase. These are the people responsible for giving business to achieve a product what they are looking for. As data analytics life cycle is an iterative process, all each phase of the development cycle, the executives are communicated about the changes, measures and implementations.

To be noted being on a same page as that of the expectations and development results, the project team would update the deliverables and the statuses of changes happening in the process. Likewise, in this project development, the instructor of the course was considered as our key executive and we constantly took updates for the project modifications. These periodical updates ensures that executives are receiving the product as expected. Thus, executives play a key role from the start of development till the final deliverable is hosted on to the web server.

**PROJECT DEVELOPMENT:**

**A close up of a map

Description automatically generated**

The development of this project began from scratch like creation of a data set by web scrapping. 15k records of games data was web scrapped altogether from the google play store in different batch files. Since this project is all about the text analysis and keyword suggestions, we considered only the game description feature of all the game records in order to build the model. Post this consideration, a whole bunch of ETL operations like data cleaning with stop words, emojis, punctuations and URLs removal are performed on the text description feature. For better text processing and filtering, operations like tokenization, stemming and lemmatizing were also applied it. Till here the data cleansing and preparation are done. Moving on to the model development, the text feature was vectorized where all the words are counted and made into columns having the inverse ratio of their counts. This vectorized data is presented to the clustering model where data is grouped based on the similarity of the vectorized data under the influence of other main features like games genre. This way the model gets trained and grant us with its predictions. A consistent iteration process of this model training and key words cleaning before modelling could yield better accuracies and predictions whenever complicated descriptions are provided to the user.

**EXECUTION REQUIREMENTS:**

This project is all about rendering a research analysis of games from the Google play store. Executives are the people who are going to use the software at the end.

In order to see the project results, a software set up has to arranged and to do this and a series of instructions were documented in the setup guide for this project. On a higher level of understanding, an executive needs the following. They are:

Project code files

Google account to access google drive

Google colab

Access permission to the dataset

Web link to connect to the application

If one has a google account and permission to the games dataset available in google drive, they can open the google colab in the browser to see the results. Here the project file i.e., keywords\_suggestions.py file is loaded, give a test description at the end of the file and results can be seen in the same file itself. In future this project can be hosted on to a web server to see in a better representation of the application.

**OUTPUT DETAILS:**

This is the image that shows the output that user wants to lookup to. Here is the detailed discussion on what the executive sees:

Now let’s see when a sample existing game description is give for testing, how this model provide its suggestions:

A screenshot of a social media post

Description automatically generated

This is the description of an existing game in play store

A screenshot of a social media post

Description automatically generated

Keyword suggestions from Model and TF-IDF vectorized data

A screenshot of a social media post

Description automatically generated

Keywords that are recommended by the TF-IDF vectorised object that is trained with game genre, game ratings score and the title

**CONCLUSION:**

Through a constantly iterating data science life cycle process, the unstructured data is represented to the users in a useful way. The problem of identifying trends in the gaming industry has been solved by achieving at the steps of the lifecycle using the dataset that we created. A new developer would get some insight from the suggestions provided by this structured model. These keyword recommendations are some of the first of its kind application developed on the play store data. This project can be taken to the next level by improving the model performance and design quality of the web application. Alongside, the same dataset can be expanded by adding more games from different platforms and help a larger crowd of the developers.

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**Git Hub Link to access the project files:**

<https://github.com/KeerthiMettu/Games-Analysis>

Access links for our 2 datasets from Google drive:

1. BatchFile1.csv: <https://drive.google.com/open?id=18cRi8kB54mGyih2ufc1pPr6hHvxPDbJT>
2. Gaming.csv:

<https://drive.google.com/open?id=1WeDY7kgeDolOQ0m7iLtiwLU-LhrliTHK>

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