TABLE OF CONTENTS

1.	Introduction4
	1.1.Purpose
	1.2.Objectives
	1.3.Scope
2.	Review Of Literature
3.	Problem Statement
4.	Proposed System8
	4.1. Features and Functionality
5.	Software Requirements
6.	Implementation
7.	Result
8.	Conclusion
9.	Future Scope
Re	ferences 18

Introduction

Today, gadgets have become an individual's essential requirement as a communication and entertainment device all over the world. The technological world has grown more competitive than ever, with many manufacturers striving to offer the greatest products for their clients. The never-ending expansion of the choices space has placed clients in a quandary. Customers were left perplexed by the seemingly limitless number of available alternatives when it came to selecting the best device for them. Customers are influenced by the following aspects when choosing an electronic device to use: brand, RAM, battery life, price, year of release, novel features, personal recommendation, and so on. Even though most electronic devices contain all of the fundamental functionality, manufacturers aim to differentiate their goods by adding some new features to the current features to set them apart from the competition. Manufacturers embraced the challenge and competitiveness of creating the ideal device for their target clients by including creative ideas and features to improve the user experience. The increasing number of brands and models has increased market competitiveness. As a result, consumers have a plethora of alternatives to choose from. Our Internet is flooded with a vast amount of information about electronic gadgets. Hence it is hard for users to get a personalized recommendation about gadgets which can fulfill the requirement for the user.

By analyzing the mentioned issue an AI based recommendation system will be the feasible solution for this problem. Recommendation system information filtering technique which will provide users the desired information about the electronic gadgets according to their search interest. Recommender Systems create suggestions; the user can accept or reject them based on their preferences, and they can also give implicit or explicit feedback, either immediately or later. Users' activities and input can be saved in the recommender database and utilized to generate new suggestions in subsequent user-system interactions. Because of the economic potential of these recommender systems, several of the largest e-commerce companies (such as Amazon.com and Snapdeal.com) and the online movie rental firm Netflix have made these algorithms a prominent element of their websites. High-quality tailored recommendations enrich the user experience. Web personalized recommendation systems have lately been used to give various forms of individualized information to their respective users. Recommender Systems have boosted the economies of several e-commerce companies such as FANNG, who have made these systems prominent features of their websites.

1.1 Purpose

The purpose is to implement this recommendation system to suggest the desired electronic gadget to the user according to their requirement. It will give sufficient assistance to the user to buy the best electronic gadget. Also it will help the business of electronic gadgets to grow their sphere of influence.

1.2 Objectives

- To collect a wide range of data and cleaned it using data pre-processing using pandas.
- To provide personalized recommendations based on recorded information on the users preferences using count vectorization and cosine similarity.
- To create user-friendly interface using html,css and connect the model with flask.

1.3 Scope

- Can be applied in e-commerce website which are concentrate on electronic products to enhance its business.
- Can recommend electronic gadgets like Camera, T.V, Laptop, Gaming Consoles, headphones, speakers, etc.

Literature Review

Sr.no	Paper Name	Author	Findings		
1	A Gadget Recommen dation System using Data Science	A Pushpalatha*, Harish Sanmugam J, Jeya Pradeepa, Madhu Bala S	They have developed an AI-based recommendation system tailored specifically for electrical devices. It offers a basic user interface with several functions. They gathered a diverse set of data and sanitized it with data pre-processing. Then, utilizing a wide variety of acquired data, we particularly customized the K-means algorithm to fit the user input.		
2	A Movie Recommen der System: MOVREC	Manoj Kumar, D.K.Yadav	The proposed system have developed Movie recommendation system using K means algorithm by sanitizing the content through count vectorization. It also calculates the cosine similarity metrics.		
3.	Online Book Recommen dation System	Nursultan Kurmashov, Konstantin Latuta , Konstantin Latuta	The recommender system focuses on simplicity and speed. The user makes a registration and is asked to select 10 favorite books from at least 3 categories (genres) Based on this information the system makes recommendations. A user can continue to rate the books, buy them and add them to read list and thus allow the quality of recommendations to improve.		

Problem Statement

In the era of digitization, gadgets have become an individual's fundamental necessity as a communication and entertainment instrument. The technology world has become more competitive than ever, with numerous firms vying to provide the best products to their customers. Clients are perplexed by the never-ending increase of the options space. When it came to choosing the right gadget for them, customers were bewildered by the seemingly infinite amount of accessible options. When selecting an electronic gadget to use, customers are affected by the following factors: brand, RAM, battery life, price, year of release, unique features, personal recommendation, and so on. Despite the fact that most electronic gadgets have all of the basic capabilities, manufacturers strive to differentiate their products by adding some new features to the existing features to set them apart from the competitors. Manufacturers accepted the challenge and competitiveness of developing the best gadget for their target consumers by including innovative ideas and features to enhance the user experience. The growing number of brands and models has heightened market competition. As a result, customers have a wealth of options from which to pick. Our Internet is filled with a wealth of information regarding electrical devices. As a result, it is difficult for users to receive individualised recommendations on products that can meet their needs.

Proposed System

The model will take product name as an argument and give the top five recommendation using content based filtering approach. Content-based filtering uses item features to recommend other items similar to what the user likes, based on their previous actions or explicit feedback. The model will interact with user interface made by html and css and connected with flask.

CONTENT-BASED FILTERING

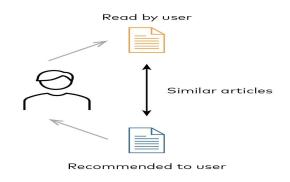


Fig 4.1 Content- based Filtering Approach

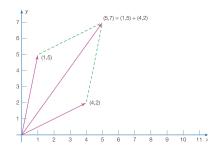
• Pre process the data:

- To fit into the model there is the need to convert the data into desired format
- The product_name,brand and model are converted into tags and merged into one column.
- Rest unnecessary attributes like price, ratings, user_Id are removed.

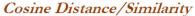
	Product_ld	Picture URL	Product Name	tags
0	0132793040	https://i.gadgets360cdn.com/products/cameras/l	Panasonic Lumix DC GH5GA DSLR Camera (20.3MP,	[Panasonic, Lumix, DC, GH5GA, DSLR, Camera, (2
1	0321732944	https://i.gadgets360cdn.com/products/cameras/l	Panasonic Lumix DMC GX8K DSLR Camera (20.3MP,	[Panasonic, Lumix, DMC, GX8K, DSLR, Camera, (2
2	0439886341	https://i.gadgets360cdn.com/products/cameras/l	Nikon D7200 DSLR Camera (24.2MP, Black)	[Nikon, D7200, DSLR, Camera, (24.2MP,, Black),
3	0439886341	https://i.gadgets360cdn.com/products/cameras/l	Panasonic Lumix DMC GH4A DSLR Camera (16MP, Bl	[Panasonic, Lumix, DMC, GH4A, DSLR, Camera, (1
4	0439886341	https://i.gadgets360cdn.com/products/cameras/l	Panasonic LUMIX G7 DSLR Camera (16.5MP, Black)	[Panasonic, LUMIX, G7, DSLR, Camera, (16.5MP,,

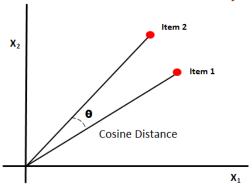
Count Vectorization:

- Count Vectorization is used to transform a given text into a vector on the basis of the frequency (count) of each word that occurs in the entire text.
- We are also using the concept of Count Vectorization to convert tags into vectors



- Calculating Distance metrics:
 - Cosine Similarity will be the best algorithm to calculate distance between vectors.
 - In data analysis, cosine similarity is a measure of similarity between two sequences of numbers.
 - For defining it, the sequences are viewed as vectors in an inner product space, and the cosine similarity is defined as the cosine of the angle between them, that is, the dot product of the vectors divided by the product of their lengths.
 - It follows that the cosine similarity does not depend on the magnitudes of the vectors, but only on their angle.
 - The distance will be calculated for all the vectors.





similarity(A,B) =
$$\frac{A \cdot B}{\|A\| \times \|B\|} = \frac{\sum_{i=1}^{n} A_i \times B_i}{\sqrt{\sum_{i=1}^{n} A_i^2} \times \sqrt{\sum_{i=1}^{n} B_i^2}}$$

- Defining recommendation function:
 - After converting to vectors and calculating the distance, a recommend function is defined.
 - The function will match product name with the indexes in dataframe and fetch the tags.
 - Than it will calculate similarity of those tags and recommend top five products whose tags are nearest to the original product tags.

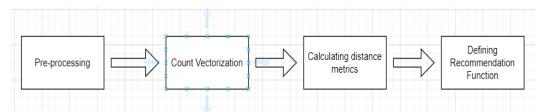


Fig 4.2: Methodology steps

4.1 Features & Functionality:

1. Search Bar:

- The search bar is given to mention the product name and it's key words to get recommendations related to it.
- The key words will find out nearest vectors to recommend products

2. Product Description:

- The product description will be visible when we click on the recommendations.
- Description will consist of product name, model , brand and price.

3. Other recommendation:

- The another set of products will also be recommended below the product description card.
- It can be customize into same brand or different brand product recommendation.

Software Requirements

• Front-end:

For front-end we are using react.js framework. React is an open source, JavaScript library for developing user interface (UI) in web application. It can be assisted by technologies such as Cascading Style Sheets (CSS) and scripting languages such as JavaScript. Web browsers receive HTML documents from a web server or from local storage and render the documents into multimedia web pages. CSS is designed to enable the separation of presentation and content, including layout, colors, and fonts. This separation can improve content accessibility; provide more flexibility and control in the specification of presentation characteristics; enable multiple web pages to share formatting by specifying the relevant CSS in a separate .css file, which reduces complexity and repetition in the structural content; and enable the .css file to be cached to improve the page load speed between the pages that share the file and its formatting.

• Back-end:

For back-end we are using python. Python is a high-level, general-purpose programming language. Its design philosophy emphasizes code readability with the use of significant indentation. It is also the high demand language for machine learning and data science based projects. The libraries like flask can be used to connect web pages with python based machine learning models. Flask is an API of Python that allows us to build up web-applications. It was developed by Armin Ronacher. Flask's framework is more explicit than Django's framework and is also easier to learn because it has less base code to implement a simple web-Application. Flask is based on the WSGI (Web Server Gateway Interface) toolkit and Jinja2 template engine.

Also to run model we are using jupyter notebook. The Jupyter Notebook is the original web application for creating and sharing computational documents. It offers a simple, streamlined, document-centric experience, program used to mix code, comments, and visualizations in an interactive document called notebook that can be shared, reused, and reworked in a web browser. Jupyter Notebook (formerly IPython Notebooks) is a web-based interactive computational environment for creating notebook documents. A Jupyter Notebook document is a browser-based REPL containing an ordered list of input/output cells which can contain code, text (using Markdown), mathematics, plots.

Implementation

EG df= EG df[['Product Id','Picture URL','Brand','Product Name','Model']] EG df.head() Product_ld Picture URL Brand **Product Name** Model Panasonic Lumix DC GH5GA DSLR Camera (20.3MP, Lumix DC GH5GA 20.33MP DSLR 0 0132793040 https://i.gadgets360cdn.com/products/cameras/l... Panasonic Panasonic Lumix DMC GX8K DSLR Camera (20.3MP. Lumix DMC GX8K 20.3MP DSLR 1 0321732944 https://i.gadgets360cdn.com/products/cameras/l... Panasonic D7200 24.2MP DSLR Camera 2 0439886341 https://i.gadgets360cdn.com/products/cameras/l... Nikon D7200 DSLR Camera (24.2MP, Black) Panasonic Lumix DMC GH4A DSLR Camera (16MP, Lumix DMC GH4A 16.05MP DSLR 3 0439886341 https://i.gadgets360cdn.com/products/cameras/l... Panasonic 4 0439886341 https://i.gadgets360cdn.com/products/cameras/l... Panasonic Panasonic LUMIX G7 DSLR Camera (16.5MP, Black) LUMIX G7 16.0MP DSLR Camera EG_df['Product Tag'] = EG_df['Product Name'].apply(lambda x:x.split()) EG_df['Model Tag'] = EG_df['Model'].apply(lambda x:x.split()) <ipython-input-23-bbc265e55992>:1: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-ve EG_df['Product Tag'] = EG_df['Product Name'].apply(lambda x:x.split())

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

EG_df['Model Tag'] = EG_df['Model'].apply(lambda x:x.split())

<ipython-input-23-bbc265e55992>:2: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.

EG_df.head()

_						
Product_Id	Picture URL	Brand	Product Name	Model	Product Tag	Model Tag
0 0132793040	https://i.gadgets360cdn.com/products/cameras/l	Panasonic	Panasonic Lumix DC GH5GA DSLR Camera (20.3MP,	Lumix DC GH5GA 20.33MP DSLR Camera	[Panasonic, Lumix, DC, GH5GA, DSLR, Camera, (2	[Lumix, DC, GH5GA, 20.33MP, DSLR, Camera]
1 0321732944	https://i.gadgets360cdn.com/products/cameras/l	Panasonic	Panasonic Lumix DMC GX8K DSLR Camera (20.3MP,	Lumix DMC GX8K 20.3MP DSLR Camera	[Panasonic, Lumix, DMC, GX8K, DSLR, Camera, (2	[Lumix, DMC, GX8K, 20.3MP, DSLR, Camera]
2 0439886341	https://i.gadgets360cdn.com/products/cameras/l	Nikon	Nikon D7200 DSLR Camera (24.2MP, Black)	D7200 24.2MP DSLR Camera	[Nikon, D7200, DSLR, Camera, (24.2MP,, Black)]	[D7200, 24.2MP, DSLR, Camera]
3 0439886341	https://i.gadgets360cdn.com/products/cameras/l	Panasonic	Panasonic Lumix DMC GH4A DSLR Camera (16MP, Bl	Lumix DMC GH4A 16.05MP DSLR Camera	[Panasonic, Lumix, DMC, GH4A, DSLR, Camera, (1	[Lumix, DMC, GH4A, 16.05MP, DSLR, Camera]
4 0439886341	https://i.gadgets360cdn.com/products/cameras/l	Panasonic	Panasonic LUMIX G7 DSLR Camera (16.5MP, Black)	LUMIX G7 16.0MP DSLR Camera	[Panasonic, LUMIX, G7, DSLR, Camera, (16.5MP,,	[LUMIX, G7, 16.0MP, DSLR, Camera]

```
: EG_df['tags'] = EG_df['Product Tag']+EG_df['Model Tag']
  EG_df= EG_df[['Product_Id','Picture URL','Product Name','tags']]
: EG_df.head()
       Product Id
                                                  Picture URL
                                                                 Panasonic Lumix DC GH5GA DSLR Camera (20.3MP,
                                                                                                                     [Panasonic, Lumix, DC, GH5GA, DSLR, Camera,
   0 0132793040 https://i.gadgets360cdn.com/products/cameras/l..
                                                                 Panasonic Lumix DMC GX8K DSLR Camera (20.3MP,
                                                                                                                     [Panasonic, Lumix, DMC, GX8K, DSLR, Camera,
   1 0321732944 https://i.gadgets360cdn.com/products/cameras/l...
   2 0439886341 https://i.gadgets360cdn.com/products/cameras/l..
                                                                         Nikon D7200 DSLR Camera (24.2MP, Black)
                                                                                                                    [Nikon, D7200, DSLR, Camera, (24.2MP,, Black),...
                                                                   Panasonic Lumix DMC GH4A DSLR Camera (16MP,
                                                                                                                     [Panasonic, Lumix, DMC, GH4A, DSLR, Camera,
   3 0439886341 https://i.gadgets360cdn.com/products/cameras/l..
   4 0439886341 https://i.gadgets360cdn.com/products/cameras/l...
                                                                  Panasonic LUMIX G7 DSLR Camera (16.5MP, Black) [Panasonic, LUMIX, G7, DSLR, Camera, (16.5MP, ....
```

```
EG_df['tags']=EG_df['tags'].apply(lambda x:" ".join(x))
```

```
EG_df['tags']=EG_df['tags'].apply(lambda x:x.lower())
```

```
from sklearn.feature_extraction.text import CountVectorizer
cv = CountVectorizer(max_features=5000,stop_words='english')
vectors = cv.fit_transform(EG_df['tags']).toarray()
```

from sklearn.metrics.pairwise import cosine_similarity
similarity=cosine_similarity(vectors)

```
def recommend(product):
    index = EG_df[EG_df['Product Name'] == product].index[0]
    distances = sorted(list(enumerate(similarity[index])),reverse=True,key = lambda x: x[1])
    for i in distances[1:6]:
        print(EG_df.iloc[i[0]])
```

Result

```
: recommend('Nikon D7200 DSLR Camera (24.2MP, Black)')
  Product_Id
                                                                0528881469
  Picture URL
                    https://i.gadgets360cdn.com/products/cameras/l...
  Product Name
                               Nikon D3400 DSLR Camera (24.2MP, Black)
                   nikon d3400 dslr camera (24.2mp, black) d3400 ...
  Name: 22, dtype: object
  Product_Id
Picture URL
                                                                0528881469
                    https://i.gadgets360cdn.com/products/cameras/l.
  Product Name
                               Nikon D5600 DSLR Camera (24.2MP, Black)
  tags
                    nikon d5600 dslr camera (24.2mp, black) d5600 ...
  Name: 33,
            dtype: object
  Product_Id
  Picture URL
                    https://i.gadgets360cdn.com/products/cameras/l..
  Product Name
                               Nikon D5500 DSLR Camera (24.2MP, Black)
                   nikon d5500 dslr camera (24.2mp, black) d5500 ...
  tags
  Name: 54, dtype: object
  Product_Id
Picture URL
                                                                0594481813
                    https://gadgets.ndtv.com/static/icons/cameras.
                   Nikon D3300 DSLR Camera (24.2MP, Black)
nikon d3300 dslr camera (24.2mp, black) d3300 ...
  Product Name
  tags
  Name: 113, dtype: object
  Product_Id
                                                               0594033896
                    https://gadgets.ndtv.com/static/icons/cameras....
Nikon D3500 DSLR Camera (24.78MP, Black)
  Picture URL
  Product Name
                   nikon d3500 dslr camera (24.78mp, black) d3500...
  tags
  Name: 53, dtype: object
```

Fig 7.1 Results in jupyter notebook

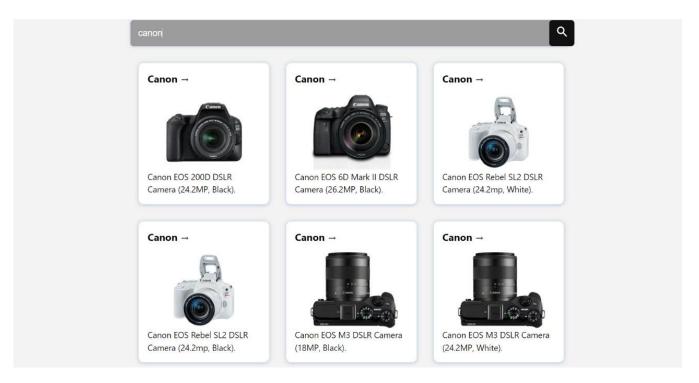


Fig 7.2 User Interface with results

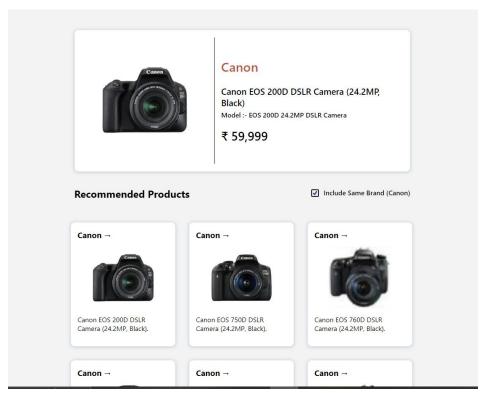


Fig 7.3 Product Description with same brand name recommendation

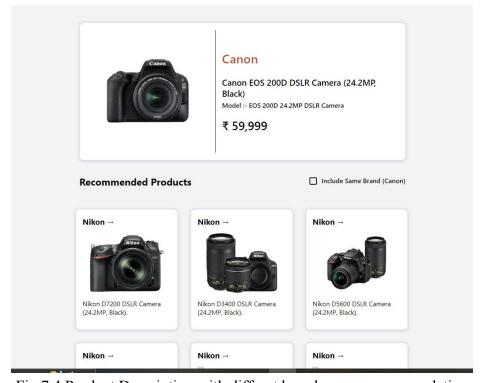


Fig 7.4 Product Description with diffrent brand name recommendation

Conclusion

Machine learning techniques are used to gather and retain data about each device and user in order to establish a link between those users and the gadgets. A user preference technique was used in this work to pick gadgets. Thus, in the realm of personalized recommendation systems, the critical issue of determining what device users require to enhance and satisfy their specific needs from in-depth personalization services has been resolved. The findings indicate that a gadget suggestion system may significantly improve service quality.

Electronic devices have become indispensable in meeting people's fundamental needs. Because of technological advancements, it is now important to meet various functional demands of end-users. As a result, it is critical to recommend devices to clients based on their unique preferences. With the fast advancement of technology, smart gadgets and communication networks have sprung up to cover every part of customer behavior. These data may also be taught and modeled for future usage in order to deal with potential technological advances.

Future Scope

- In the future the model can be updated into a deep learning project using CNN.
- Furthermore the website can convert to mobile application and can use camera settings for image based recommendation.
- The audio processing can also be used to fetch the key words from audio and match the recommendations in the model.

References

- [1] A. Pushpalatha, S. J. Harish, P. K. Jeya and S. Madhu Bala, "Gadget Recommendation System using Data Science," 2020 3rd International Conference on Intelligent Sustainable Systems (ICISS), 2020, pp. 1003-1005, doi: 10.1109/ICISS49785.2020.9315980.
- [2] Kumar, Manoj & Yadav, Dharmendra & Singh, Ankur & Kr, Vijay. (2015). A Movie Recommender System: MOVREC. International Journal of Computer Applications. 124. 7-11. 10.5120/ijca2015904111.
- [3] Kurmashov, Nursultan & Latuta, Konstantin & Nussipbekov, Abay. (2015). Online book recommendation system. 1-4. 10.1109/ICECCO.2015.7416895.