# **NLP Classification task for Shopping Products**

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**Abstract** 1

> available online. Product search and browsing is a 51 product categorizing available online. Product search and browsing is a 52 methods in natural language processing. shopping easier and consequently improving sales. The categorization of products must be 53 2 done properly to assist in product search. For effective organization of products on e-commerce 54 In order to train the data on the classifier, we will category. A product is typically represented by 56 word embeddings - where the embeddings the classification task. Using these features, we 65 models specified above. plan to evaluate different classification models, 66 Various combinations of features classification models will be explored and 67 3 compared to find a good product categorizing 68

#### 31 **1 Description and Motivation**

32 33 In this project, the problem we will be handling is 73 embedding. As we believe classification of 34 the multi-class classification for an E-commerce 74 products in an E-commerce domain is a highly 35 text dataset. This dataset contains 50k+ rows 75 paramount task in modern society, we wish to 36 wherein each row has category labels and 76 have a clear understanding of exactly which 37 descriptions. Overall, there are 4 different product 77 embeddings work best with which models and 38 type labels. We will investigate different 78 why. We believe this project would give future 39 embeddings and their influence on the outcome 79 developers and researchers further knowledge 40 and accuracy of classification models. We are also 80 which they can leverage if they were to build a 41 interested in finding ways to combine the 81 model that pertained to E-commerce. 42 embeddings for better performance. 43

44 Over recent years, we have witnessed a surge in 45 online shopping. On such E-commerce platforms, With the growth of online shopping since the 46 we have noticed that sometimes products may not pandemic, it is essential to keep the momentum 47 be shown to the user due to wrongful going in the e-commerce industry. One of the 48 categorization. This could cause issues for both main reasons people prefer e-commerce platforms 49 sellers and buyers as well. This motivated us to for market research and purchase is the ease of the 50 work on developing an efficient and accurate process and the variety of options that are 51 product categorizing model using advanced

### **Proposed Solution**

websites, each product is usually assigned a 55 convert the text into useful features using various several features like title, description, and image 57 capture the appropriate relationships required. and is identified by a category label. In this work, 58 We plan on working with embeddings given by we will focus on the prediction of product 59 BERT, GLUE, Glove, Word2Vec, TF-IDF, and categories based on their titles and textual 60 our custom-defined embeddings. With the help descriptions. Our main focus will be on the data processing and feature extraction process. We shall explore different kinds of feature extraction methods such as TF-IDF, and word embeddings

60 our custom-defined embeddings. With the help of these embeddings, we will categorize products different models such as BERT, SVM, and methods such as TF-IDF, and word embeddings

61 on received the embeddings of the embeddings and the different models such as BERT, SVM, and some process of the embeddings and the different models such as TF-IDF, and word embeddings. to extract the semantic information that will guide 64 approaches of the embeddings and the different

## How the proposed solution fits with previous research done

69 While we did go through many papers that 70 attempted to perform classification on E-71 commerce data, none of them went deeper into 72 their reasoning for choosing a specific

## 83 4 Blocks experienced or anticipate 84 experiencing.

- The data set contains product title and description, using both of them as input will be challenging as the product descriptions are lengthy (5-10 sentences per product).
- 90 2. We are uncertain of the baseline model91 to be used.
- 92 3. We are unsure if the size of the chosen
  93 dataset is sufficient to train, validate and
  94 test the model.
- 95 4. Defining the embeddings and using these
  96 kinds of embeddings to get the
  97 appropriate results will be a challenge.
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