

# AI Based : Events and Festival Planner

Mahesh N, Nireeksha A R, Mir Afnaan Ali, S Saila

1 Student, Department of Information Science and Engineering, T John Institute of Technology, Karnataka, India.

2 Student, Department of Information Science and Engineering, T John Institute of Technology, Karnataka, India.

3 Student, Department of Information Science and Engineering, T John Institute of Technology, Karnataka, India.

4 Assistant professor, Department of Information Science and Engineering, T John Institute of Technology, Karnataka, India.

## ABSTRACT

Now a days we are very much concerned about an event where many guests are invited. It is very difficult to know the likes and dislikes of all the guests and to satisfy everyone. The main aim of the proposed system is to develop an intelligent app that guides the user for the requirements of the upcoming events. To build the recommendation system that can access the user's messages, emails and calendar and process them using natural language processing NLP. The proposed android based application can track the user's movement, collects the liked places and activities and process them in the server using AI. To list out necessary goods or products required for a particular event. To trace the user's location and find out the required outlets nearby that will be helpful for the upcoming events. The proposed algorithms can predict the individual's likes and confirms the same by the user from the android app. The mobile app keeps track of all the invitations an individual receives and connections and based on that prepares the guest list for an events. Using machine learning and location tracing helps the user to get the list of the items required for the upcoming festival or events and where to get these items. Our application can suggest gifts of friends which they like.

**Keyword:-** NLP, Event Recommendation, Store Recommendation, Guest list preparation, Gift Recommendation, Menu Recommendation

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## 1. INTRODUCTION:

### 1.1 Natural Language Processing:

The essence of Natural Language Processing lies in making computers understand the natural language. That's not an easy task though. Computers can understand the structured form of data like spreadsheets and the tables in the database, but human languages, texts, and voices form an unstructured category of data, and it gets difficult for the computer to understand it, and there arises the need for Natural Language Processing.

There's a lot of natural language data out there in various forms and it would get very easy if computers can understand and process that data. We can train the models in accordance with expected output in different ways. Humans have been writing for thousands of years, there are a lot of literature pieces available, and it would be great if we make computers understand that. But the task is never going to be easy. There are various challenges floating

out there like understanding the correct meaning of the sentence, correct Named-Entity Recognition (NER), correct prediction of various parts of speech, coreference resolution(the most challenging thing in my opinion).

Computers can't truly understand the human language. If we feed enough data and train a model properly, it can distinguish and try categorizing various parts of speech(noun, verb, adjective, supporter, etc...) based on previously fed data and experiences. If it encounters a new word it tried making the nearest guess which can be embarrassingly wrong few times.

It's very difficult for a computer to extract the exact meaning from a sentence. For example – The boy radiated fire like vibes. The boy had a very motivating personality or he actually radiated fire? As you see over here, parsing English with a computer is going to be complicated.

There are various stages involved in training a model. Solving a complex problem in Machine Learning means building a pipeline. In simple terms, it means breaking a complex problem into a number of small problems, making models for each of them and then integrating these models. A similar thing is done in NLP. We can break down the process of understanding English for a model into a number of small pieces.

It would be really great if a computer could understand that San Pedro is an island in Belize district in Central America with a population of 16,444 and it is the second largest town in Belize. But to make the computer understand this, we need to teach computer very basic concepts of written language.

So let's start by creating an NLP pipeline.

## 2. STEPS TO CREATE NLP PIPELINE:

### Step 1: Sentence Segmentation

Breaking the piece of text in various sentences.

**Input:** San Pedro is a town on the southern part of the island of Ambergris Caye in the Belize District of the nation of Belize, in Central America. According to 2015 mid-year estimates, the town has a population of about 16, 444. It is the second-largest town in the Belize District and largest in the Belize Rural South constituency.

**Output:** San Pedro is a town on the southern part of the island of Ambergris Caye in the 2. Belize District of the nation of Belize, in Central America. According to 2015 mid-year estimates, the town has a population of about 16, 444. It is the second-largest town in the Belize District and largest in the Belize Rural South constituency.

For coding a sentence segmentation model, we can consider splitting a sentence when it encounters any punctuation mark. But modern NLP pipelines have techniques to split even if the document isn't formatted properly.

### Step 2: Word Tokenization

Breaking the sentence into individual words called as tokens. We can tokenize them whenever we encounter a space, we can train a model in that way. Even punctuations are considered as individual tokens as they have some meaning.

**Input:** San Pedro is a town on the southern part of the island of Ambergris Caye in the Belize District of the nation of Belize, in Central America. According to 2015 mid-year estimates, the town has a population of about 16, 444. It is the second-largest town in the Belize District and largest in the Belize Rural South constituency.

**Output:** 'San Pedro', 'is', 'a', 'town' and so on.

### Step 3: Predicting Parts of Speech for each token

Predicting whether the word is a noun, verb, adjective, adverb, pronoun, etc. This will help to understand what the sentence is talking about. This can be achieved by feeding the tokens ( and the words around it) to a pre-trained part-of-speech classification model. This model was fed a lot of English words with various parts of speech tagged to them so that it classifies the similar words it encounters in future in various parts of speech. Again, the models don't

really understand the ‘sense’ of the words, it just classifies them on the basis of its previous experience. It’s pure statistics. The process will look like this:

**Input:** Part of speech classification model

**Output:** Town - common noun    Is- verb    The – determiner    And similarly, it will classify various tokens.

#### **Step 4: Lemmatization**

Feeding the model with root word.

**For example** –There’s a Buffalo grazing in the field.

There are Buffaloes grazing in the field.

Here, both Buffalo and Buffaloes mean the same. But the computer can confuse it as two different terms as it does not know anything. So, we have to teach the computer that both terms mean the same. We must tell a computer that both sentences are talking about the same concept. So, we need to find out the most basic form or root form or lemma of the word and feed it to the model accordingly. In a similar fashion, we can use it for verbs too. ‘Play’ and ‘Playing’ should be considered as same.

#### **Step 5: Identifying stop words**

There are various words in the English language that are used very frequently like ‘a’, ‘and’, ‘the’ etc. These words make a lot of noise while doing statistical analysis. We can take these words out. Some NLP pipelines will categorize these words as stop words; they will be filtered out while doing some statistical analysis. They are needed to understand the dependency between various tokens to get the exact sense of the sentence. The list of stop words varies and depends on what kind of output are you expecting.

#### **Step 6: Dependency Parsing:**

This means finding out the relationship between the words in the sentence and how they are related to each other. We create a parse tree in dependency parsing, with root as the main verb in the sentence. If we talk about the first sentence in our example, then ‘is’ is the main verb and it will be the root of the parse tree. We can construct a parse tree of every sentence with one root word (main verb) associated with it. We can also identify the kind of relationship that exists between the two words. In our example, ‘San Pedro’ is the subject and ‘island’ is the attribute. Thus, the relationship between ‘San Pedro’ and ‘is’, and ‘island’ and ‘is’ can be established.

Just like we trained a Machine Learning model to identify various parts of speech, we can train a model to identify the dependency between words by feeding many words. It’s a complex task though. In 2016, Google released a new dependency parser Parsey McParseface which used a deep learning approach.

#### **Step 7: Finding Noun Phrases**

We can group the words that represent the same idea. For example – It is the second-largest town in the Belize District and largest in the Belize Rural South constituency. Here, tokens ‘second’, ‘largest’ and ‘town’ can be grouped together as they together represent the same thing ‘Belize’. We can use the output of dependency parsing to combine such words. Whether to do this step or not completely depends on the end goal, but it’s always quick to do this if we don’t want much information about which words are adjective, rather focus on other important details.

#### **Step 8. Named Entity Recognition (NER)**

San Pedro is a town on the southern part of the island of Ambergris Caye in the 2. Belize District of the nation of Belize, in Central America. Here, the NER maps the words with the real world places. The places that actually exist in the physical world. We can automatically extract the real world places present in the document using NLP.

If the above sentence is the input, NER will map it like this way:

San Pedro - Geographic Entity

Ambergris Caye - Geographic Entity

Belize - Geographic Entity

## Central America - Geographic Entity

NER systems look for how a word is placed in a sentence and make use of other statistical models to identify what kind of word actually it is. For example – ‘Washington’ can be a geographical location as well as the last name of any person. A good NER system can identify this.

### 3. EXISTING SYSTEM :

- In the existing system user prepares the guest list manually.
- User must think about the arrangement for any festival or occasion.
- For any events or festival, user need to sit and plan accordingly.
- The gift selection may not be liked by the friends or relatives.
- User has to save reminders for events on his calendar manually.
- User has to plan menu for guests manually.

#### 3.1 Disadvantages :

- Manual operation
- Time consuming
- User may miss out to invite important people
- Manual guest list preparation

### 4. PROPOSED SYSTEM :

- User can add the event or festival details in the app calendar.
- The proposed system can check the upcoming events or festival, and based on user’s GPS location, suggests the user for required goods or articles for the events or festival.
- Suggests the shops where it might be available.
- Suggests gifts for friends which he/she likes.

#### 4.1 Advantages :

- Auto advice for dishes for any occasion.
- No need to sit and plan everything for any events.

### 5. OBJECTIVES :

- To remind about the upcoming events.
- Guest list recommendation.
- Tension free event arrangement.
- Quality product recommendation required for any events.
- Based on user’s GPS Location, it gives the store recommendation.

## 6. METHODS FOR FEATURES :

App will have permissions to users' SMS, Email, Contact list, Call list, Location.

### **Event recommendation :**

App will analyse the users' text messages and if the text is related to an event then based on it it will set remainder for that particular event.

### **Invitation :**

If the messages sent by a person is related to inviting the other person, then the app will automatically add him in the guest list and invite that person for the event.

### **Guest list preparation :**

Guest list recommendation is based on the contact list which the app has access to. It is also based on the previous invitations for an event.

### **Gift recommendation :**

Gifts are recommended by analysing the likes of a particular person through the messages sent by the users.

### **Menu Recommendation :**

Based on input details entered by user during launch of the app and also based on dataset provided by developer , a menu is recommended in such a way that it benefits each and every person attending the event.

### **Store Recommendation :**

Based on location using GPS, nearby stores are recommended when traveling for a particular event.

## 7. IMPLEMENTATION:

### **7.1 KNN Algorithm (Introduction):**

K-nearest neighbors (KNN) algorithm is a type of supervised ML algorithm which can be used for both classification as well as regression predictive problems. However, it is mainly used for classification predictive problems in industry.

The following two properties would define KNN well:

**Lazy learning algorithm:** KNN is a lazy learning algorithm because it does not have a specialized training phase and uses all the data for training while classification.

**Non-parametric learning algorithm:** KNN is also a non-parametric learning algorithm because it doesn't assume anything about the underlying data.

**Working of KNN Algorithm (Process1):** K-nearest neighbors (KNN) algorithm uses 'feature similarity' to predict the values of new data points which further means that the new data point will be assigned a value based on how closely it matches the points in the training set. We can understand its working with the help of following steps:

#### **Step 1:**

For implementing any algorithm, we need dataset. So during the first step of KNN, we must load the training as well as test data.

#### **Step 2:**

Next, we need to choose the value of K i.e. the nearest data points. K can be any integer.

**Step 3:** For each point in the test data do the following –

- **3.1:** Calculate the distance between test data and each row of training data with the help of any of the method namely: Euclidean, Manhattan or Hamming distance. The most commonly used method to calculate distance is Euclidean.
- **3.2:** Now, based on the distance value, sort them in ascending order.
- **3.3:** Next, it will choose the top K rows from the sorted array.
- **3.4:** Now, it will assign a class to the test point based on most frequent class of these rows.

**Step 4:** End

#### **Working of KNN Algorithm (Process2):**

- a. Load the data
- b. Initialize K to your chosen number of neighbors
- c. For each example in the data
  1. Calculate the distance between the query example and the current example from the data.
  2. Add the distance and the index of the example to an ordered collection
- d. Sort the ordered collection of distances and indices from smallest to largest.
- e. Pick the first K entries from the sorted collection.
- f. Get the labels of the selected K entries
- g. If regression, return the mean of the K labels
- h. If classification, return the mode of the K labels

## **7.2 NLTK**

### **POS (Part-Of-Speech) Tagging & Chunking with NLTK POS**

#### **Tagging**

Parts of speech Tagging is responsible for reading the text in a language and assigning some specific token (Parts of Speech) to each word.

e.g.

Input: Everything to permit us.

Output: [('Everything', NN), ('to', TO), ('permit', VB), ('us', PRP)]

#### **Steps Involved:**

- Tokenize text (word\_tokenize)
- apply pos\_tag to above step that is nltk.pos\_tag(tokenize\_text)

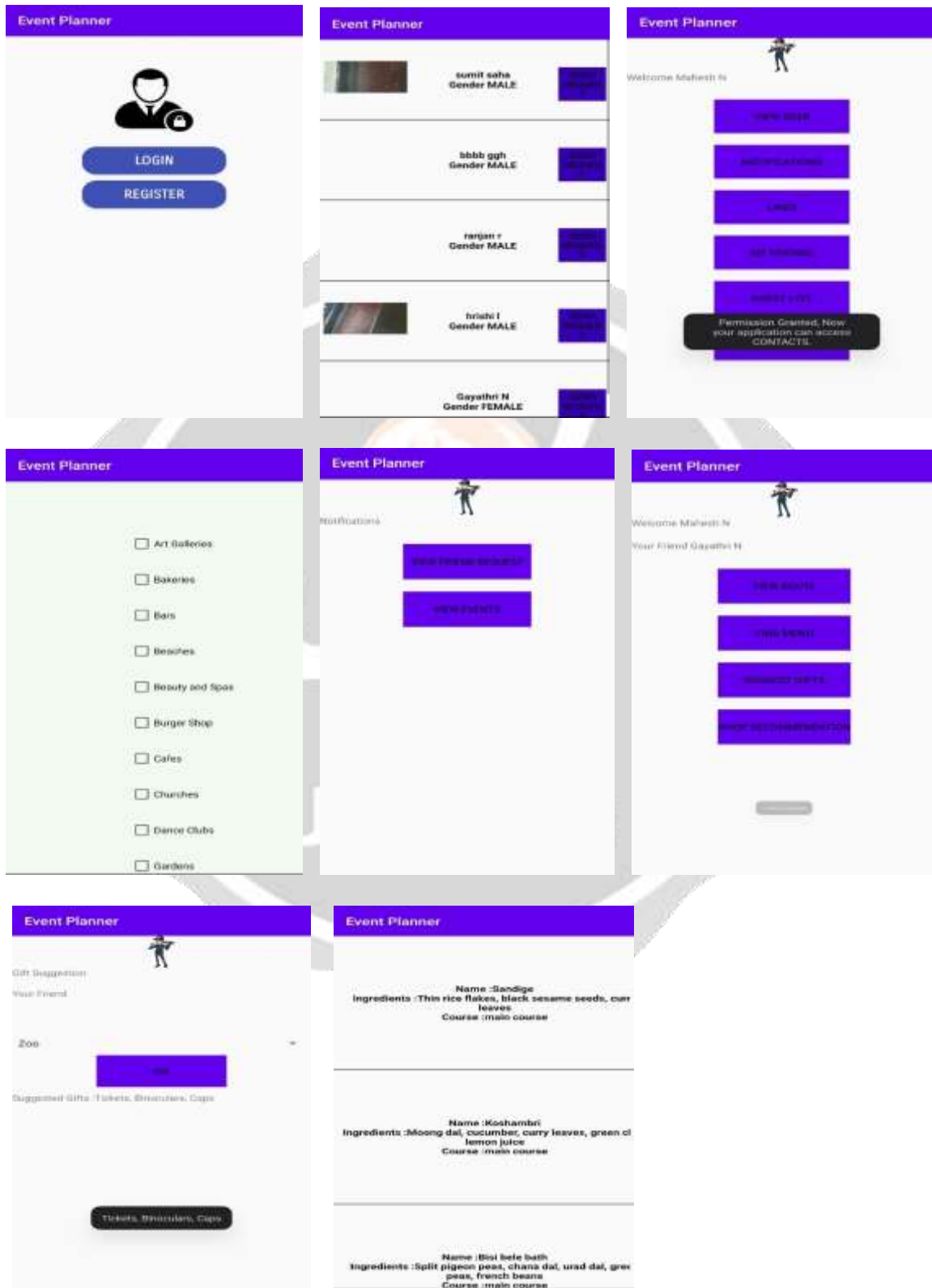
#### **Chunking:**

Chunking is used to add more structure to the sentence by following parts of speech (POS) tagging. It is also known as shallow parsing. The resulted group of words is called "chunks." In shallow parsing, there is maximum one level between roots and leaves while deep parsing comprises of more than one level. Shallow Parsing is also called light parsing or chunking. The primary usage of chunking is to make a group of "noun phrases." The parts of speech are combined with regular expressions.

#### **Rules for Chunking:**

There are no pre-defined rules, but you can combine them according to need and requirement. For example, you need to tag Noun, verb (past tense), adjective, and coordinating junction from the sentence.

### 8. RESULTS:



## 9. CONCLUSION:

We studied Natural Language Processing and how it can be used to teach computer basic concepts of written language. We also studied how NLP pipeline can be created using the seven steps under Natural Language Processing. Using this system, the user is able to plan his event tension free because the basic planning under all categories is done by the app. It makes users' work less. The basic Event recommendation, Invitation, Guest List preparation, Store recommendation, Gift recommendation and Menu recommendation is done by the system. This gives a good foundation for future Manual Event Management discoveries.

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