

NLP Analysis on Restaurant Reviews



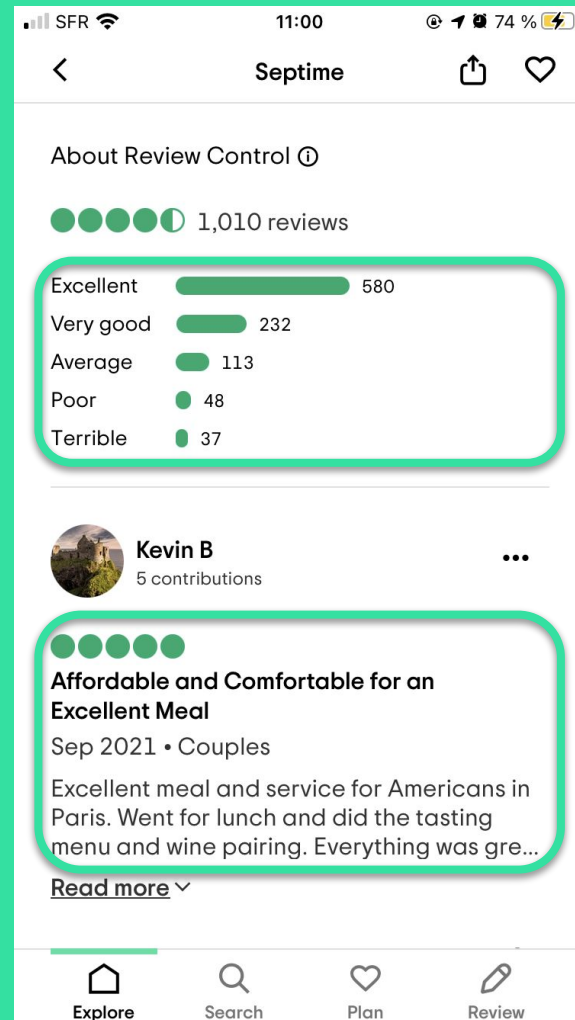
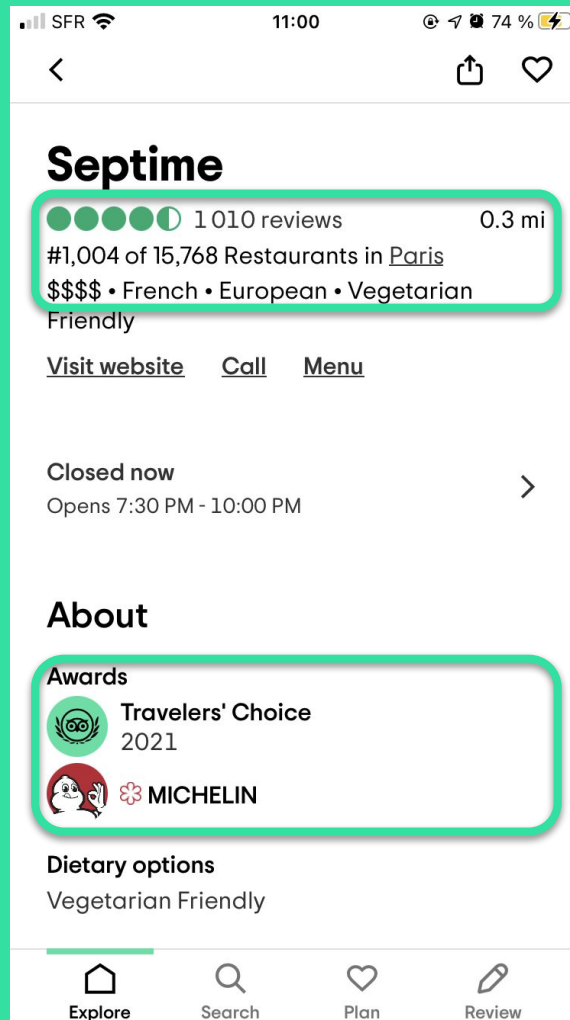
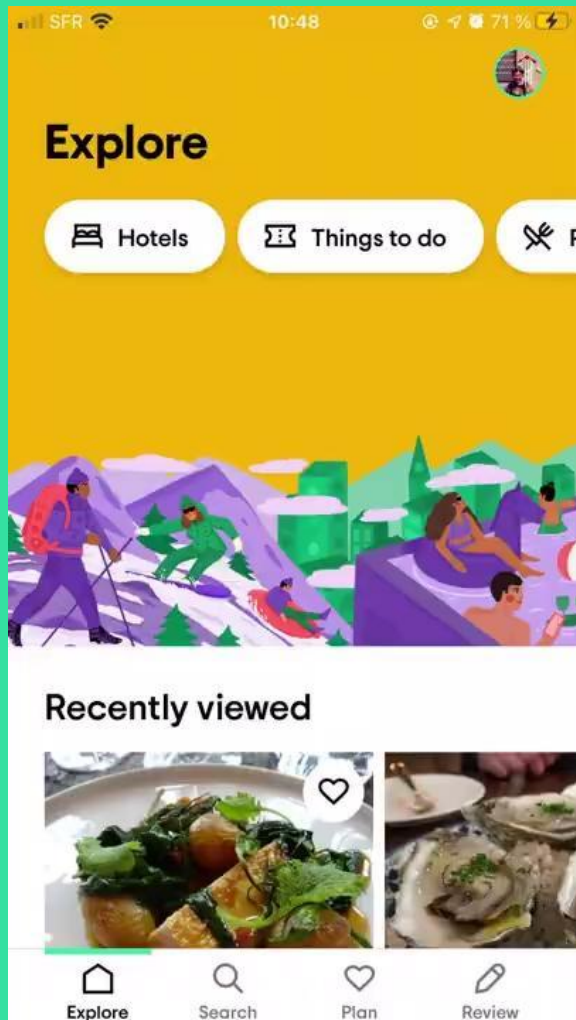
Tripadvisor[®]

NLP-filter

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IRONHACK
DAFT NOV21
Feb 11th, Paris



Quantitative scoring

Michelin-stars

Macarons, toques

Customer rating

Qualitative feedback

Food critics blogs

Word-of-mouth

Customer reviews

How do you
score
customer
reviews?



Data Collection

download Octoparse

All reviews for one restaurant:

- > 372 reviews

With 8 features:

- > review title

- > review text

- > review author

- > count of author's total reviews

- > date of visit

- > date of rating

- > likes

- > rating



Roadmap:

1. Clean, trim,
process text

Lemmatization

2. Compute
polarity score on
reviews

**Sentiment
Analysis**

3. Display words
that describe the
place best

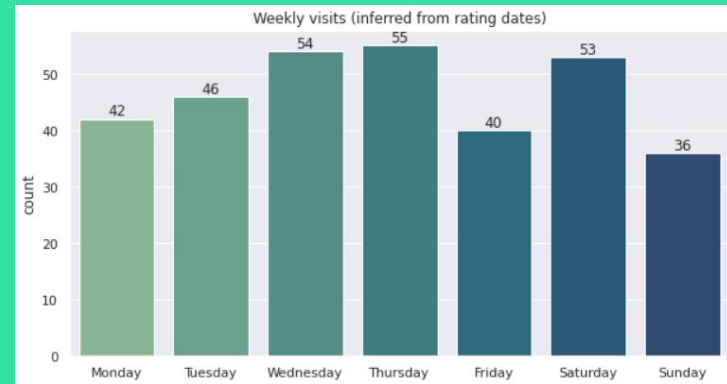
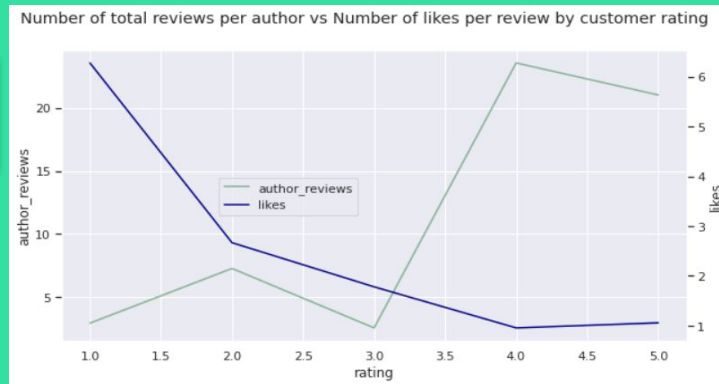
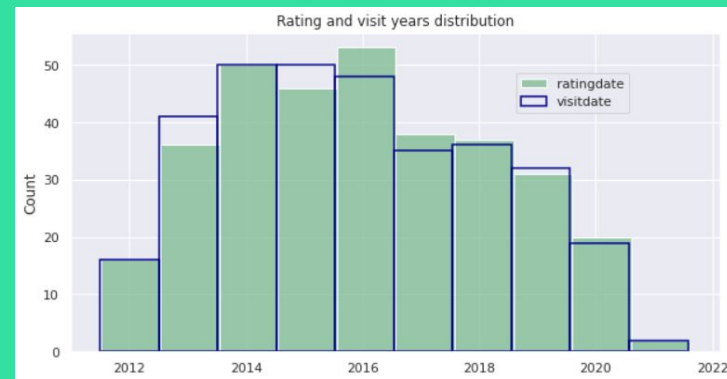
Word Cloud

EDA:

explore

the

data



1. Lemma tization

Expand contractions

Lowercase text

Clean digits and special
characters

Remove stop words

Lemmatize

```
import Spacy
```

> "dunno" → "do not know"

> "SepTime" → "septime"

> "hi-5!" → " "

> " " → " "

> "I really loved this
restaurant" → "really loved"

> "perfected tastings" →
"perfect tasting"

1. Lemma tization

Raw

'We had a wonderful experience. Great taste, great presentation. All the textures, flavors since the starter until the desert. Great chef',

Cleaned

'we had a wonderful experience great taste great presentation all the textures flavors since the starter until the desert great chef'

Lemmatized

'wonderful experience great taste great presentation texture flavor starter desert great chef',

```
import TextBlob  
.sentiment.polarity
```

2. Sentiment Analysis

Rating = 5/5

"excellent food,totally fresh,simple ingredients mixed to haute cuisine,great,caring service,interesting wine and beverage-list,one of the very best restaurants in paris,yes,it's worth the sometimes troublesome reservation!"

Polarity = 0.85

Rating = 1/5

'Very little plates, pretentious dishes, pretentious cook, everybody left the restaurant being hungry (if not for the bread they keep serving to fill us up).'

Polarity score = -0.262

2. Sentiment Analysis



> Polarity consistent with rating

> Outlier identified and dropped

> Polarized extremes

> Ratings 2 & 3 share

similar polarities



3. Word Cloud

```
import CountVectorizer
```

20 most occurring words and their frequencies

food	271	not	139
good	253	table	127
wine	209	great	121
menu	188	time	109
course	178	meal	109
reservation	160	staff	103
dish	157	place	98
paris	149	experience	92
service	149	like	84
lunch	147	taste	81

3. Word Cloud

Highest rating scores



Highest
polarity scores



3. Word Cloud

Lowest rating scores



Lowest
polarity scores



With this additional polarity-based wordcloud feature, Tripadvisor would be able to provide:

1. **Users** with a grouped, concise and customized overview of the restaurants's reviews

2. **Restaurants** with a visual tool to take corrective actions or to target their communication



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