



Avis Restau

Étude préliminaire des fonctionnalités pour “Détecter les sujets d’insatisfaction” et “Labelliser automatiquement les photos postées”

Projet

Détecter les sujets d'insatisfaction présents dans les commentaires postés sur la plateforme.

Labelliser automatiquement les photos postées sur la plateforme. Par exemple, identifier les photos relatives à la nourriture, au décor dans le restaurant ou à l'extérieur du restaurant.

```
graph TD; A[Détecter les sujets d'insatisfaction présents dans les commentaires postés sur la plateforme.] --> C[Étude préliminaire des fonctionnalités "Détecter les sujets d'insatisfaction"]; B[Labelliser automatiquement les photos postées sur la plateforme. Par exemple, identifier les photos relatives à la nourriture, au décor dans le restaurant ou à l'extérieur du restaurant.] --> C; C --> D[S'assurer de la possibilité de collecter de nouvelles données];
```

Étude préliminaire des fonctionnalités
"Détecter les sujets d'insatisfaction"

S'assurer de la possibilité de collecter de nouvelles données

Étapes

Analyser les commentaires pour détecter les différents sujets d'insatisfaction



Prétraiter des données textuelles



Utiliser les techniques de réduction de dimension



Visualiser des données de grandes dimensions

Analyser les photos pour déterminer les catégories des photos



Prétraiter des images



Utiliser les techniques de réduction de dimension



Visualiser des données de grandes dimensions

Collecter un échantillon (environ 200 restaurants) de données via l'API Yelp



Récupérer uniquement les champs nécessaires



Stocker les résultats dans un fichier exploitable (par exemple csv)

Description des données

2 jeux de données au format JSON

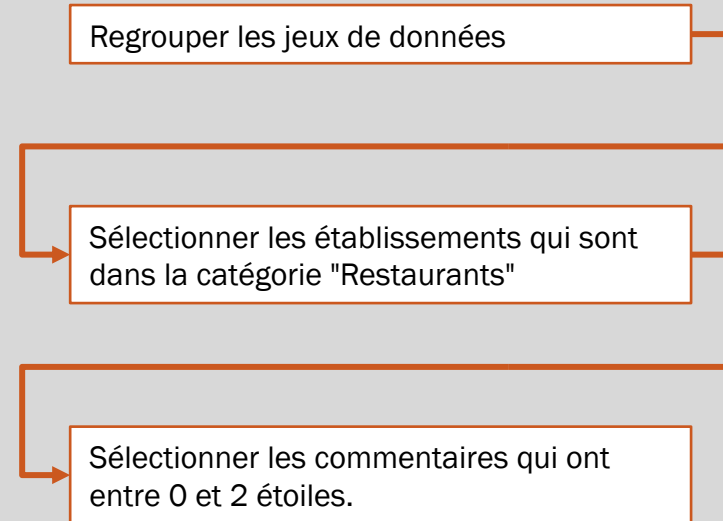
yelp_academic_dataset_business.json

```
{
  "business_id": "Xb0VJtN8ldxBsrgmEinHIA",
  "name": "Taco Bell",
  "address": "26 E Schrock Rd",
  "city": "Westerville",
  "state": "OH",
  "postal_code": "43081",
  "latitude": 40.1100433197,
  "longitude": -82.924741194,
  "stars": 2.5,
  "review_count": 24,
  "is_open": 1,
  "attributes": {
    ...
  },
  "categories": "Breakfast & Brunch, Restaurants, Tacos, Mexican, Fast Food",
  "hours": {
    ...
  }
}
```

yelp_academic_dataset_review.json

```
{
  "review_id": "m68FT2nDHXAyYZtBtE0n2Q",
  "user_id": "kTnqh6o9zPeiks3JVv-DyQ",
  "business_id": "Xb0VJtN8ldxBsrgmEinHIA",
  "stars": 2.0,
  "useful": 2,
  "funny": 0,
  "cool": 0,
  "text": "I will say Westerville needed Taco Bell back in town. The food is really good but I must say this is the slowest location I have ever been to. Takes approximately 10 minutes to get your food every time and this is supposed to be fast food. If it was my choice, I would not come here but my boys like it. They need to step up their game.",
  "date": "2016-08-14 21:48:02"
}
```

Prétraitement des données



Traitement des données texte

Very nice décor, with band playing jazz, but don't go with a group larger than 4 after 6:30PM

Mise en minuscule du texte

very nice décor, with band playing jazz,
but don't go with a group larger than 4
after 6:30PM

Suppression des accents

very nice decor, with band playing jazz,
but don't go with a group larger than 4
after 6:30PM

Expansion des contractions

very nice decor, with band playing jazz,
but do not go with a group larger than 4
after 6:30PM

Suppression de la ponctuation

very nice decor, with band playing jazz
but do not go with a group larger than 4
after 6:30PM

Suppression des mots avec des chiffres

very nice decor with band playing jazz
but do not go with a group larger than
after

Stemmisation

not open at very nice decor with band
play jazz but do not go with a group
large than after

Stop words

very nice decor band play jazz not group
large

very nice decor band play jazz not group large

Répartition des mots



Utilisation de 60 000 commentaires
sur 1 126 980

36392 mots uniques

Latent Dirichlet Allocation (LDA)

11 catégories réparties en 5 groupes



Service

- Call, manag, custom, would, told
- Veri, restaur, dish, servic, lobster



Nourriture

- Taco, steak, food, chip, good
- Food, chicken, tast, like, ric
- Sushi, coffe, roll, ice, tuna



Commande

- pizza, salad order, good like
- Burger, fri , order, chees, sandwich



Emplacement

- Order, go, food, place, time
- Food, place, like, good, one



Attente

- Order, food, us, minut, wait
- Tabl, us, wait, bar, seat

11 catégories avec un score
de cohérence de 0,40

Latent Dirichlet Allocation (LDA)

Exemple

Ordered take out that was to be 30 mins. 1 hour later we finally received our order. Pirate Pak burger dry, cold and unfit for consumption. Fish and chips also cold but did eat the fish which was at least fresh. Milk shake melted. I suspect that the order was ready at the 30 minute mark but somehow they either had a mix up or put the order somewhere where it was missed. Will give them the benefit of the doubt on temperature. No excuses on the Pirate Pak. If you must use this White Spot, check your take out order before you leave the restaurant for quality and temperature. I will not return.



Attente

- Order, food, us, minut, wait

K-means

11 catégories réparties en 5 groupes



Service

- food, place, good, servic, like
- bar, bartend, drink, beer, servic



Nourriture

- bbq, rib, brisket, pork, sauc



Commande

- burger, fri, order, food, good
- sandwich, order, bread, chees, good
- chicken, rice, order, fri, food
- taco, order, food, bell, mexican



Emplacement

- coffe, cup, place, breakfast, shop
- pizza, order, crust, chees, place



Attente

- us, tabl, food, wait, minut
- order, wait, call, time, minu

11 catégories avec un score de silhouette de 0,0074

K-means

Exemple

Ordered take out that was to be 30 mins. I hour later we finally received our order. Pirate Pak burger dry, cold and unfit for consumption. Fish and chips also cold but did eat the fish which was at least fresh. Milk shake melted. I suspect that the order was ready at the 30 minute mark but somehow they either had a mix up or put the order somewhere where it was missed. Will give them the benefit of the doubt on temperature. No excuses on the Pirate Pak. If you must use this White Spot, check your take out order before you leave the restaurant for quality and temperature. I will not return.



Attente

- order, wait, call, time, minu

Non Negative Matrix Factorization (NMF)

8 catégories réparties en 4 groupes



Service

- food, servic, good, terribl, horribl



Nourriture

- chicken, tast, veri, sauc, good,
- pizza, crust, chees, slice, top
- burger, fri, bun, onion, chees



Emplacement

- place, go, time, like, get
- us, tabl, ask, drink, server



Attente

- wait, minut, hour, min, long
- order, call, deliveri, time, said

8 catégories avec un score de silhouette de 0,0042

Non Negative Matrix Factorization (NMF)

Exemple

Ordered take out that was to be 30 mins. I hour later we finally received our order. Pirate Pak burger dry, cold and unfit for consumption. Fish and chips also cold but did eat the fish which was at least fresh. Milk shake melted. I suspect that the order was ready at the 30 minute mark but somehow they either had a mix up or put the order somewhere where it was missed. Will give them the benefit of the doubt on temperature. No excuses on the Pirate Pak. If you must use this White Spot, check your take out order before you leave the restaurant for quality and temperature. I will not return.



Attente

- order, call, deliveri, time, said

API Yelp

API GraphQL

Requête

```
{
  search(
    term: "restaurants",
    location: "$location",
    categories: "restaurants",
    limit: $limit,
    offset: $offset
  ){
    total,
    business {
      name
      id
      reviews(){
        text
        rating
      }
    }
  }
}
```

Limité à un extrait de 3 commentaires par établissement

Format JSON

```
{
  "name": "Fuckoffee",
  "id": "-22PaX8rQyh0ZQrdCiDz4A",
  "reviews": [
    {
      "text": "They make perfect flat white, I love the fruity taste. Nice and comfortable place, friendly people. \nI miss the crowd in the shop before this COVID-19...",
      "rating": 5
    },
    {
      "text": "Lol more like fuck racist establishments. This business has an a-frame sign that they're proudly boasting, which contains a very explicit racist message...",
      "rating": 1
    },
    {
      "text": "Do not support racist businesses. I don't care how good your coffee is, your ignorance has no place here.",
      "rating": 1
    }
  ]
},
```

Format CSV

"They make perfect flat white, I love the fruity taste. Nice and comfortable place, friendly people. \nI miss the crowd in the shop before this COVID-19...", 5

"Lol more like fuck racist establishments. This business has an a-frame sign that they're proudly boasting, which contains a very explicit racist message...", 1

"Do not support racist businesses. I don't care how good your coffee is, your ignorance has no place here.", 1

Les données sont prêtes à être traitées avec le traitement standard

Traitement des images

Objectif : Labelliser automatiquement les photos postées sur la plateforme.

200 000 images

5 catégories

- Interior
- Food
- Outside
- Menu
- drink

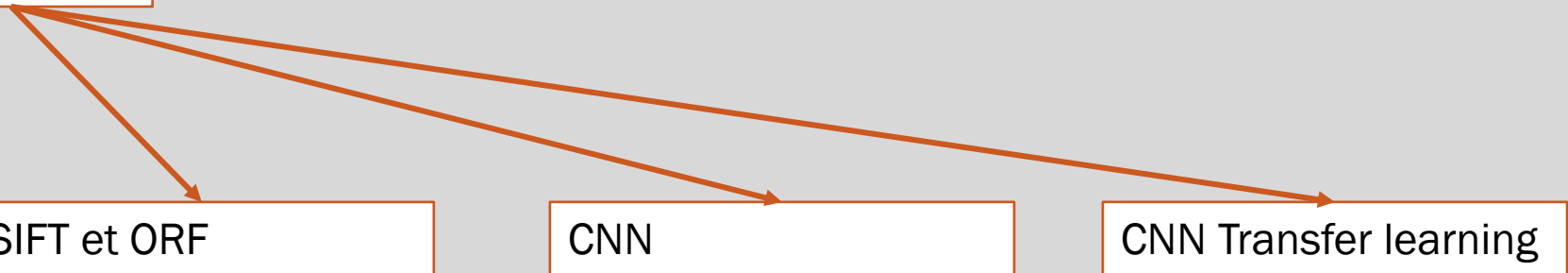
```
{  
  "photo_id": "Un_Og6jfhazVn7CxszkKEw",  
  "business_id": "RlsIqGfKpF5f3HV3vyNsbg",  
  "caption": "Drinks! Front is beer obviously. The pink one is the blackberry lemonade and the yellow one is the spicy margarita they made",  
  "label": "drink"  
}
```

Algorithmes

SIFT et ORF

CNN

CNN Transfer learning



SIFT et ORB

Traitement de l'image

Original



Nuances de gris



Mise à l'échelle

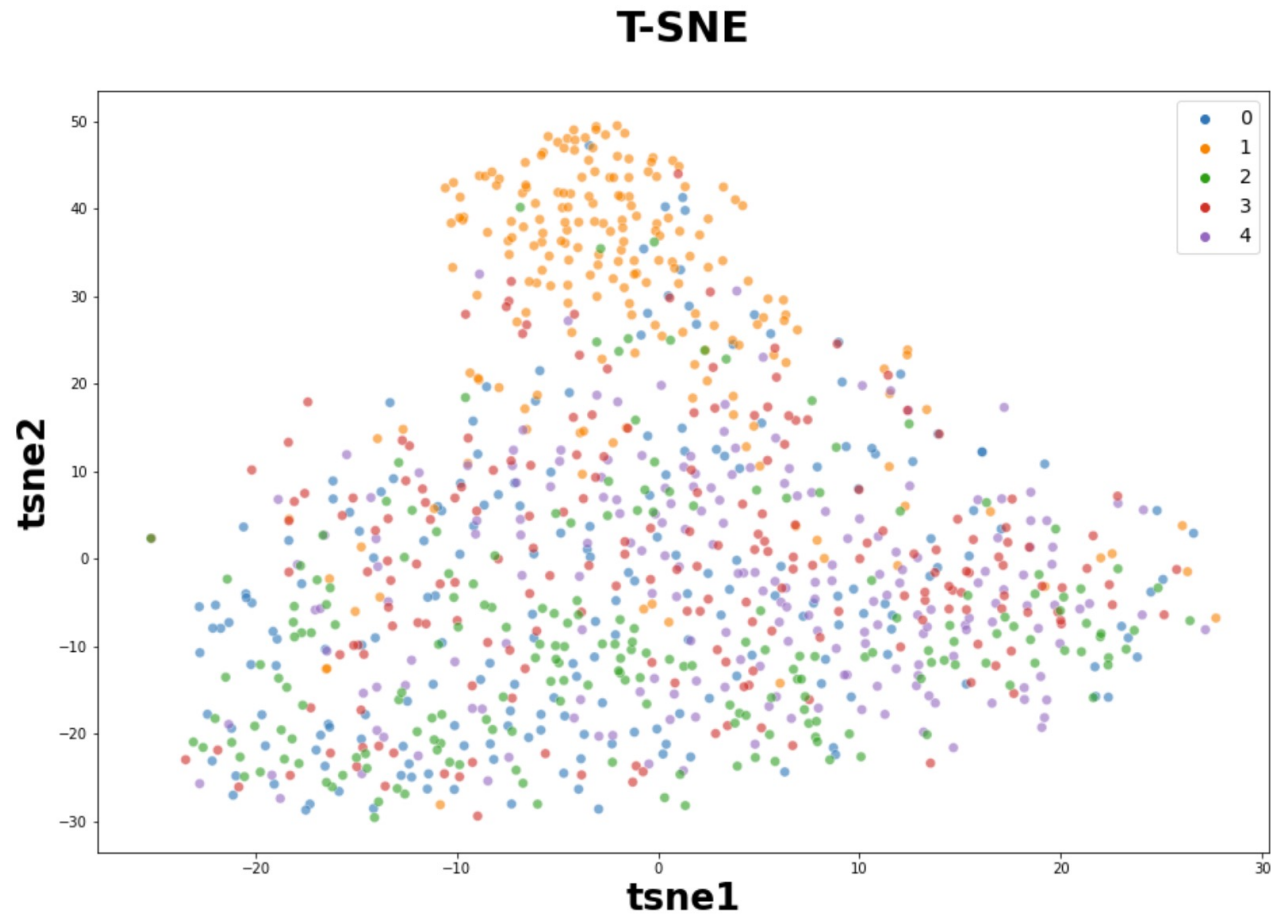


Histogramme

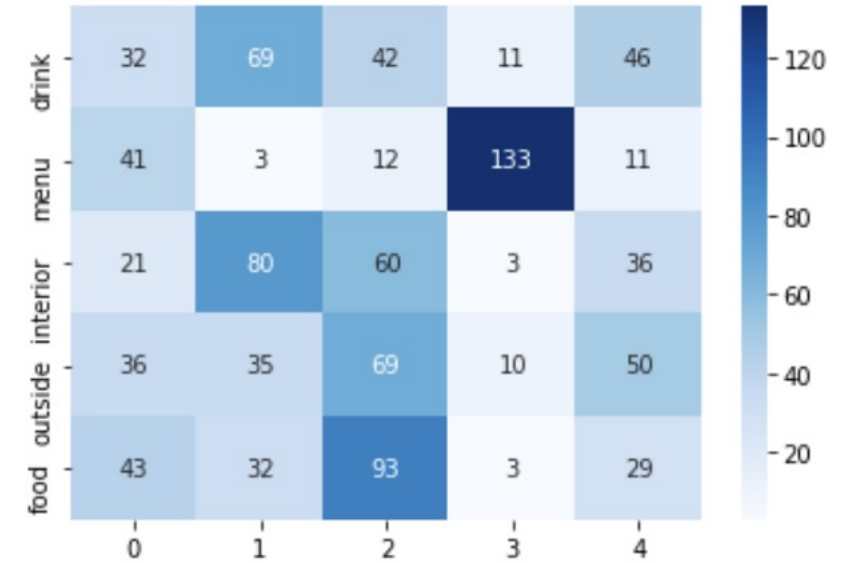


SIFT

Visualisation des catégories

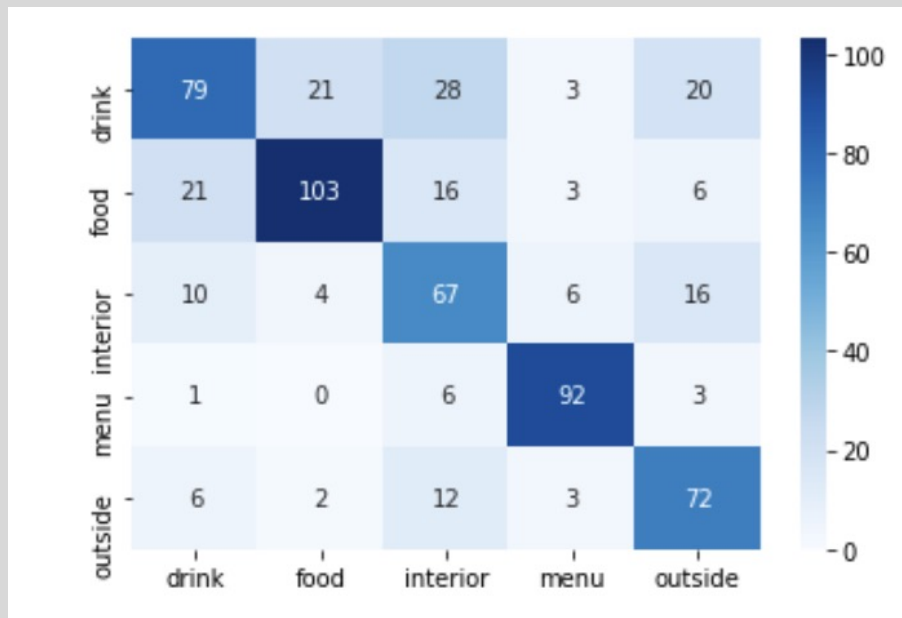
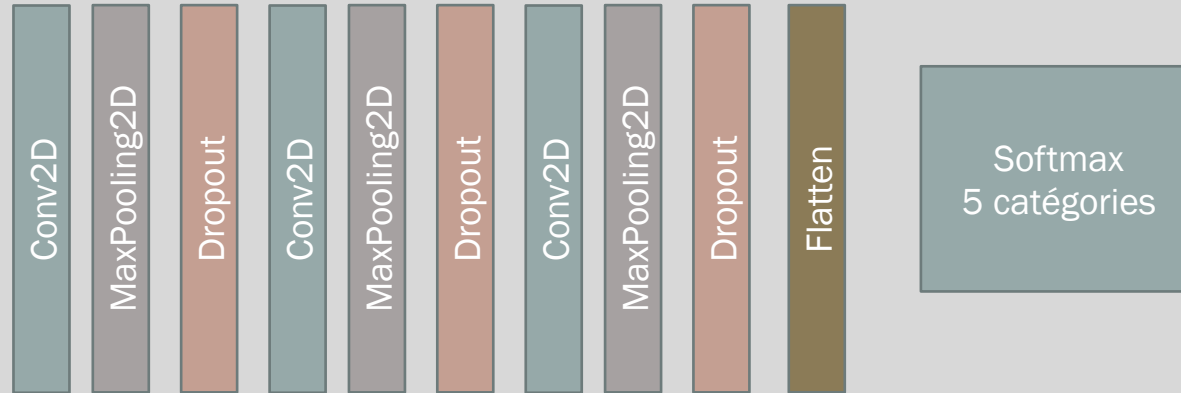


Réduction de dimension PCA, puis T-SNE



Réseau de neurones convolutifs (CNN)

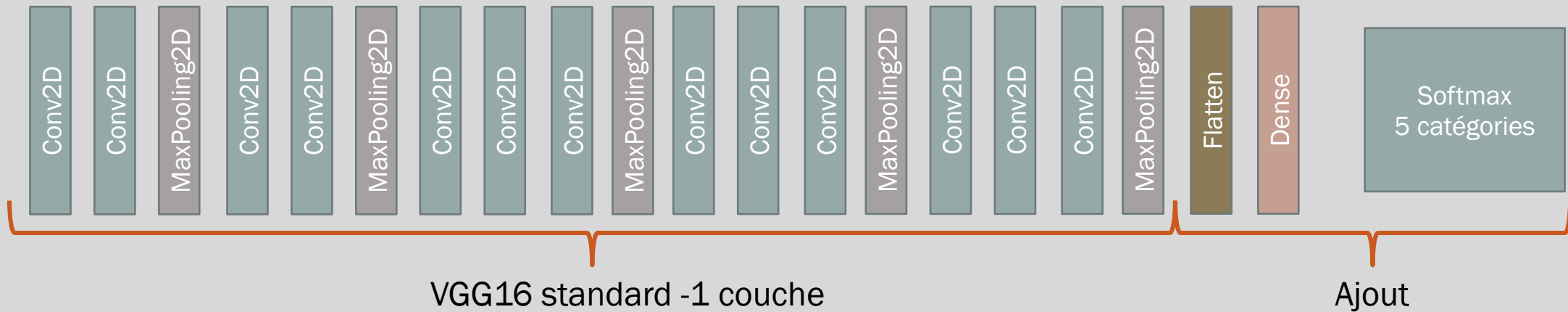
Précision de 65%



Intérieur détecté

Apprentissage par transfert (VGG16)

Précision de 86%



Intérieur détecté

Conclusion (texte)

Faisabilité de détection des sujets d'insatisfaction en utilisant **LDA**, mais avec des aménagements

- Détection de la langue

Avantage

- Rapidité d'exécution des prédictions

Inconvénients

- Difficulté d'estimer les catégories

Conclusion (images)

Faisabilité de classification d'images
avec un modèle de Deep Learning CNN
avec Transfer Learning

Avantage

- Bonnes prédictions à 86%
- Rapidité d'exécution des prédictions

Inconvénients

- Nécessite plus de ressources machine
- Utilise des images labellisées pour l'entraînement