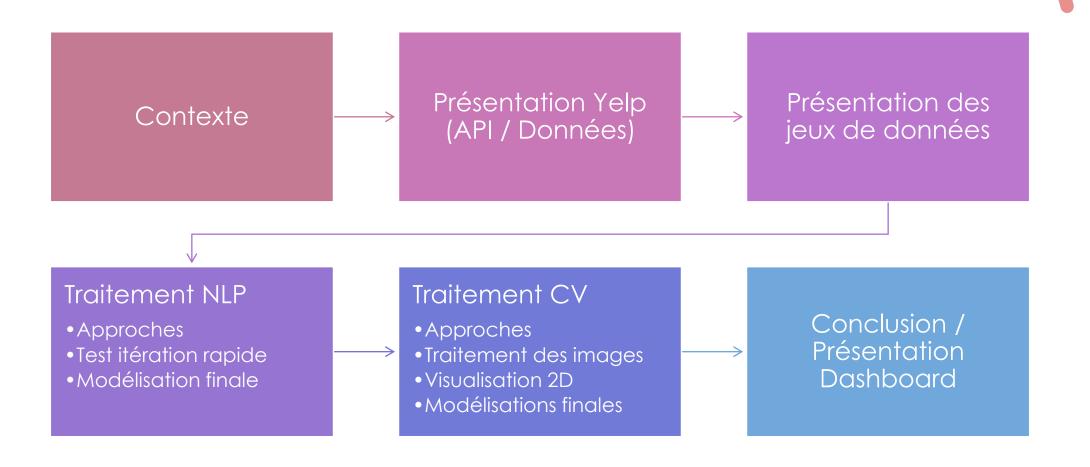
Avis Restau

Amélioration des produits IA (CV / NLP)



Sommaire





Contexte



Contexte







Nouvelle fonctionnalité de collaboration

Sujet d'insatisfaction

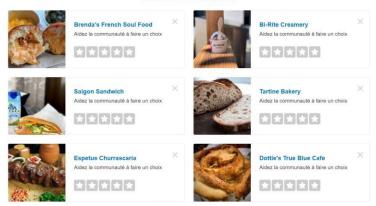
Labélisation automatique des images







Your Next Review Awaits







API Yelp



The first time dining with Eggmann's at their original location I ordered the BLT K's way with avocado and an egg. The BLT was good. Waffle fries were solid. I ordered the mimosa flight, which was a really fun addition to my brunch. My boyfriend got the California burrito with black beans. The burrito definitely needed the addition of hot sauce. The coffee was delicious. Overall it was a really good experience.

Recently, we dined at the new Wyandot golf course location. There are no longer cocktails offered on the menu, but I was told I could go to the golf course's pro shop to order a cocktail and bring it back to my table. The lady working the pro shop was super friendly and fun. She made me a vodka, orange juice, and pineapple juice concoction for only \$5. To eat, I ordered the chicken salad croissant with waffle fries and my boyfriend got the chicken and waffle. The chicken salad is tasty and the croissant is nicely toasted. I forgot to add avocado to the order, which definitely would have been nice! The waffle fries are good. My boyfriend's chicken and waffle was a large portion. The sriracha flavor really came through in the waffle. The mashed potatoes were alright. The addition of hot sauce elevated the meal.

The biggest difference to me is the atmosphere and energy of the new location. It's much quieter with less character. The removal of cocktails from the menu with the ability to go get a cocktail myself from a different location was a little strange. Even just a list of available drinks at the pro shop would have been helpful since it wasn't clear what I could order.













API Usage

Daily API limit	5000
Remaining API Calls	5000
30-Day Hit Total	446
30-Day Daily Mean Hits	14.866666666666667
30-Day Daily Median Hits	0.0



```
"reviews": [
       "text": "The first time dining with Eggmann's at their original location I ordered the BLT K's way with avoca
       "rating": 4,
       "time created": "2021-08-31 09:44:14",
           "image_url": "https://s3-media3.fl.yelpcdn.com/photo/GJqoBbR7-BbRipcFBN-1Lw/o.jpg",
           "name": "Lauren E."
       "rating": 5,
       "user": {
          "profile url": "https://www.yelp.com/user details?userid= NpzT8PeNBFTln8FGape9Q",
           "image_url": "https://s3-media2.fl.yelpcdn.com/photo/vI-MYMWWmpKDvNniRoeS1w/o.jpg",
           "name": "Mike K."
       "time_created": "2021-05-02 12:05:56",
          "profile url": "https://www.yelp.com/user details?userid=UiyxQ-7APGGMSeZJOn-oPA",
           "image_url": "https://s3-media2.fl.yelpcdn.com/photo/1LhN30Fr6I5jv_ijk9wf2Q/o.jpg",
```



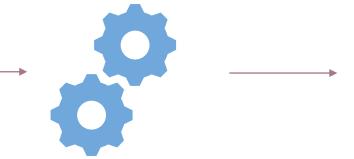


ETL (« Extract Transform and Load »)

```
«EXTRACT»
"text": "The first time dining with Eggmann's at their original location I ordered the BLT K's way with avoca
"time created": "2021-08-31 09:44:14",
"user": {
    "image_url": "https://s3-media3.fl.yelpcdn.com/photo/GJqoBbR7-BbRipcFBN-1Lw/o.jpg",
    "name": "Lauren E."
"text": "OMG... This place is Awesome...\n\nIt is really in a not easy to find location but it is worth find
"rating": 5,
"user": {
    "image_url": "https://s3-media2.fl.yelpcdn.com/photo/vI-MYMWWmpKDvNniRoeS1w/o.jpg",
    "name": "Mike K."
"text": "This place is absolutely amazing!!\nGreat setting on a country road and part of a golf course.\nGreat
"time_created": "2021-05-02 12:05:56",
"user": {
    "image url": "https://s3-media2.fl.yelpcdn.com/photo/1LhN30Fr6I5jv_ijk9wf2Q/o.jpg",
```

«TRANSFORM»

«LOAD»



- yelp_academic_dataset_business_1632329699.csv
- yelp_academic_dataset_review_1632329699.csv
- yelp_academic_dataset_user_1632329699.csv





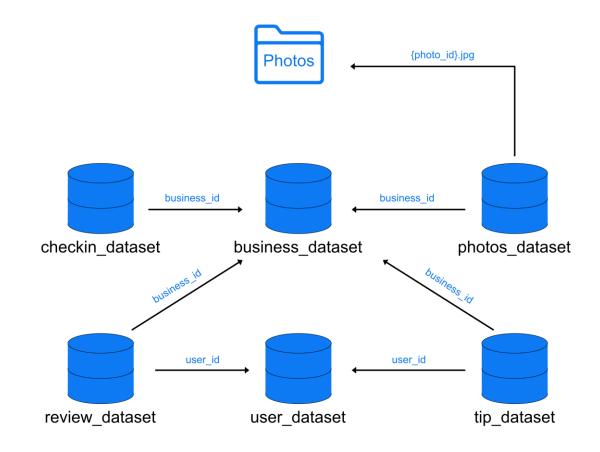
Données Yelp

9 Go fichiers compressés.

16,5 Go décompressés

=>10,5 Go Users / Reviews / Business

=> 6 Go Photos



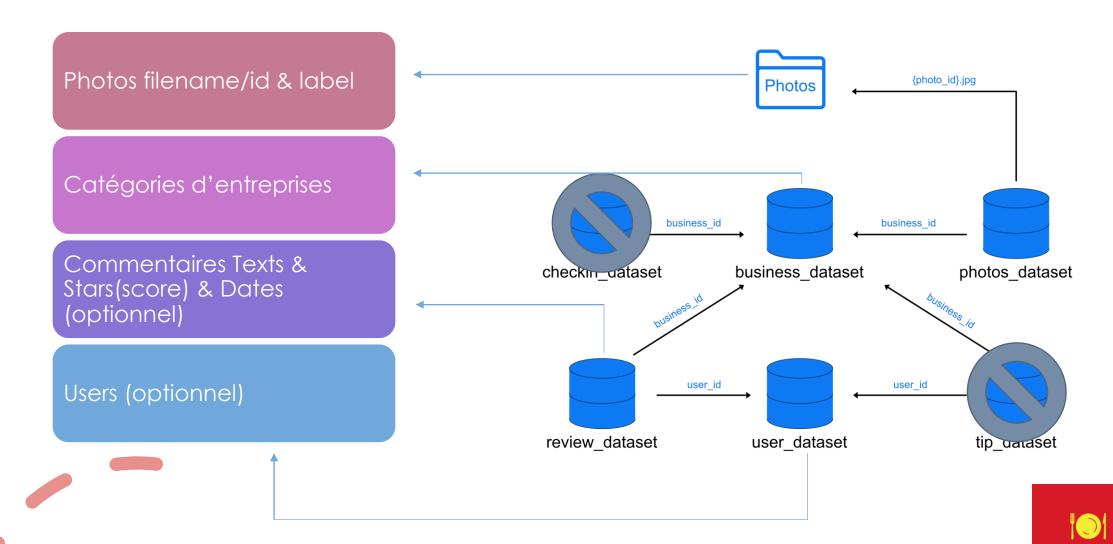




Présentation / Traitement du jeu de données



Sélection des fichiers pertinents



Avis Restau

Traitement business_dataset



1331 catégories d'entreprises et 160585 entreprises



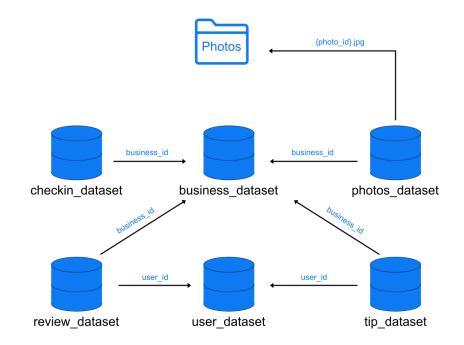
28 catégories d'entreprises et 65927 entreprises restantes



Fichier finaux

Dimensionnalité du nouveau DataFrame : (198979, 3)

	photo_id	label	photo
0	Un_Og6jfhazVn7CxszkKEw	drink	./data/dataset_photo/photos/Un_Og6jfhazVn7Cxsz
1	BFE1AFOs27scnnfeBf99ZA	drink	/data/dataset_photo/photos/BFE1AFOs27scnnfeBf



Dimensionnalité du nouveau DataFrame : (6196998, 8)

user_id	name_user	complete_address	name_business	date	text	stars	review_id
ak0TdVmGKo4pwqdJSTLwWw	Mel	MA - (Beverly) :\n240 Rantoul St \nPostal code	Prides Osteria	2014-10-11 03:34:02	Apparently Prides Osteria had a rough summer a	4.0	0 IWC-xP3rd6obsecCYsGZRg
YoVfDbnlSlW0f7abNQAClg	Brandon	MA - (Framingham) :\n400 Cochituate Rd \nPosta	Target	2015-07-03 20:38:25	This store is pretty good. Not as great as Wal	4.0	1 8bFej1QE5LXp4O05qjGqXA



Traitement NLP (Natural Language Processing)



Approches des recherches

Nettoyage des commentaires

Préparation des données

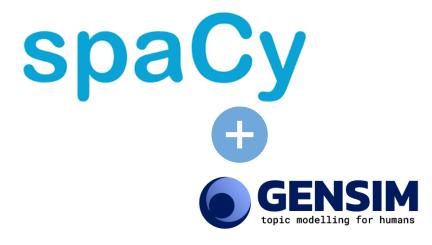
Recherche LDA

Recherche LSA

Nettoyage des données



- Suppression espace
- Caractère alphabétique uniquement
- Suppression de caractères (so goooood => so good)



- Tokenization
- Sélection de « tag » (Nom, Adjectif, Adverbe, Verbe)
- Trigramme
- Lemmatization



Préparation des données

« Every time I come to this location I'm always so disappointed and frustrated with how slow the service is. »

Se.

« It took about 15 minutes to come out. Horrible service I'm waiting 15 minute seating on my chair to get my meal! »



« time come location disappointed frustrated slow service »

&

« take minute come horrible service wait minute seating chair meal »



Préparation des données (Bag Of Word)

« time come location disappointed frustrated slow service »
&

« take minute come horrible service wait minute seating chair meal »

Mot	Occurrence_1	Occurrence_2
time	1	0
come	1	1
location	1	0
disappointed	1	0
frustrated	1	0
slow	1	0
service	1	1
take	0	1
minute	0	2
horrible	0	1
wait	0	1
seating	0	1
chair	0	1
meal	0	1



Préparation des données (TF-IDF) : TF(Time) = 1 : IDF = Log(n+1/df+1) + 1 = Log(3/2)+1

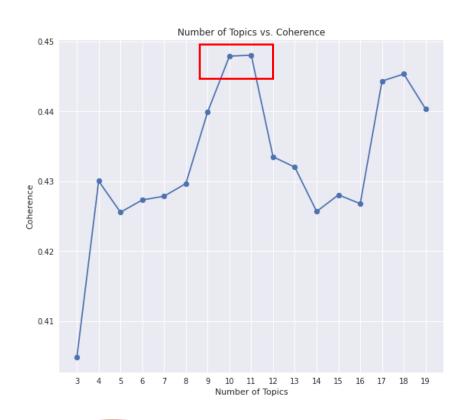
Mot	Occurrence_1	Occurrence_2	TF-IDF_1	TF-IDF_2
time	1	0	1,405	0
come	1	1	1	1
location	1	0	1,405	0
disappointed	1	0	1,405	0
frustrated	1	0	1,405	0
slow	1	0	1,405	0
service	1	1	1	1
take	0	1	0	1,405
minute	0	2	0	2,81
horrible	0	1	0	1,405
wait	0	1	0	1,405
seating	0	1	0	1,405
chair	0	1	0	1,405
meal	0	1	0	1,405

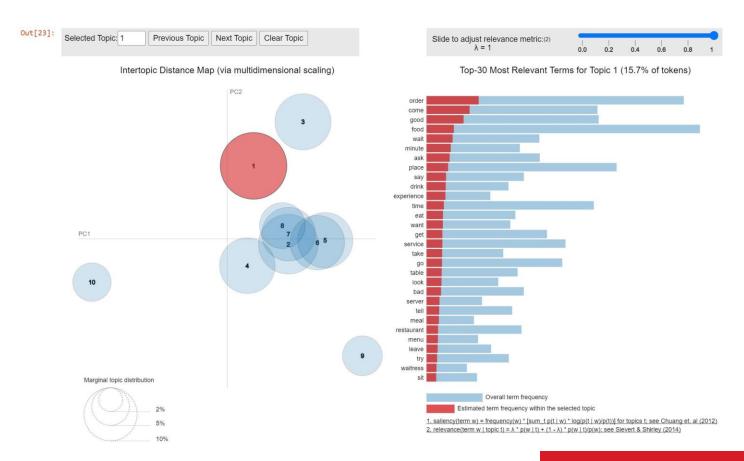
wait 1.4054651081081644

- TFIDF = 1*1,405 = 1,405

```
tfidf = TfidfVectorizer(norm=None)
data = tfidf.fit_transform(["time come location disappointed frustrated slow service",
                           "take minute come horrible service wait minute seating chair meal"])
for row in data.toarray():
   print("----")
   for k,v in zip(tfidf.get_feature_names_out(),row):
       print(k,v)
-----
chair 0.0
come 1.0
disappointed 1.4054651081081644
frustrated 1.4054651081081644
horrible 0.0
location 1.4054651081081644
meal 0.0
minute 0.0
seating 0.0
service 1.0
slow 1.4054651081081644
take 0.0
time 1.4054651081081644
wait 0.0
-----
chair 1.4054651081081644
come 1.0
disappointed 0.0
frustrated 0.0
horrible 1.4054651081081644
location 0.0
meal 1.4054651081081644
minute 2.8109302162163288
seating 1.4054651081081644
service 1.0
slow 0.0
take 1.4054651081081644
                                                                         Avis Restau
time 0.0
```

Recherches LDA







Recherches LSA

Test itératif

Topic #0

restaurant bad table ask get go minute wait good service come time place order food

Topic #1

min customer take order manager drink sit hour seat say ask tell table minute wait

Topic #2

sauce pick topping hour deliver say tell slice phone crust cheese call delivery order pizza

Topic #3

horrible price well bar great terrible staff good rude customer bad food place service pizza



Topic #0

restaurant bad table ask get go minute wait good service come time place order food

opic #1

min customer take order manager drink sit seat hour say ask tell table minute wait

Topic #

sauce pick topping hour deliver say tell slice phone crust cheese call delivery order pizza

Topic #3

take meal say sauce ask come salad get sandwich rice minute burger fry chicken order

Topic #4

min time cold take horrible delivery terrible hour slow minute bad wait order service food

Topic #5

restaurant food waiter bar waitress good sit come server seat drink minute wait table pizza

Topic #6

location people hour good drive coffee long sandwich time line fry minute place wait burger

Tonic #7

terrible cheese take waitress bartender server come ask bar beer fry order service drink burger

onic #8

hour coffee menu time delivery sushi bartender price food good beer bar place drink order

Topic #9

location terrible wait taste customer bartender bad slow minute bar sandwich service coffee drink chicken

opic #10

ask small bread server service come good cheese breakfast slow location time price sandwich coffee

onic #11

long bad year waste visit salad night second drink server eat go chicken location time

Topic #12

go egg waitress food water breakfast cold taco get taste eat coffee place come bad

onic #13

experience wait fish time dish burger rice service restaurant table roll place order bad sushi

onic #14

dirty sit breakfast server slow restaurant cheese order service bread lunch salad table place sandwich

Topic #15

bartender pay steak eat get salad restaurant experience sandwich price wait drink bar hour bad

Topic #1

well chicken horrible money service time worth ask pay charge wait sushi price place come

Tonic #17

place night pretty reservation close pm tell coffee service say call delivery bad hour good

Topic #18

go sauce sit sandwich min good customer sushi bar fish service table cheese



Traitement CV (Computer Vision)



Approches des recherces

Traitement des images

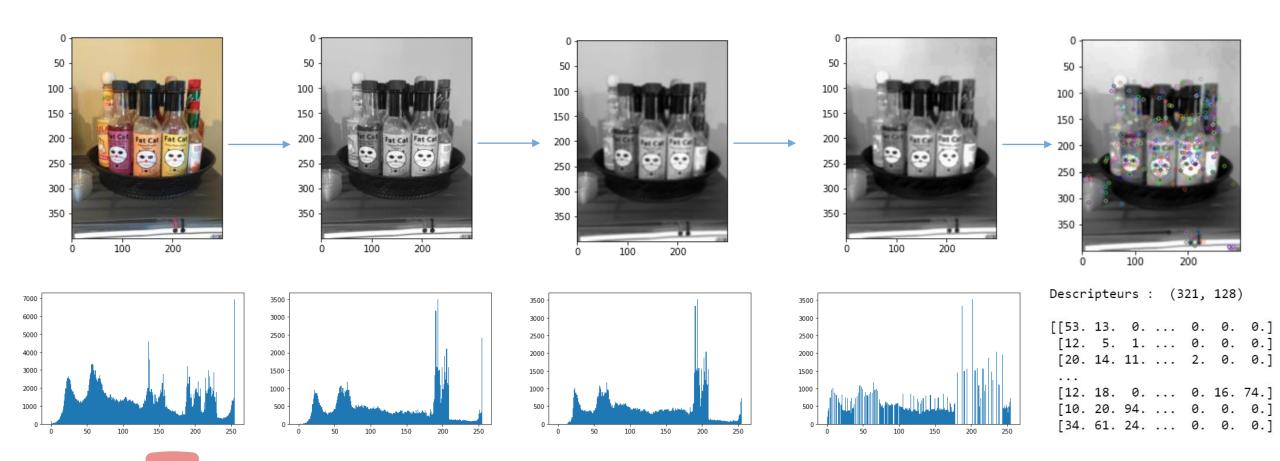
Visualisation SIFT/Kmeans (2D)

Classification avec SIFT & Bag Of Visual Word

Approches CNN



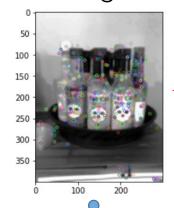
Traitement des images





Visualisation SIFT / Kmeans

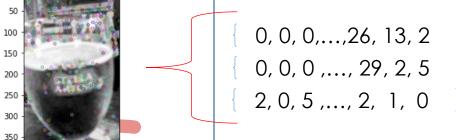
Descripteur par images



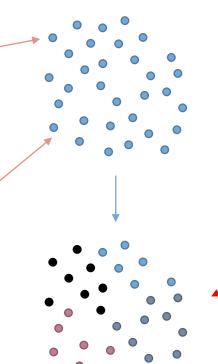
Descripteur de toutes les images

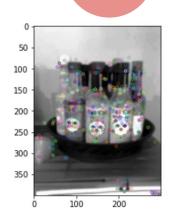
53, 13, 0,...,0, 0, 0 12, 5, 1,..., 0, 0, 0 20,14, 11,...,2, 0, 0





Kmeans des descripteurs





53, 13, 0,...,0, 0, 0 12, 5, 1 ,..., 0, 0, 0 20,14, 11,...,2, 0, 0



« Prédiction » kmeans

Bag Of Visual Word

0.004, 0.2, 0, ..., 0, 0.02



Visualisation SIFT / Kmeans

Bag Of Visual Word

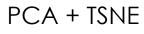
0.004, 0.2, 0, ..., 0, 0.02

0.01, 0, 0.1, ..., 0, 0.05



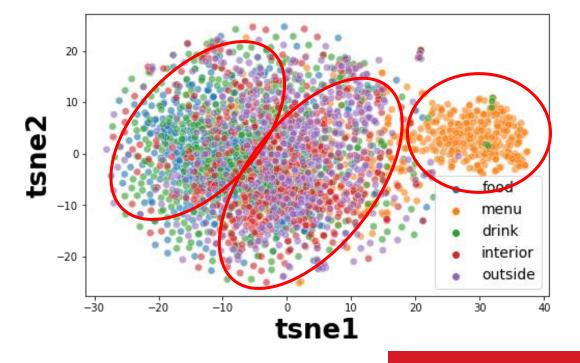
0.02, 0, 0, ..., 0.5, 0.009

0.006, 0, 0, ..., 0.2, 0.02





TSNE selon les vraies classes

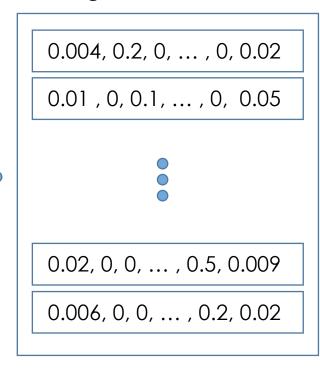




Classification SIFT

Bag Of Visual Word

12500 images : =>10000 train_set =>2500 test_set



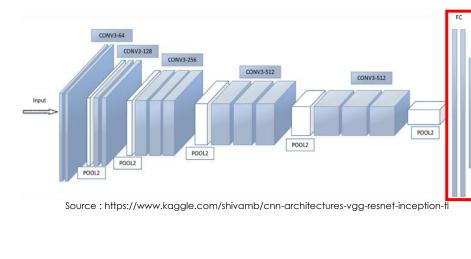




Classification CNN

Model: "model_8"

-		
Layer (type)	Output Shape	Param #
input_7 (InputLayer)	[(None, 224, 224, 3)]	0
block1_conv1 (Conv2D)	(None, 224, 224, 64)	1792
block1_conv2 (Conv2D)	(None, 224, 224, 64)	36928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	0
block2_conv1 (Conv2D)	(None, 112, 112, 128)	73856
block2_conv2 (Conv2D)	(None, 112, 112, 128)	147584
block2_pool (MaxPooling2D)	(None, 56, 56, 128)	0
block3_conv1 (Conv2D)	(None, 56, 56, 256)	295168
block3_conv2 (Conv2D)	(None, 56, 56, 256)	590080
block3_conv3 (Conv2D)	(None, 56, 56, 256)	590080
block3_pool (MaxPooling2D)	(None, 28, 28, 256)	0
block4_conv1 (Conv2D)	(None, 28, 28, 512)	1180160
block4_conv2 (Conv2D)	(None, 28, 28, 512)	2359808
block4_conv3 (Conv2D)	(None, 28, 28, 512)	2359808
block4_pool (MaxPooling2D)	(None, 14, 14, 512)	0
block5_conv1 (Conv2D)	(None, 14, 14, 512)	2359808
block5_conv2 (Conv2D)	(None, 14, 14, 512)	2359808
block5_conv3 (Conv2D)	(None, 14, 14, 512)	2359808
block5_pool (MaxPooling2D)	(None, 7, 7, 512)	0
flatten_3 (Flatten)	(None, 25088)	0
fc1 (Dense)	(None, 128)	3211392
dropout_3 (Dropout)	(None, 128)	0



	food	menu	drink	interior	outside
food	87	0	10	2	1
menu	1	83	1	3	0
drink	3	0	102	1	2
interior	2	2	2	97	8
outside	0	1	2	7	83

Total params: 17,926,725

predictions (Dense)

Trainable params: 3,212,037 Non-trainable params: 14,714,688

(None, 5)

645



Conclusion / Dashbord



Conclusion



PREMIER PROJET NLP / CV



LABÉLISATION DES IMAGES FAISABLES

&
TOPIC MODELING POSSIBLE MAIS
DIFFICILE



AMÉLIORATION POSSIBLE DES SOLUTIONS FINALES



DIFFICULTÉ À TROUVER DES SUJETS INTÉRESSANTS (NLP)





Merci de votre attention, avez-vous des questions?