Context

Although this dataset was originally contributed to the UCI Machine Learning repository nearly 30 years ago, mushroom hunting (otherwise known as "shrooming") is enjoying new peaks in popularity. Learn which features spell certain death and which are most palatable in this dataset of mushroom characteristics. And how certain can your model be?

Content

This dataset includes descriptions of hypothetical samples corresponding to 23 species of gilled mushrooms in the Agaricus and Lepiota Family Mushroom drawn from The Audubon Society Field Guide to North American Mushrooms (1981). Each species is identified as definitely edible, definitely poisonous, or of unknown edibility and not recommended. This latter class was combined with the poisonous one. The Guide clearly states that there is no simple rule for determining the edibility of a mushroom; no rule like "leaflets three, let it be" for Poisonous Oak and lvy.

Time period: Donated to UCI ML 27 April 1987

Inspiration

What types of machine learning models perform best on this dataset?

Which features are most indicative of a poisonous mushroom?



```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

!pip install dataprep

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Requirement already satisfied: dataprep in /usr/local/lib/python3.7/dist-packages (@Requirement already satisfied: pydantic<2.0,>=1.6 in /usr/local/lib/python3.7/dist-packages Requirement already satisfied: dask[array,dataframe,delayed]<3.0,>=2.25 in /usr/locaRequirement already satisfied: numpy<2,>=1 in /usr/local/lib/python3.7/dist-packages Requirement already satisfied: nltk<4.0,>=3.5 in /usr/local/lib/python3.7/dist-packa Requirement already satisfied: regex<2021.0.0,>=2020.10.15 in /usr/local/lib/python3 Requirement already satisfied: scipy<2,>=1 in /usr/local/lib/python3.7/dist-packages Requirement already satisfied: bokeh<3,>=2 in /usr/local/lib/python3.7/dist-packages Requirement already satisfied: tqdm<5.0,>=4.48 in /usr/local/lib/python3.7/dist-packages Requirement already satisfied: pandas<2,>=1 in /usr/local/lib/python3.7/dist-packages Requirement already satisfied: pandas<2,>=1 in /usr/local/lib/python3.7/dist-packages
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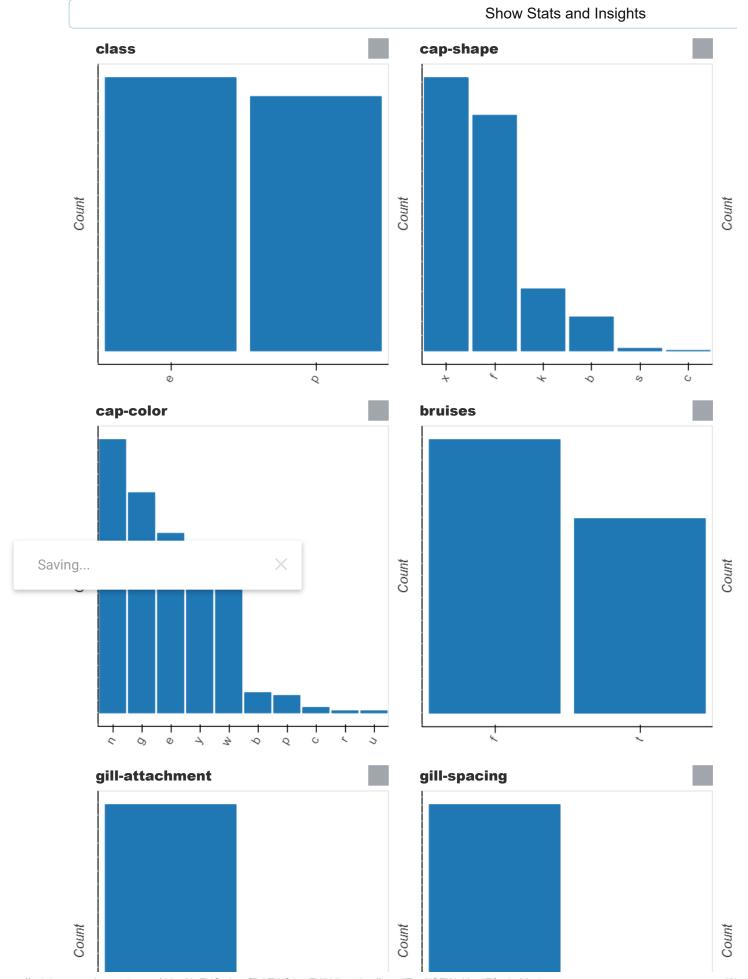
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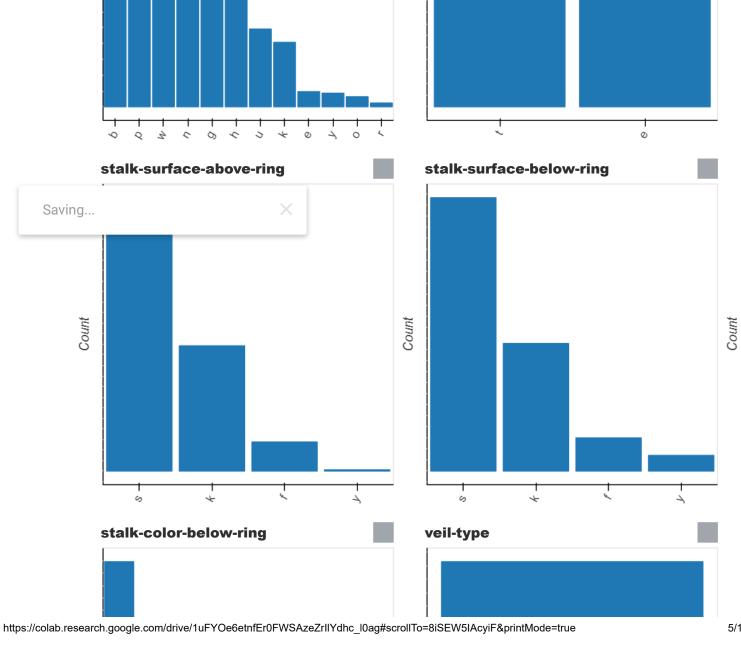
```
from dataprep.eda import create report, plot, plot correlation, plot missing
df = pd.read csv("/content/drive/MyDrive/Colab Notebooks/kaggle/mushrooms.csv")
df.head()
```

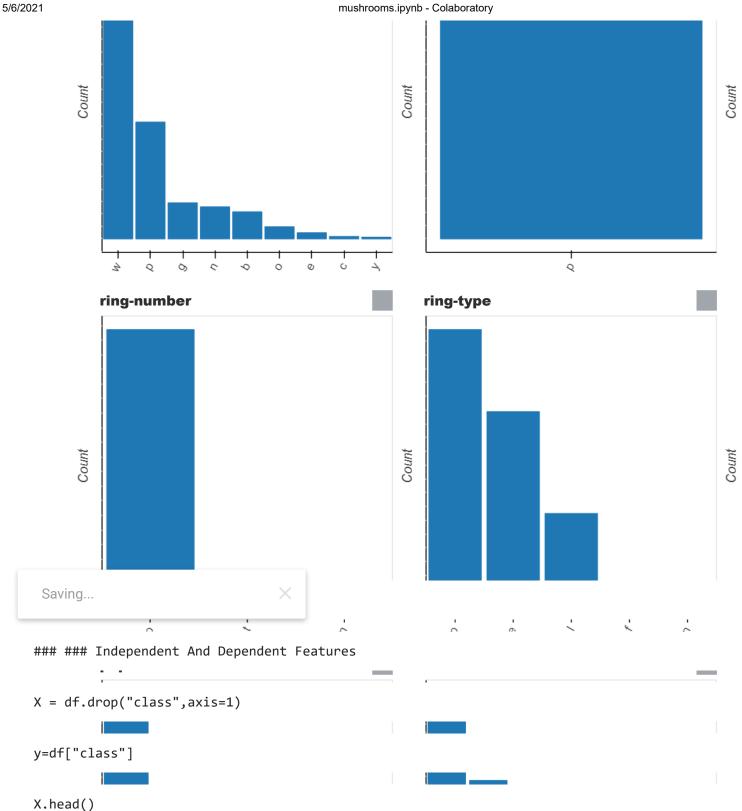
https://colab.research.google.com/drive/1uFYOe6etnfEr0FWSAzeZrllYdhc_l0ag#scrollTo=8iSEW5IAcyiF&printMode=true

	class	cap- shape	cap- surface	cap- color	bruises	odor	gill- attachment	gill- spacing	gill- size	gill- color	stall shap
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1	е	х	S	у	t	а	f	С	b	k	
2	е	b	S	W	t	1	f	С	b	n	
3	р	Х	у	W	t	р	f	С	n	n	
4	е	х	s	g	f	n	f	W	b	k	



gill-color





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                   cap-
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y.value counts(normalize=True)
     e
          0.517971
          0.482029
     Name: class, dtype: float64
!python setup.py install
!python -m pip install --upgrade pip setuptools wheel
!pip install evalml
     python3: can't open file 'setup.py': [Errno 2] No such file or directory
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Python3: can't open file 'setup.py': [Errno 2] No such file or directory
Requirement already satisfied: pip in /usr/local/lib/python3.7/dist-packages (21.1.1
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WARNING: Running pip as root will break packages and permissions. You should install
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     Requirement already satisfied: ipython>=4.0.0 in /usr/local/lib/python3.7/dist-packa
import evalml
from evalml import AutoMLSearch
### Train Test Split
X train,X test,y train,y test = evalml.preprocessing.split data(X,y,problem type="binary")
 Saving...
                                   ll problem types
     [<ProblemTypes.BINARY: 'binary'>,
      <ProblemTypes.MULTICLASS: 'multiclass'>,
      <ProblemTypes.REGRESSION: 'regression'>,
      <ProblemTypes.TIME_SERIES_REGRESSION: 'time series regression'>,
      <ProblemTypes.TIME SERIES BINARY: 'time series binary'>,
      <ProblemTypes.TIME_SERIES_MULTICLASS: 'time series multiclass'>]
type(X_train)
     woodwork.datatable.DataTable
X train.head()
```

Data Column	cap-shape	cap-surface	cap-color	bruises	odor	gill- attachn
Physical Type	category	category	category	category	category	categor
Logical Type	Categorical	Categorical	Categorical	Categorical	Categorical	Categor
Semantic Tag(s)	['category']	['category']	['category']	['category']	['category']	['cate¿
6373	f	S	n	f	f	
4331	k	у	n	f	n	
6496	f	8	P.	f	V	

AutoMLSearch

AutoMLSearch give us most effective and accurate ML model or model to fit datasets

automl.search()

```
**********
    *********
    * Beginning pipeline search *
    * Beginning pipeline search *
    **********
    *********
    Optimizing for F1.
    Optimizing for F1.
    Greater score is better.
    Greater score is better.
   Using SequentialEngine to train and score pipelines.
   Using SequentialEngine to train and score pipelines.
    Searching up to 1 batches for a total of 9 pipelines.
    Searching up to 1 batches for a total of 9 pipelines.
    Allowed model families: xgboost, decision tree, lightgbm, linear model, extra trees, ca
   Allowed model families: xgboost, decision tree, lightgbm, linear model, extra trees, ca
    Evaluating Baseline Pipeline: Mode Baseline Binary Classification Pipeline
    Evaluating Baseline Pipeline: Mode Baseline Binary Classification Pipeline
    All predicted probabilities have the same value: [0.]
    Extreme threshold of 0.0
    All predicted probabilities have the same value: [0.]
    Extreme threshold of 0.0
    All predicted probabilities have the same value: [0.]
    Extreme threshold of 0.0
                                 ation Pipeline:
Saving...
                              ation Pipeline:
                                 ion
           Starting cross validation
                   Training and scoring on fold 0
                           Fold 0: starting training
                           Fold 0: finished training
                           Fold 0: Optimal threshold found (0.000)
                           Fold 0: Scoring trained pipeline
                           Fold 0: F1 score: 0.000
                   Training and scoring on fold 1
                           Fold 1: starting training
                           Fold 1: finished training
                           Fold 1: Optimal threshold found (0.000)
                           Fold 1: Scoring trained pipeline
                           Fold 1: F1 score: 0.000
                   Training and scoring on fold 2
                           Fold 2: starting training
                           Fold 2: finished training
                           Fold 2: Optimal threshold found (0.000)
                           Fold 2: Scoring trained pipeline
                           Fold 2: F1 score: 0.000
           Finished cross validation - mean F1: 0.000
           Finished cross validation - mean F1: 0.000
```

```
**********
    **********
    * Evaluating Batch Number 1 *
    * Evaluating Batch Number 1 *
    **********
    ***********
    Elastic Net Classifier w/ Imputer + One Hot Encoder + Standard Scaler:
    Elastic Net Classifier w/ Imputer + One Hot Encoder + Standard Scaler:
           Starting cross validation
           Starting cross validation
                   Training and scoring on fold 0
                           Fold 0: starting training
                           Fold 0: finished training
                           Fold 0: Optimal threshold found (0.533)
                           Fold 0: Scoring trained pipeline
                           Fold 0: F1 score: 0.888
                   Training and scoring on fold 1
                           Fold 1: starting training
                           Fold 1: finished training
                           Fold 1: Optimal threshold found (0.535)
                           Fold 1: Scoring trained pipeline
                           Fold 1: F1 score: 0.896
                   Training and scoring on fold 2
                           Fold 2: starting training
                           Fold 2: finished training
                           Fold 2: Optimal threshold found (0.535)
                           Fold 2: Scoring trained pipeline
                           Fold 2: F1 score: 0.889
           Finished cross validation - mean F1: 0.891
           Finished cross-validation - mean F1: 0.891
Saving...
   Extreme threshold of 1.0
   Decision Tree Classifier w/ Imputer + One Hot Encoder:
   Decision Tree Classifier w/ Imputer + One Hot Encoder:
           Starting cross validation
           Starting cross validation
                   Training and scoring on fold 0
                           Fold 0: starting training
                           Fold 0: finished training
                           Fold 0: Optimal threshold found (0.778)
                           Fold 0: Scoring trained pipeline
                           Fold 0: F1 score: 0.950
                   Training and scoring on fold 1
                           Fold 1: starting training
                           Fold 1: finished training
                           Fold 1: Optimal threshold found (0.818)
                           Fold 1: Scoring trained pipeline
                           Fold 1: F1 score: 0.946
                   Training and scoring on fold 2
                           Fold 2: starting training
                           Fold 2: finished training
                           Fold 2: Optimal threshold found (0.826)
                           Fold 2: Scoring trained pipeline
                           Fold 2: F1 score: 0.947
```

```
Finished cross validation - mean F1: 0.948
            Finished cross validation - mean F1: 0.948
    Random Forest Classifier w/ Imputer + One Hot Encoder:
    Random Forest Classifier w/ Imputer + One Hot Encoder:
            Starting cross validation
            Starting cross validation
                    Training and scoring on fold 0
                            Fold 0: starting training
                            Fold 0: finished training
                            Fold 0: Optimal threshold found (0.397)
                            Fold 0: Scoring trained pipeline
                            Fold 0: F1 score: 0.999
                    Training and scoring on fold 1
                            Fold 1: starting training
                            Fold 1: finished training
                            Fold 1: Optimal threshold found (0.500)
                            Fold 1: Scoring trained pipeline
                            Fold 1: F1 score: 0.994
                    Training and scoring on fold 2
                            Fold 2: starting training
                            Fold 2: finished training
                            Fold 2: Optimal threshold found (0.453)
                            Fold 2: Scoring trained pipeline
                            Fold 2: F1 score: 0.999
            Finished cross validation - mean F1: 0.997
            Finished cross validation - mean F1: 0.997
    LightGBM Classifier w/ Imputer + One Hot Encoder:
    LightGBM Classifier w/ Imputer + One Hot Encoder:
            Starting cross validation
            Starting cross validation
                    Training and scoring on fold 0
                            Fold 0: starting training
                                   : finished training

    : Optimal threshold found (1.000)

Saving...
                                   : Scoring trained pipeline
                            Fold 0: F1 score: 1.000
                    Training and scoring on fold 1
                            Fold 1: starting training
                            Fold 1: finished training
                            Fold 1: Optimal threshold found (1.000)
                            Fold 1: Scoring trained pipeline
                            Fold 1: F1 score: 1.000
                    Training and scoring on fold 2
                            Fold 2: starting training
                            Fold 2: finished training
                            Fold 2: Optimal threshold found (1.000)
                            Fold 2: Scoring trained pipeline
                            Fold 2: F1 score: 0.998
            Finished cross validation - mean F1: 0.999
            Finished cross validation - mean F1: 0.999
    Logistic Regression Classifier w/ Imputer + One Hot Encoder + Standard Scaler:
    Logistic Regression Classifier w/ Imputer + One Hot Encoder + Standard Scaler:
            Starting cross validation
            Starting cross validation
                    Training and scoring on fold 0
                            Fold 0: starting training
                            Fold 0: finished training
                            Fold A. Ontimal threshold found (A Q5Q)
```

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רסבע ש. טענבווומב נווו כאווטבע וטעווע (ש. דיס)
                            Fold 0: Scoring trained pipeline
                            Fold 0: F1 score: 1.000
                    Training and scoring on fold 1
                            Fold 1: starting training
                            Fold 1: finished training
                            Fold 1: Optimal threshold found (0.983)
                            Fold 1: Scoring trained pipeline
                            Fold 1: F1 score: 0.999
                    Training and scoring on fold 2
                            Fold 2: starting training
                            Fold 2: finished training
                            Fold 2: Optimal threshold found (0.986)
                            Fold 2: Scoring trained pipeline
                            Fold 2: F1 score: 0.997
            Finished cross validation - mean F1: 0.998
            Finished cross validation - mean F1: 0.998
    XGBoost Classifier w/ Imputer + One Hot Encoder:
   XGBoost Classifier w/ Imputer + One Hot Encoder:
            Starting cross validation
            Starting cross validation
                    Training and scoring on fold 0
                            Fold 0: starting training
                            Fold 0: finished training
                            Fold 0: Optimal threshold found (0.933)
                            Fold 0: Scoring trained pipeline
                            Fold 0: F1 score: 1.000
                    Training and scoring on fold 1
                            Fold 1: starting training
                            Fold 1: finished training
                            Fold 1: Optimal threshold found (0.952)
                            Fold 1: Scoring trained pipeline
                            Fold 1: F1 score: 1.000
                                   coring on fold 2
                               × starting training
Saving...
                             ____: finished training
                            Fold 2: Optimal threshold found (0.971)
                            Fold 2: Scoring trained pipeline
                            Fold 2: F1 score: 0.996
            Finished cross validation - mean F1: 0.998
            Finished cross validation - mean F1: 0.998
    Extra Trees Classifier w/ Imputer + One Hot Encoder:
    Extra Trees Classifier w/ Imputer + One Hot Encoder:
            Starting cross validation
            Starting cross validation
                    Training and scoring on fold 0
                            Fold 0: starting training
                            Fold 0: finished training
                            Fold 0: Optimal threshold found (0.487)
                            Fold 0: Scoring trained pipeline
                            Fold 0: F1 score: 0.997
                    Training and scoring on fold 1
                            Fold 1: starting training
                            Fold 1: finished training
                            Fold 1: Optimal threshold found (0.441)
                            Fold 1: Scoring trained pipeline
                            Fold 1: F1 score: 0.997
                    Training and scoring on fold 2
```

```
mushrooms.ipynb - Colaboratory
                        Fold 2: starting training
                        Fold 2: finished training
                        Fold 2: Optimal threshold found (0.467)
                        Fold 2: Scoring trained pipeline
                        Fold 2: F1 score: 0.999
        Finished cross validation - mean F1: 0.998
        Finished cross validation - mean F1: 0.998
CatBoost Classifier w/ Imputer:
CatBoost Classifier w/ Imputer:
        Starting cross validation
        Starting cross validation
                Training and scoring on fold 0
                        Fold 0: starting training
                        Fold 0: finished training
                        Fold 0: Optimal threshold found (0.835)
                        Fold 0: Scoring trained pipeline
                        Fold 0: F1 score: 0.983
                Training and scoring on fold 1
                        Fold 1: starting training
                        Fold 1: finished training
                        Fold 1: Optimal threshold found (0.841)
                        Fold 1: Scoring trained pipeline
                        Fold 1: F1 score: 0.984
                Training and scoring on fold 2
                        Fold 2: starting training
                        Fold 2: finished training
```

Fold 2: Optimal threshold found (0.816)

automl.rankings

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2 1				IIIu	silloonis.ipyrib - Colaboratory			
	i	d p	ipeline_name	mean_cv_score	standard_deviation	n_cv_score	validation_score	per
	0 4	4	LightGBM Classifier w/ Imputer + One Hot Encoder	0.998881		0.001109	0.999521	
	1 :	5	Logistic Regression Classifier w/ Imputer + On	0.998400		0.001210	0.999521	
:	2 (6	XGBoost Classifier w/ Imputer + One Hot Encoder	0.998238		0.002223	0.999521	
	o .	7	Extra Trees Classifier w/	0 007500		0 001272	0 007121	
utoml	.des	crib	e_pipeline(4)					
************************************* * LightGBM Classifier w/ Imputer + One Hot Encoder * * LightGBM Classifier w/ Imputer + One Hot Encoder * **********************************								
		a Т	······································	~				
Saving		Fam	nly: LightGBM					
P = = 1	ipel	ine ==== ==== pute	* categorical * categorical * numeric_imp * numeric_imp * categorical * categorical * numeric_fil	impute_strategoute_stratego: oute_stratego: oute_stratego:fill_value: Nfill_value: N	mean None			
2	. On	е На	r numeric_fii t Encoder	l_value : None				

* features_to_encode : None

One Hot Encoder
 One Hot Encoder

* top_n : 10 * top_n : 10

```
* features to encode : None
              * categories : None
               categories : None
              * drop : if binary
              * drop : if binary
              * handle_unknown : ignore
              * handle unknown : ignore
              * handle_missing : error
              * handle missing : error
     3. LightGBM Classifier
     LightGBM Classifier
              * boosting_type : gbdt
              * boosting type : gbdt
              * learning_rate : 0.1
              * learning rate : 0.1
              * n estimators : 100
              * n estimators : 100
              * max depth : 0
              * max depth : 0
              * num leaves : 31
              * num leaves : 31
              * min child samples : 20
pipeline = automl.get_pipeline(4)
print(pipeline.parameters)
     {'Imputer': {'categorical_impute_strategy': 'most_frequent', 'numeric_impute_strategy':
Saving...
                                                             One Hot Encoder
                                                                                             b
                    Imputer
                                                                                             le
                                                        top n:10
                                                                                             n
    egorical impute strategy: most frequent
                                                        features to encode: None
                                                                                             \mathbf{m}
                                                        categories: None
    meric impute strategy: mean
                                                                                             n
    egorical fill value: None
                                                        drop: if binary
                                                                                             \mathbf{m}
    meric fill value: None
                                                        handle unknown: ignore
                                                                                             n
                                                        handle missing : error
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```

▼ Conclusion

We have a simple overview of some techniques and algorithms in machine learning. Furthermore, there are more and more techniques apply machine learning as a solution. The best model seem to

be the LightGBM Classifier w/ Imputer + One Hot Encoder model, which is the model with the mushrooms dataset The model generated a 'F1 score' of 0.998881.LightGBM Classifier w/ Imputer + One Hot Encoder is a great algorithm for classification problems to produce a predictive model.its default hyperparameters already return great results. So this project gives different ways of adressing the task with unbalanced data.Using Supervised learning Elastic Net Classifier, XGBoost Classifier, CatBoost Classifier, DecisionTreeClassifier,Logistic Regression and Random Forest.

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