



< Return to Classroom

# Disaster Response Pipeline

**REVIEW** 

HISTORY

### **Requires Changes**

## 2 specifications require changes

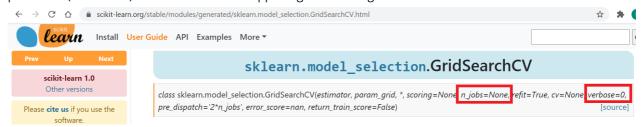
Dear Learner,

Awesome work is done in this submission. You have completed the tasks amazingly well. Overall brilliant work.

You almost completed the project, few minor additions, which I am sure you can fix in no time. Kindly follow the comments in the corresponding specifications.

I believe you will have further fun and a learning experience working through them. Feel free to post your specific doubts at "https://knowledge.udacity.com/"

As far as your doubt is concerned, one of the ways to improve the time is to use the n\_jobs parameter(n\_jobs=-1) to use all the cores of the CPU. You can also make use of the verbose parameter(verbose=4) to visualize what is happening in the background.



You can refer to the document at- https://scikit-

 $learn.org/stable/modules/generated/sklearn.model\_selection.GridSearchCV.html\\$ 

Keep learning.

Good luck !!!

## **Github & Code Quality**



All project code is stored in a GitHub repository and a link to the repository has been provided for reviewers. The student made at least 3 commits to this repository.

Awesome, All project code is pushed to your GitHub repository.



The README file includes a summary of the project, how to run the Python scripts and web app, and an explanation of the files in the repository. Comments are used effectively and each function has a docstring.

Perfect, your README file is properly documented mentioning about every step very clearly.



Scripts have an intuitive, easy-to-follow structure with code separated into logical functions. Naming for variables and functions follows the PEP8 style guidelines.

Nice!,

Scripts have an intuitive, easy-to-follow structure with code separated into logical functions. It's nice to see that you have used proper "docstrings" into your functions.

#### **ETL**



The ETL script, process\_data.py, runs in the terminal without errors. The script takes the file paths of the two datasets and database, cleans the datasets, and stores the clean data into a SQLite database in the specified database file path.

Yes, The ETL script, process\_data.py, runs in the terminal without errors accepting required arguments.

#### Recommendation

```
def save_data(df, database_filename):
 """used to save given dataframe to SQLite database
 args:
 - df: dataframe
                                                                        It is nice to see that you have
 - database_filename: SQLite DB name
                                                                        added code to drop the table if
                                                                        it exists.
 table_name = 'message_category'
# instantiate DB engine
 engine = create_engine(f'sqlite:///{databas
                                              filename}')
 connection = engine.raw_connection()
 # get DB cursor so that we can execute SQL query
 cursor = connection.cursor()
 query = f"DROP TABLE IF EXISTS {table_name}"
                                                                              An easy way to implement the same thing
 cursor.execute(query)
                                                                              is to just add if_exists= 'replace' as one of the parameters.
 connection.commit() # commit the change
 df.to_sql(table_name, engine, index=False)
 # shut down DB engine
 cursor.close()
 engine.dispose()
```



The script successfully follows steps to clean the dataset. It merges the messages and categories datasets, splits the categories column into separate, clearly named columns, converts values to binary, and drops duplicates.

- It merges the messages and categories datasets  $\sqrt{\phantom{a}}$
- Splits the categories column into separate, clearly named columns 🗸
- Converts values to binary X
- Drops duplicates

# **Explanation**

The values in your cleaned data\_frame should contain only binary values(i.e either 0 or 1). But if you watch closely the "categories" column into the disaster\_categories.csv file. You will find that certain "related" values are "related-2" which doesn't make any sense.

Snippet from disaster\_categories.csv:-

	Α	В	С	D	Е	F	G	Н
118	144	related-1;re	equest-1;off	er-0;aid_rela	ated-1;med	ical_help-0;r	nedical_pro	ducts-0;sea
119	146	related-2 re	equest-0;off	er-0;aid_rela	ated-0;medi	ical_help-0;r	nedical_pro	ducts-0;sea
120	147	related-1;re	equest-1;off	er-0;aid_rela	ated-1;medi	ical_help-0;r	nedical_pro	ducts-0;sea
121	149	related-1;re	equest-1;off	er-0;aid_rela	ated-1;medi	ical_help-0;r	nedical_pro	ducts-0;sea
122	150	related-1;re	equest-1;off	er-0;aid_rela	ated-1;medi	ical_help-0;r	nedical_pro	ducts-0;sea
123	151	related-0;re	equest-0;off	er-0;aid_rela	ated-0;medi	ical_help-0;r	nedical_pro	ducts-0;sea

#### Hint

You can check for different columns something like:-

Kindly clean that. You can drop the entire row as the cleaning process.

# **Machine Learning**



The machine learning script, train\_classifier.py, runs in the terminal without errors. The script takes the database file path and model file path, creates and trains a classifier, and stores the classifier into a pickle file to the specified model file path.



The script uses a custom tokenize function using nltk to case normalize, lemmatize, and tokenize text. This function is used in the machine learning pipeline to vectorize and then apply TF-IDF to the text.



The script builds a pipeline that processes text and then performs multi-output classification on the 36 categories in the dataset. GridSearchCV is used to find the best parameters for the model.



The TF-IDF pipeline is only trained with the training data. The f1 score, precision and recall for the test set is outputted for each category.

# **Deployment**



The web app, run.py, runs in the terminal without errors. The main page includes at least two visualizations using data from the SQLite database.

The web app, run.py, runs in the terminal without errors.  $\sqrt{\phantom{a}}$ 



The main page must include at least two visualizations using data from the SQLite database. KIndly add atleast one more plot.

Just complete the TODO task in the /app/run.py module.



When a user inputs a message into the app, the app returns classification results for all 36 categories.

Classification results are provided based on user inputs.





Learn the best practices for revising and resubmitting your project.



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