

EXPLORE || DIGITAL SKILLS

Setting up Comet

Version control



Overview

Requirements

Setting up Comet account

Your first experiment

Conclusion

Learning Objectives

Version Control

Comet.ml

Why this is important

Project management
Reproducibility
Collaboration

Easy version control
You choose what to keep
Records per experiment



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Have your Comet starter notebook open (found under trains) if you would like to follow along for the “Your first experiment” section.



comet





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- Visit <https://www.comet.ml> and **create a free account**

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- Sign up with GitHub or fill in your details

Step 3

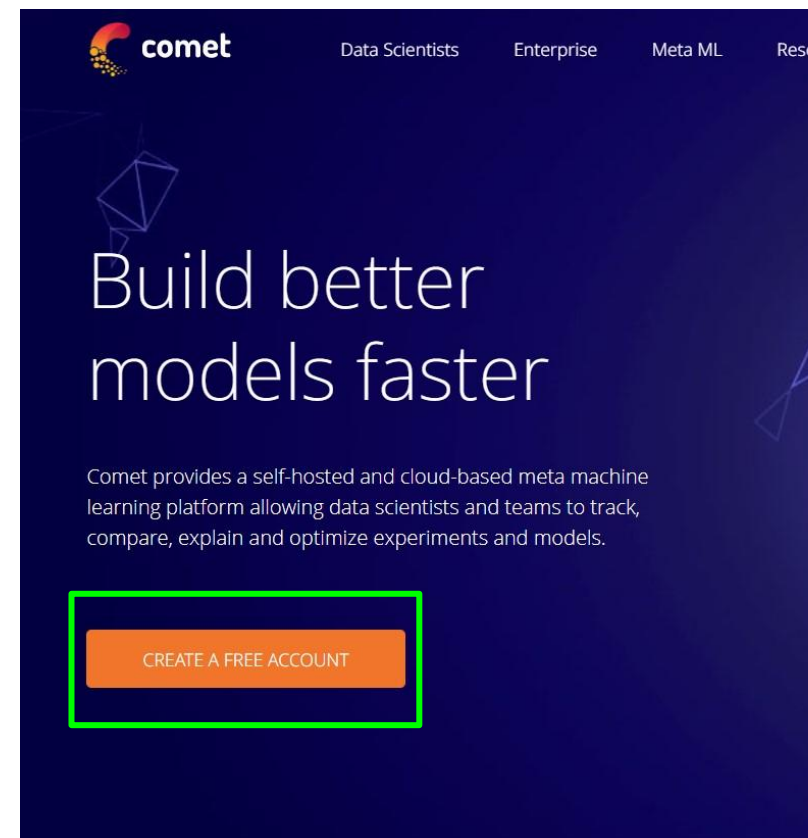
- Choose the **individual** option

Step 4

- On the quick start page, have a look at the sklearn guide

Step 5

- `pip install comet_ml` from your notebook or shell (If you are using colab you will have to do this every time)



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Sign up for Comet Back


[Sign in with GitHub](#)

or

Email *

Username *

First name *

Last name *

Password *

Repeat password *

☐ Send me monthly Comet product updates

[Create account](#)

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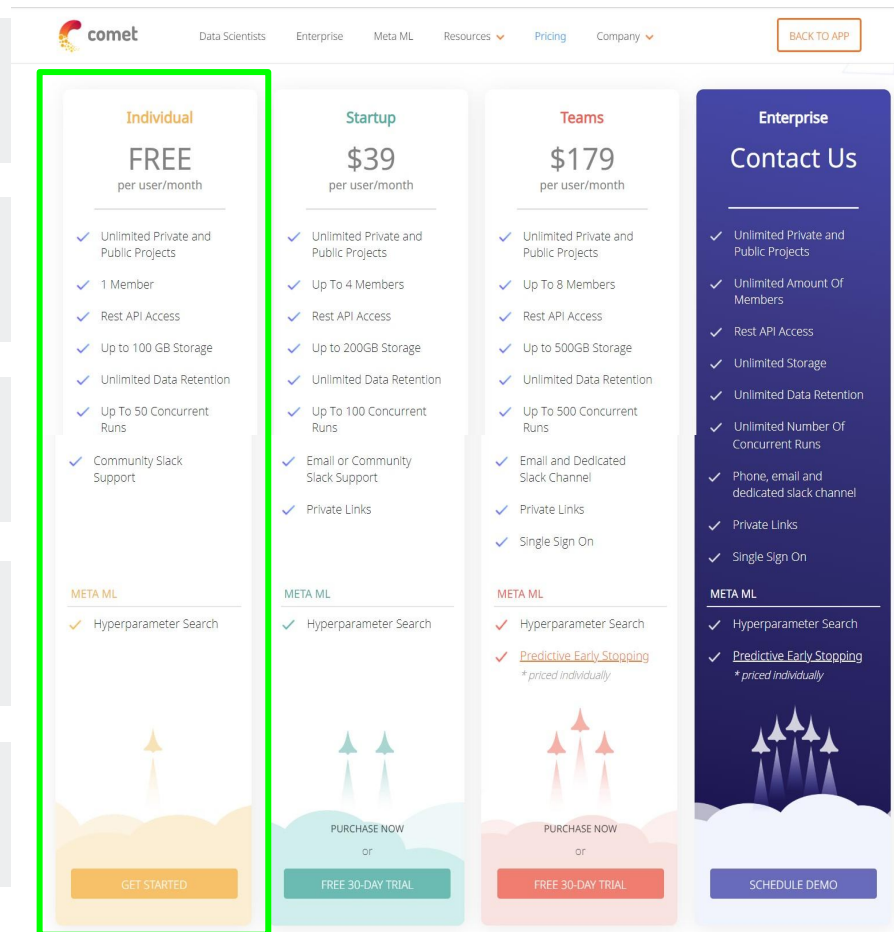
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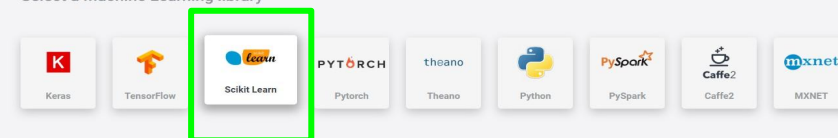
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Get Started with Comet

Select a Machine Learning library

1 Install Comet Using `pip`

```
$ | pip install comet_ml
$ | pip3 install comet_ml
```

2 Add Tracking Code for Scikit Learn

Full Scikit Learn example link [on our Github Quickstart Guide](#)

```
1 # Import comet_ml in the top of your file
2 from comet_ml import Experiment
3
4 experiment = Experiment(api_key="COMET_API_KEY",
5                         project_name="general", workspace="joanne-moonsamy")
6
```

Save this for later

3 Run your experiment

Run your training code just like you normally do (ie. `python train.py`).

4 See data on Comet

In a few moments your data will be transmitted to the experiments page. Your experiment will send data to Comet.

[Listen for my experiment](#)[Skip setup and get started](#)

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```
[1] !pip install comet_ml
```

```
Collecting comet_ml
  Downloading https://files.pythonhosted.org/packages/d3/c3/0e46ea9ccba761f005dbde4fe0bd884308251e1a854987ef610659fc09c/comet_ml-3.1.1-py3-none-any.whl (194kB)
    194kB 1.3MB/s
Collecting comet-git-pure==0.19.11
  Downloading https://files.pythonhosted.org/packages/56/2b/c1ca11a237e3fcb0e1ea53134901f30f3aa657e5ea141dcca0e0f46df0e/comet_git_pure-0.19.11-py3-none-any.whl (409kB)
    409kB 2.5MB/s
Requirement already satisfied: jsonschema<3.1.0,>=2.6.0 in /usr/local/lib/python3.6/dist-packages (from comet_ml) (2.6.0)
Requirement already satisfied: six in /usr/local/lib/python3.6/dist-packages (from comet_ml) (1.12.0)
Collecting websocket-client<0.55.0
  Downloading https://files.pythonhosted.org/packages/4c/5f/f61b420143ed1c8dc69f9eae5ff1ac36109d52c80de49d66e0c36c3dfdf/websocket_cli-0.55.0-py3-none-any.whl (204kB)
    204kB 3.2MB/s
Collecting wurlitizer<1.0.2
  Downloading https://files.pythonhosted.org/packages/24/5e/f3bd8443bdf96d2f5d10097d301076a9eb55637b7864e52d2d1a4d8c72a/wurlitizer-2.0-py3-none-any.whl (10.7kB)
    10.7kB 1.3MB/s
Requirement already satisfied: requests>=2.18.4 in /usr/local/lib/python3.6/dist-packages (from comet_ml) (2.21.0)
Requirement already satisfied: certifi in /usr/local/lib/python3.6/dist-packages (from comet-git-pure==0.19.11->comet_ml) (2019.11.28)
Requirement already satisfied: urllib3>=1.24.1 in /usr/local/lib/python3.6/dist-packages (from comet-git-pure==0.19.11->comet_ml) (1.24.1)
Requirement already satisfied: idna<2.9,>=2.5 in /usr/local/lib/python3.6/dist-packages (from requests>=2.18.4->comet_ml) (2.8)
Requirement already satisfied: chardet<3.1.0,>=3.0.2 in /usr/local/lib/python3.6/dist-packages (from requests>=2.18.4->comet_ml) (3.0.2)
Collecting configobj
  Downloading https://files.pythonhosted.org/packages/64/61/079eb60459c44929e684fa7d9e2fdeca03f67d64dd9bac27296be2e0fab/configobj-5.0.6-py2.py3-none-any.whl (31kB)
    31kB 1.3MB/s
Building wheels for collected packages: configobj
  Building wheel for configobj (setup.py) ... done
  Created wheel for configobj: filename=configobj-5.0.6-cp36-none-any.whl size=34546 sha256=e187785469ed4c714f732dc29e751bf4c4377b5dd7
  Stored in directory: /root/.cache/pip/wheels/f1/e4/16/4981ca97c2d65186b49861e0b35e2660695be7219a2d351ee0
Successfully built configobj
Installing collected packages: comet-git-pure, websocket-client, wurlitizer, netifaces, configobj, everett, comet-ml
Successfully installed comet-git-pure-0.19.11 comet-ml-3.1.1 configobj-5.0.6 everett-1.0.2 netifaces-0.10.9 websocket-client-0.57.0 wur
```

Get ready to start recording your own experiments!



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- **Import comet** at the very top of your notebook

```
[ ] from comet_ml import Experiment
```

You will see an API key button at the top of the page when you click on an experiment workspace to comet. (If a project is empty, the code below will autogenerate for you)

Step 2

- **Link your workspace** using the code from the starter-guide

```
# Setting the API key (saved as environment variable)
experiment = Experiment(#api_key='your api key here'|
                        project_name="general", workspace="jo-moon")
```

COMET INFO: Experiment is live on comet.ml <https://www.comet.ml/jo-moon/ge>

Step 3

- Create/run your notebook as you usually would

Step 4

- After your model has run, **save all your metrics** as separate variables and **create dictionaries** for the data you want to log.

Step 5

- Log the necessary items using the commands: **`experiment.log_metrics()`** and **`experiment.log_parameters()`**

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- Remember to **`experiment.end()`** if you're using a **notebook**

Import the rest of your necessary libraries as you usually would. For this demonstration we will be using the breast cancer classification so we will also import that from sklearn.

```
[ ] import numpy as np
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from sklearn.model_selection import train_test_split, GridSearchCV
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import f1_score, precision_score, recall_score, confusion_matrix
```

```
[ ] # Have a look at your dataset
cancer = load_breast_cancer()
print("cancer.keys(): {}".format(cancer.keys()))
print("Shape of cancer data: {}".format(cancer.data.shape))
print("Sample counts per class:\n{}".format(
    {n: v for n, v in zip(cancer.target_names, np.bincount(cancer.target))}))
print("\nFeature names:\n{}".format(cancer.feature_names))
```

```
cancer.keys(): dict_keys(['data', 'target', 'target_names', 'DESCR', 'feature_names', 'filename'])
Shape of cancer data: (569, 30)
```

```
Sample counts per class:
{'malignant': 212, 'benign': 357}
```

```
Feature names:
['mean radius' 'mean texture' 'mean perimeter' 'mean area'
 'mean smoothness' 'mean compactness' 'mean concavity'
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```
# Saving each metric to add to a dictionary for logging
```

```
f1 = f1_score(y_test, y_pred)
precision = precision_score(y_test, y_pred)
recall = recall_score(y_test, y_pred)
```

```
[ ] # Create dictionaries for the data we want to log
```

```
params = {"random_state": 7,
          "model_type": "logreg",
          "scaler": "standard_scaler",
          "param_grid": str(param_grid),
          "stratify": True
        }
metrics = {"f1": f1,
          "recall": recall,
          "precision": precision
        }
```

```
[ ] # Log our parameters and results
experiment.log_parameters(params)
experiment.log_metrics(metrics)
```

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You can find additional information on experiments and what data you can record at the link below:

<https://www.comet.ml/docs/python-sdk/Experiment/>

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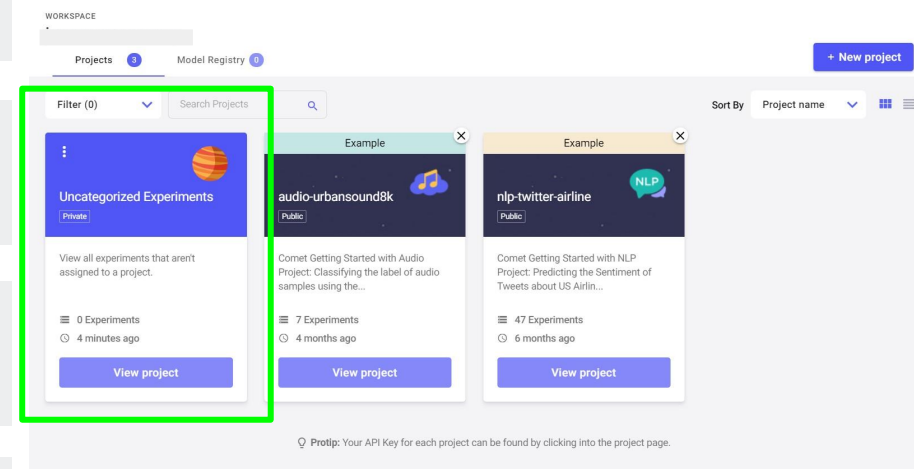
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- Remember to `experiment.end()` if you're using a **notebook**! You can check out your results immediately after by clicking on your project.

```
] experiment.end()
```

```
COMET INFO: -----
COMET INFO: Comet.ml Experiment Summary:
COMET INFO: Data:
COMET INFO:   url: https://www.comet.ml/jo-moon/general/96b60794fd8747a084b2a1c0cc015a33
COMET INFO: Metrics [count] (min, max):
COMET INFO:   f1           : (0.9832402234636872, 0.9832402234636872)
COMET INFO:   precision    : (0.9887640449438202, 0.9887640449438202)
COMET INFO:   recall       : (0.9777777777777777, 0.9777777777777777)
COMET INFO:   sys.cpu.percent.01 [31] : (0.9, 10.3)
COMET INFO:   sys.cpu.percent.02 [31] : (0.9, 7.5)
COMET INFO:   sys.cpu.percent.avg [31]: (0.95, 8.9)
COMET INFO:   sys.ram.total [31]      : (13653561344.0, 13653561344.0)
COMET INFO:   sys.ram.used [31]       : (592932864.0, 735928320.0)
COMET INFO: -----
COMET INFO: Uploading stats to Comet before program termination (may take several seconds)
```



Bonus Panel!

Using Comet in a Kaggle Kernel

You will need to use Comet inside of your Kaggle kernel - This can be done in exactly the same way that we outlined above, Just remember to switch on the internet connectivity in your kernel to install the package. You can do this as illustrated on the right.

You may be prompted to verify you account, and will then be able to proceed :)

```
!pip install comet_ml
```

```
Collecting comet_ml
  Downloading comet_ml-3.1.8-py2.py3-none-any.whl (200 kB)
    |████████████████████████████████████████| 200 kB 4.8 MB/s eta 0:00:01
```

The image shows the right-hand sidebar of a Kaggle kernel interface. It is divided into several sections: 'Data', 'Settings', and 'Code Help'. In the 'Data' section, there are folders for 'input (read-only data)', 'climate-change-belief-analysis', 'output', and '/kaggle/working'. The 'Settings' section contains options for 'Language' (set to Python), 'Environment' (set to Preferences), 'Accelerator' (set to None), and 'Internet'. The 'Internet' option is highlighted with a green rectangular box and shows a toggle switch that is currently turned 'On'.



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More info at comet.ml

If you would like to dig deeper, go and explore the tutorials on the comet website. They are robust, and insightful, and will give you a lot to play with.

This is not the only method of model version control- there are many amazing new technologies springing up specifically for Data Science applications, whereas in the past we were adapting tools created for software engineers. So go and explore, and do amazing things!

