Module 2 Setting up Machine Learning Environment

























Class Modules

- Module 1- Introduction to Machine Learning
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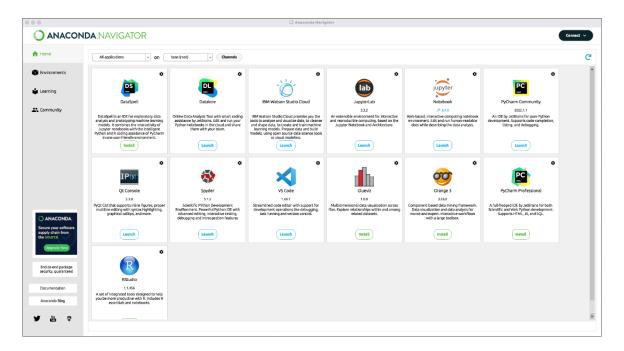
Install







- Anaconda is a distribution of the Python and R programming languages for scientific computing, that aims to simplify package management with conda environments.
- Anaconda offers the easiest way to perform data science and machine learning on a single machine.
- Install Anaconda @ https://www.anaconda.com/products/distribution









JupyterLab



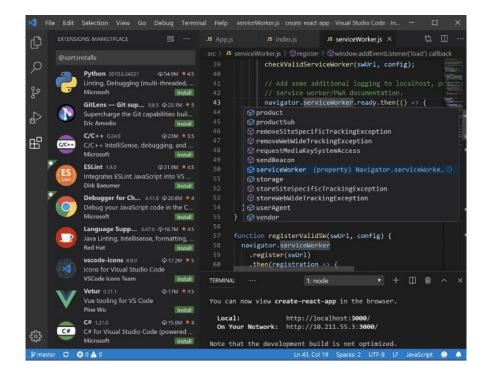
- <u>JupyterLab</u> is the latest web-based interactive development environment for notebooks, code, and data
- Jupyter's name is a reference to the three core programming languages supported by Jupyter, which are Julia, Python and R



Pedram, Jahangiry



- VS Code is one of the most popular source code editors
- Features include support for debugging, syntax highlighting, intelligent code completion, code refactoring, and embedded Git.
- Install VS code @ https://code.visualstudio.com/



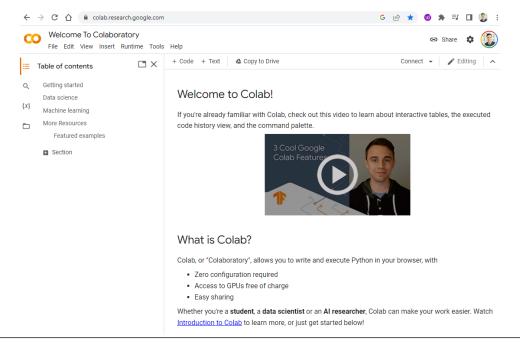








- <u>Colab</u> is a free hosted Jupyter notebook-style environment that runs entirely in the cloud and requires no setup to use. It also provides access to machine learning libraries and computing resources including GPU.
- Colab allows anybody to write and execute arbitrary python code through the browser, and is especially well suited to machine learning, data analysis and education. https://colab.research.google.com/







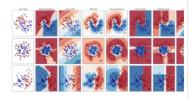




Classification

Identifying which category an object belongs to.

Applications: Spam detection, image recognition. **Algorithms:** SVM, nearest neighbors, random forest, and more...

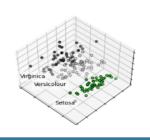


Examples

Dimensionality reduction

Reducing the number of random variables to consider.

Applications: Visualization, Increased efficiency Algorithms: PCA, feature selection, non-negative matrix factorization, and more...

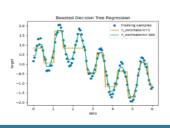


Example

Regression Predicting a contin

Predicting a continuous-valued attribute associated with an object.

Applications: Drug response, Stock prices. **Algorithms:** SVR, nearest neighbors, random forest, and more...



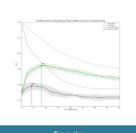
Examples

Model selection

Comparing, validating and choosing parameters and models.

Applications: Improved accuracy via parameter tuning

Algorithms: grid search, cross validation, metrics, and more...



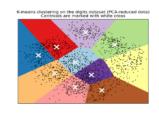
Examples

Clustering

Automatic grouping of similar objects into sets.

Applications: Customer segmentation, Grouping experiment outcomes

Algorithms: k-Means, spectral clustering, meanshift, and more...

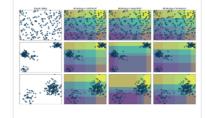


Examples

Preprocessing

Feature extraction and normalization.

Applications: Transforming Input data such as text for use with machine learning algorithms. Algorithms: preprocessing, feature extraction, and



Examples

- Scikit-learn is an open-sourced Python library and includes a variety of unsupervised and supervised learning techniques.
- It is based on technologies and libraries like Matplotlib, Pandas and NumPy and helps simplify the coding task.
- Install Scikit-learn @ https://scikit-learn.org/stable/install.html









- PyCaret is an open-source, low-code machine learning library in Python that automates machine learning workflows.
- PyCaret is essentially a Python wrapper around several machine learning libraries and frameworks
- Install PyCaret @ https://pycaret.gitbook.io/docs/get-started/installation

```
# load dataset
import pandas as pd
train = pd.read_csv('train.csv')
test = pd.read_csv('test.csv')

# init setup
from pycaret.classification import *
s = setup(train, target= 'target')

# model training and selection
best = compare_models()

# analyze best model
evaluate_model(best)

# predict on new data
predictions = predict_model(best, data =test )

# save best pipeline
save_model(best, 'my_best_pipeline')
```

```
'lr'
             Logistic Regression
             K Nearest Neighbour
'knn'
             Naives Bayes
'nb'
'dt'
             Decision Tree Classifier
             SVM - Linear Kernel
'svm'
             SVM - Radial Kernel
'rbfsvm'
'gpc'
             Gaussian Process Classifier
'mlp'
             Multi Level Perceptron
             Ridge Classifier
'ridge'
'rf'
             Random Forest Classifier
'qda'
             Quadratic Discriminant Analysis
             Ada Boost Classifier
'ada'
'gbc'
             Gradient Boosting Classifier
'lda'
             Linear Discriminant Analysis
'et'
             Extra Trees Classifier
'xgboost'
             Extreme Gradient Boosting
             Light Gradient Boosting
'lightgbm'
'catboost'
            CatBoost Classifier
```





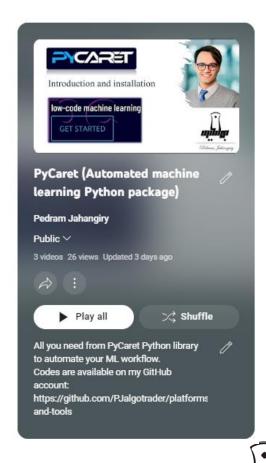


Available YouTube playlists









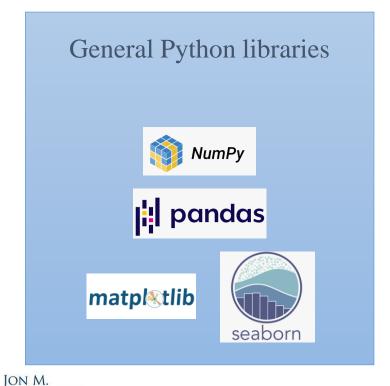




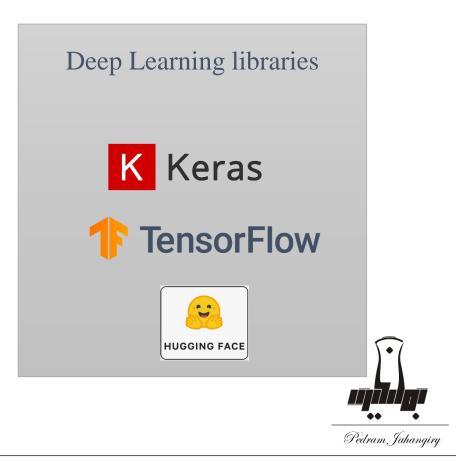
Platforms and Packages



Listed below are some Python packages and platforms that will be used in the Machine learning deep learning and deep forecasting courses.









Setting up Deep Learning Environment







Personal Workstation

Cloud Platforms

Google Colaboratory

Pros

- Full control over hardware and software
- Work offline
- Fixed cost

- Powerful computing resources
- Scalability
- Ease of use
- Cost-effective: Pay-as-you-go
- Collaboration

- Powerful computing resources (GPU, TPU)
- Ease of use
- Collaboration
- No need to set up a local environment

Cons

- Scalability
- Maintenance (both hardware and software)
- Expensive for large-scale experiments
- Dependency on the provider
- Limited control
- Internet connection
- Security

- Time limit
- Hardware limitation
- Data storage
- Limited control
- Internet connection
- Security







Kaggle Survey 2022



- Kaggle runs a yearly survey among machine learning and data science professionals worldwide.
- This survey is one of our most reliable sources about the state of the industry!!!

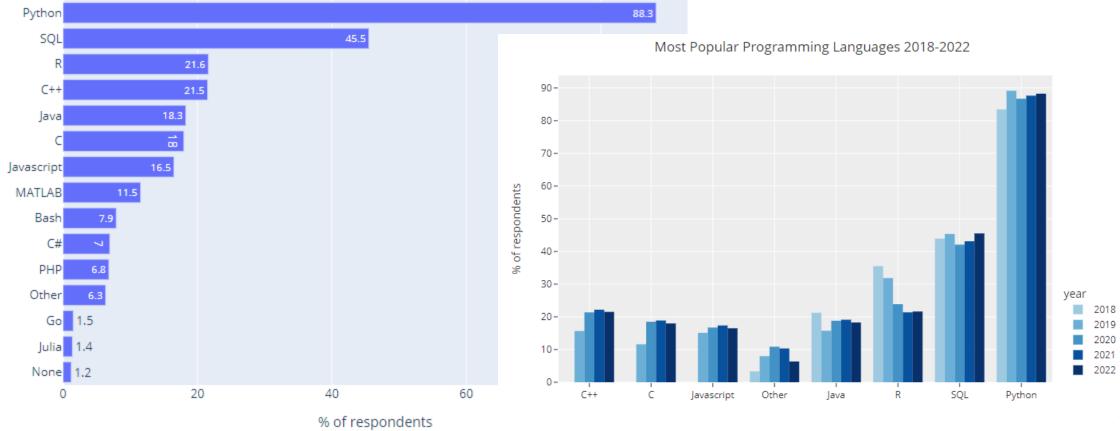






Programming Languages

Most Popular Programming Languages in 2022

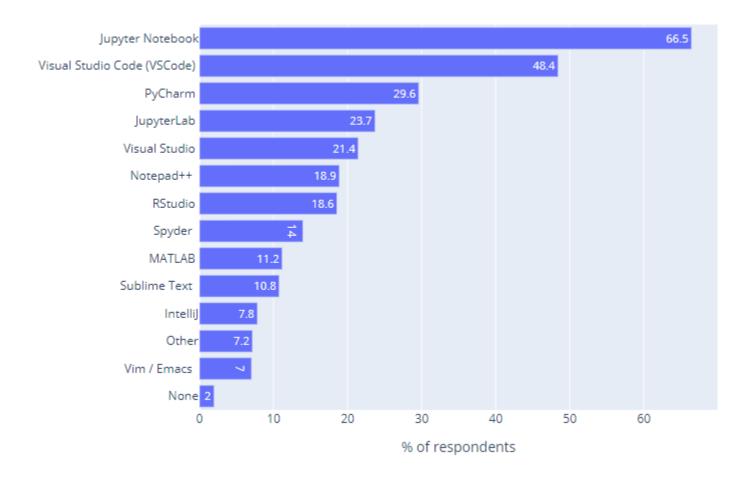






Popular IDE's

Most Popular IDE's in 2022



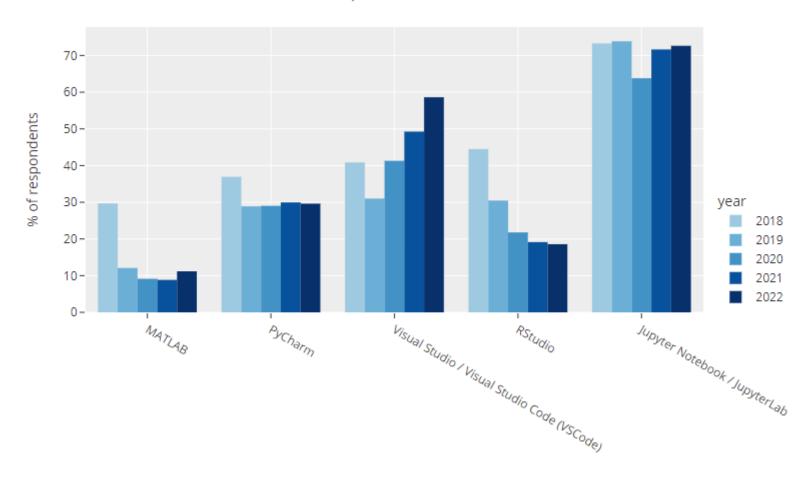






Popular IDE's





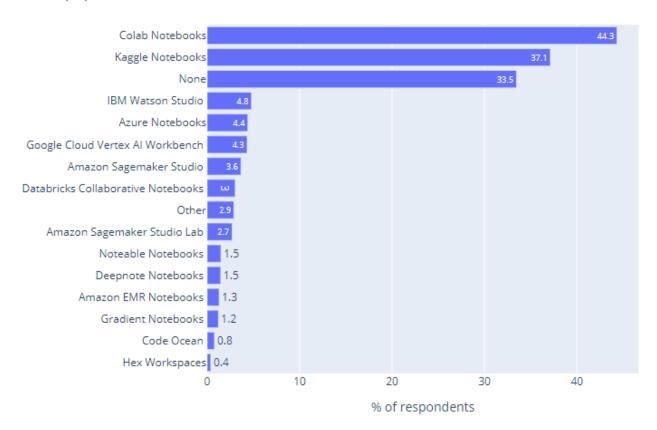


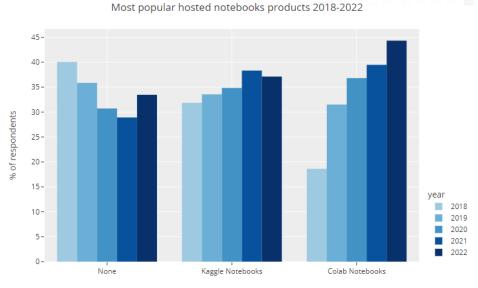




Popular notebooks

Most popular hosted notebooks in 2022 in 2022





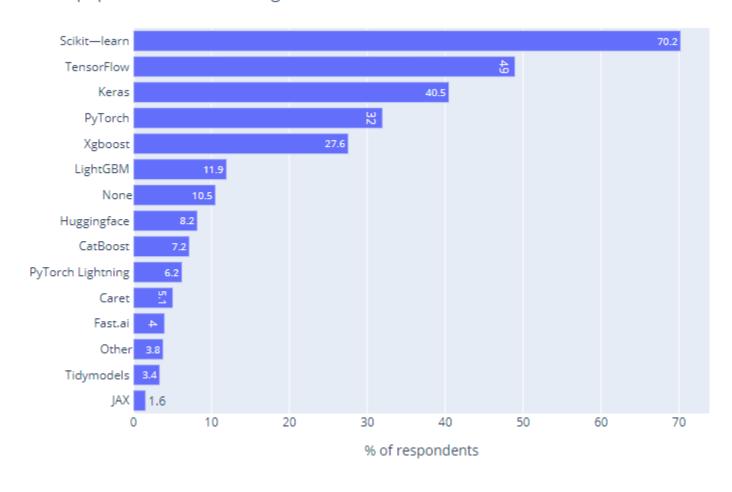






Machine Learning Frameworks

Most popular machine learning frameworks in 2022



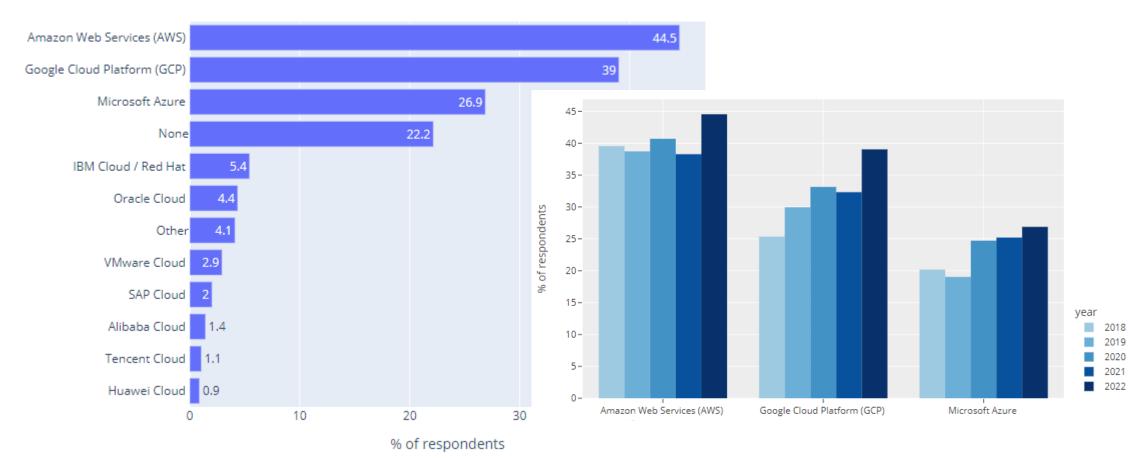






Cloud computing platforms

Most popular cloud computing platforms in 2022







The modern machine learning landscape

- From 2016 to 2022, the entire machine learning and data science industry has been dominated by these two approaches:
 - 1. Deep learning
 - 2. Gradient boosted trees
- Most practitioners of deep learning use Keras, often in combination with its parent framework TensorFlow.
- This means you'll need to be familiar with Scikit-learn, XGBoost, and Keras

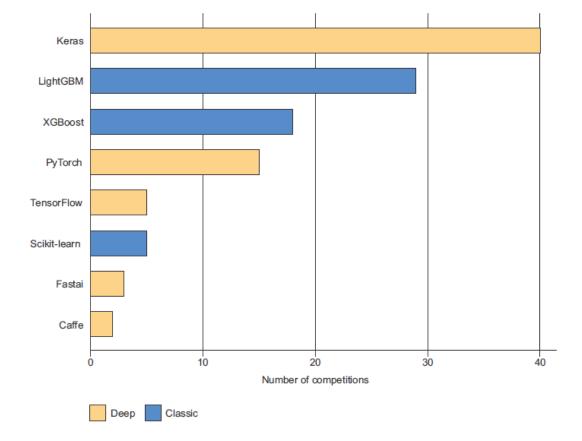


Figure 1.12 Machine learning tools used by top teams on Kaggle







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