

# Module 2 – Part I

## Setting up Deep Forecasting Environment (Python)

---





# Road map!

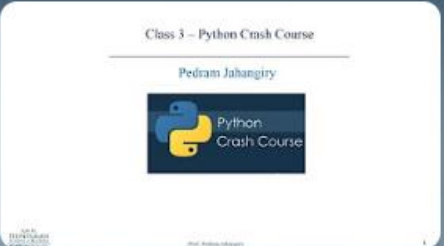
- Module 1- Demystifying Timeseries Data and Modeling
- **Module 2- Setting up Deep Forecasting Environment (Python)**
- Module 3- Exponential Smoothing
- Module 4- ARIMA models
- Module 5- Machine Learning for Time series Forecasting
- Module 6- Deep Neural Networks
- Module 7- Deep Sequence Modeling (RNN, LSTM)
- Module 8- Prophet and Neural Prophet



# Course prerequisites

Class 3 – Python Crash Course

Pedram Jahangiry



**Python Crash Course**

Pedram Jahangiry

Public


9 videos 1,612 views Updated today

 Play all  Shuffle

Codes are available on my GitHub account:  
<https://github.com/PJalgotrader/platforms-and-tools>

1. vscode Installation







**Visual Studio Code**

Pedram Jahangiry

Public






13 videos 194 views Updated 3 days ago

 Play all  Shuffle

Codes are available on my GitHub account:  
<https://github.com/PJalgotrader/platforms-and-tools>

1. Google Colab: Jumpstart!







**Google Colab**

Pedram Jahangiry


Public

3 videos 73 views Updated 6 days ago

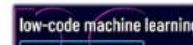
 

 Play all  Shuffle

Codes are available on my GitHub account:  
<https://github.com/PJalgotrader/platforms-and-tools>



Introduction and installation







**PyCaret (Automated machine learning Python package)**

Pedram Jahangiry

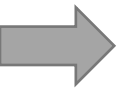
Public

3 videos 26 views Updated 3 days ago

 Play all  Shuffle

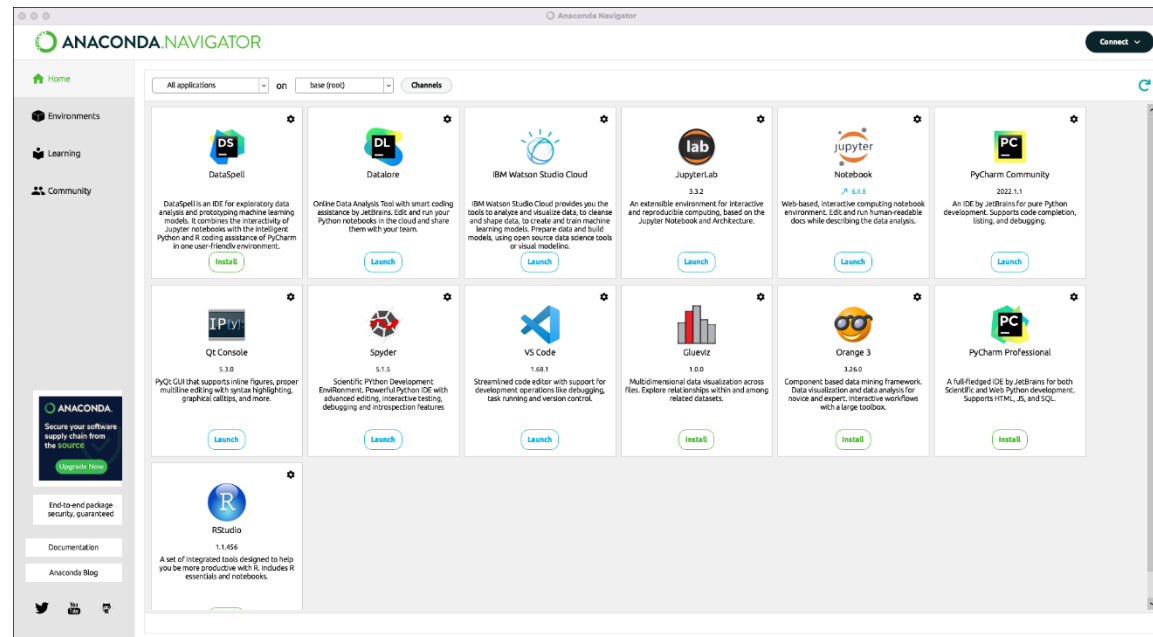
All you need from PyCaret Python library to automate your ML workflow.  
Codes are available on my GitHub account:  
<https://github.com/PJalgotrader/platforms-and-tools>

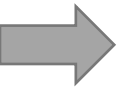


# Install through



- 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/>

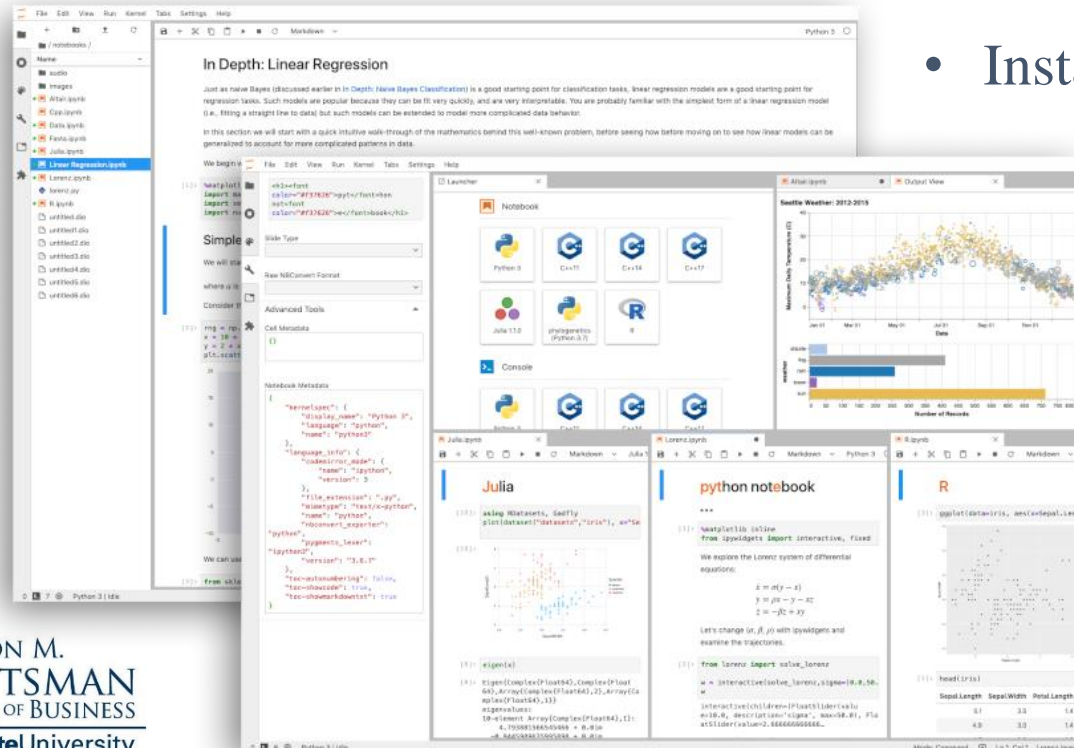




# JupyterLab



- JupyterLab 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**

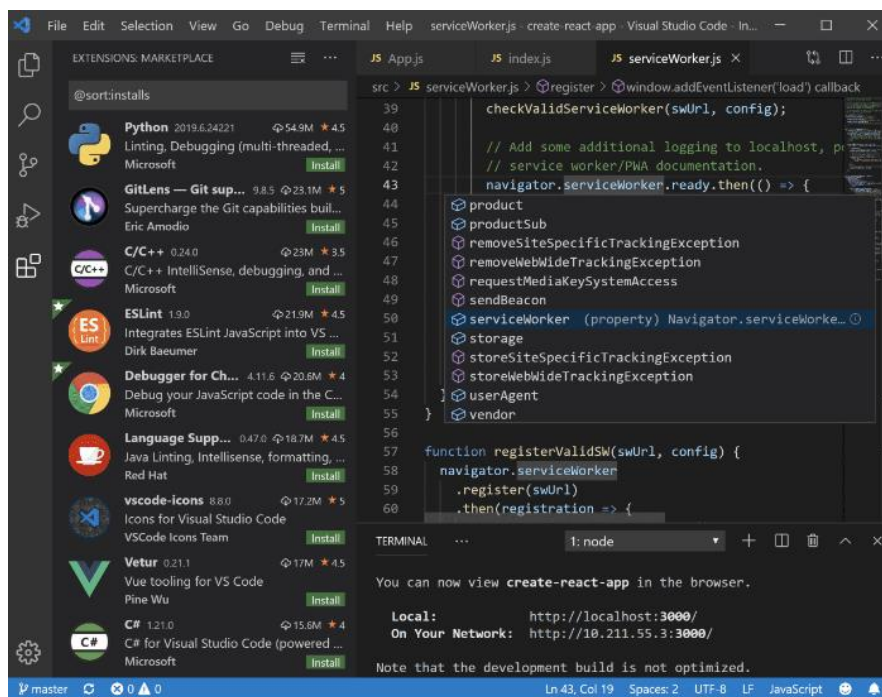


- Install JupyterLab @ <http://jupyter.org/install>

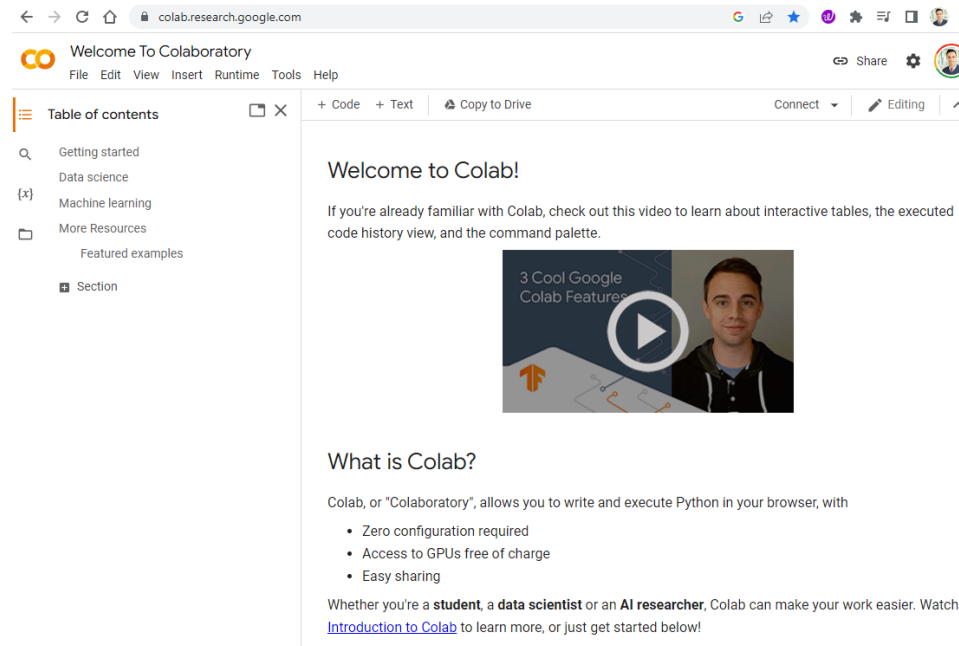


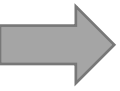
# VS Code

- 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/>



- Colab 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/>



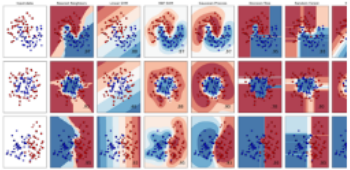


### Classification

Identifying which category an object belongs to.

**Applications:** Spam detection, image recognition.

**Algorithms:** SVM, nearest neighbors, random forest, and more...



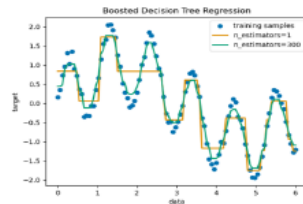
Examples

### Regression

Predicting a continuous-valued attribute associated with an object.

**Applications:** Drug response, Stock prices.

**Algorithms:** SVR, nearest neighbors, random forest, and more...



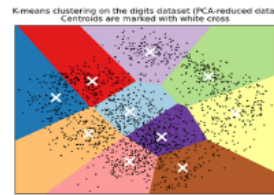
Examples

### Clustering

Automatic grouping of similar objects into sets.

**Applications:** Customer segmentation, Grouping experiment outcomes

**Algorithms:** k-Means, spectral clustering, mean-shift, and more...



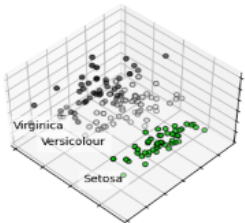
Examples

### Dimensionality reduction

Reducing the number of random variables to consider.

**Applications:** Visualization, Increased efficiency

**Algorithms:** PCA, feature selection, non-negative matrix factorization, and more...



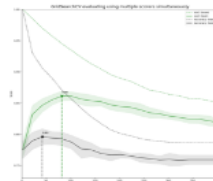
Examples

### Model selection

Comparing, validating and choosing parameters and models.

**Applications:** Improved accuracy via parameter tuning

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



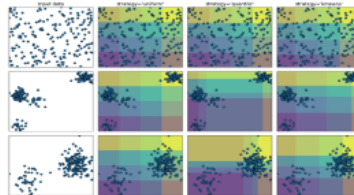
Examples

### Preprocessing

Feature extraction and normalization.

**Applications:** Transforming input data such as text for use with machine learning algorithms.

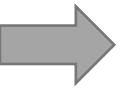
**Algorithms:** preprocessing, feature extraction, and more...



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/statistical learning library in Python that automates machine learning workflows.
- PyCaret is essentially a **Python wrapper** around several machine/statistical learning libraries and frameworks
- Install PyCaret @ <https://pycaret.gitbook.io/docs/get-started/installation>

```
from pycaret.time_series import *  
  
exp = TSForecastingExperiment()  
exp.setup(data = df, target='Passengers' , fh = 12, coverage=0.95)  
  
# comparing all models  
best = exp.compare_models()
```



- List of **Timeseries models** available in PyCaret
- This is the [Full] version.
- `pip install pycaret[full]`

```
from pycaret.time_series import *  
  
exp = TSForecastingExperiment()  
exp.setup(data = df, target='Passengers')  
  
# comparing all models  
best = exp.compare_models()
```

ID	Description
naive	Naive Forecaster
grand_means	Grand Means Forecaster
snaive	Seasonal Naive Forecaster
polytrend	Polynomial Trend Forecaster
arima	ARIMA (Autoregressive Integrated Moving Average)
auto_arima	Auto ARIMA
exp_smooth	Exponential Smoothing
ets	ETS (Error, Trend, Seasonality)
theta	Theta Forecaster
stlf	STLF (Seasonal and Trend decomposition using Loess)
croston	Croston's method for intermittent demand forecasting
lr_cds_dt	Linear Regression with Conditional Deseasonalizing and Detrending
en_cds_dt	Elastic Net with Conditional Deseasonalizing and Detrending
ridge_cds_dt	Ridge Regression with Conditional Deseasonalizing and Detrending
lasso_cds_dt	Lasso Regression with Conditional Deseasonalizing and Detrending
llar_cds_dt	Lasso Least Angular Regressor with Cond. Deseasonalizing and Detrending
br_cds_dt	Bayesian Ridge with Cond. Deseasonalizing and Detrending
huber_cds_dt	Huber Regressor with Cond. Deseasonalizing and Detrending
omp_cds_dt	Orthogonal Matching Pursuit with Cond. Deseasonalizing and Detrending
knn_cds_dt	K Neighbors Regressor with Cond. Deseasonalizing and Detrending
dt_cds_dt	Decision Tree Regressor with Cond. Deseasonalizing and Detrending
rf_cds_dt	Random Forest Regressor with Cond. Deseasonalizing and Detrending
et_cds_dt	Extra Trees Regressor with Cond. Deseasonalizing and Detrending
gbr_cds_dt	Gradient Boosting Regressor with Cond. Deseasonalizing and Detrending
ada_cds_dt	AdaBoost Regressor with Cond. Deseasonalizing and Detrending
xgboost_cds_dt	XGBoost Regressor with Cond. Deseasonalizing and Detrending
lightgbm_cds_dt	LightGBM Regressor with Cond. Deseasonalizing and Detrending
catboost_cds_dt	CatBoost Regressor with Cond. Deseasonalizing and Detrending



# K Keras

- Keras is a **high-level**, open-source **neural network** library written in Python. It was developed to make it easier for researchers and developers to build and experiment with deep learning models.
- The Keras API became the official high-level API for TensorFlow 2.0 in **2019**.  
<https://keras.io/>

```
model = keras.Sequential(  
    [  
        keras.Input(shape=input_shape),  
        layers.Conv2D(32, kernel_size=(3, 3), activation="relu"),  
        layers.MaxPooling2D(pool_size=(2, 2)),  
        layers.Conv2D(64, kernel_size=(3, 3), activation="relu"),  
        layers.MaxPooling2D(pool_size=(2, 2)),  
        layers.Flatten(),  
        layers.Dropout(0.5),  
        layers.Dense(num_classes, activation="softmax"),  
    ]  
)  
  
model.summary()
```

# → Platforms and Packages

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



## General Python libraries



## Machine Learning libraries



## Deep Learning libraries





# Setting up Deep Learning Environment



Personal Workstation



Cloud Platforms



Google Colaboratory

## Pros

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

## Cons

- Scalability
- Maintenance (both hardware and software)

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

- Expensive for large-scale experiments
- Dependency on the provider
- Limited control
- Internet connection
- Security

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

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

# ➔ Kaggle Survey 2022

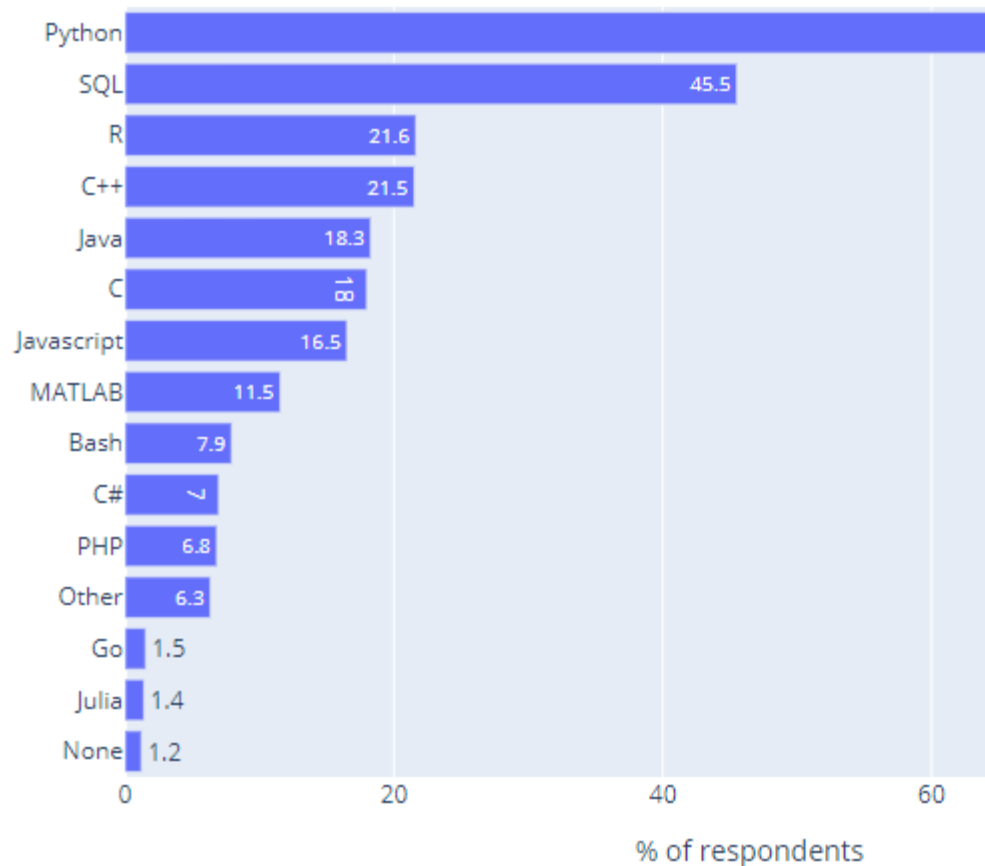


- Kaggle runs a yearly survey among machine learning and data science professionals worldwide.
- This survey is one of our **most reliable** sources about the **state of the industry!!!**

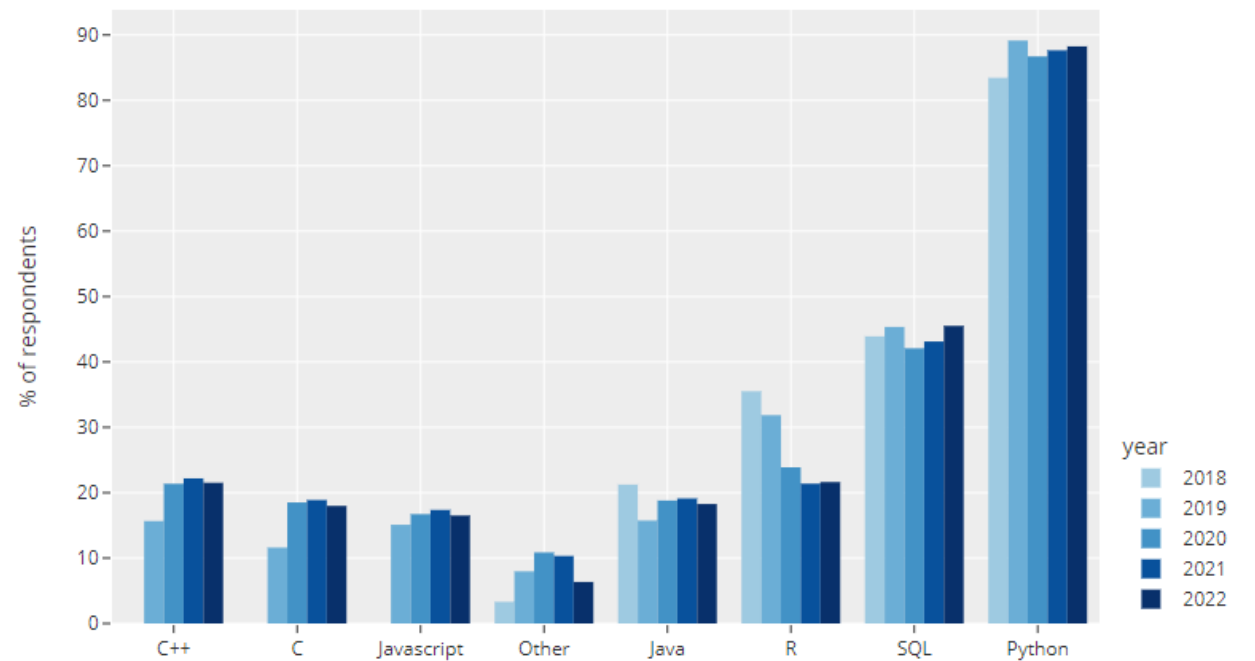


# Programming Languages

Most Popular Programming Languages in 2022



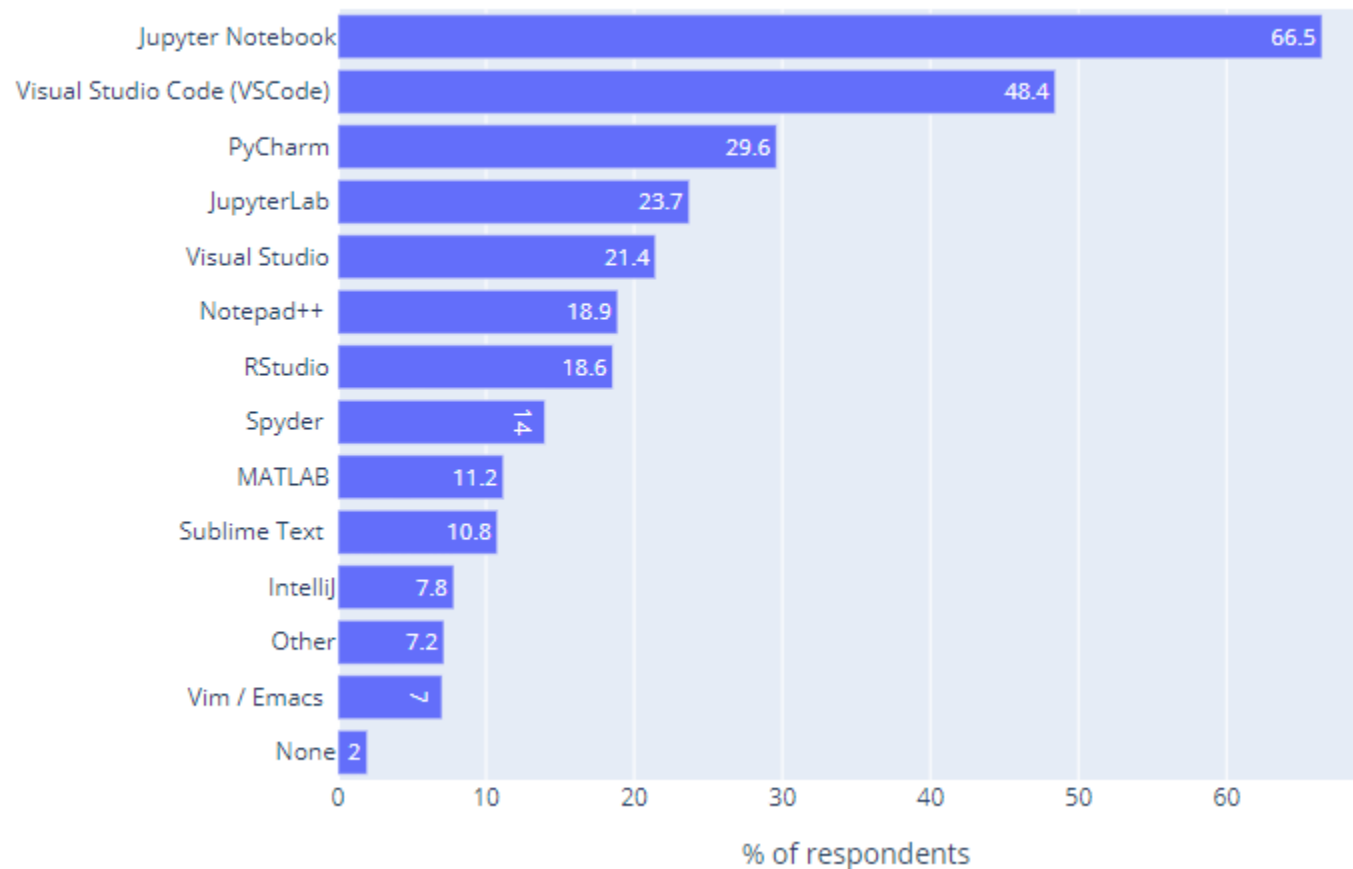
Most Popular Programming Languages 2018-2022





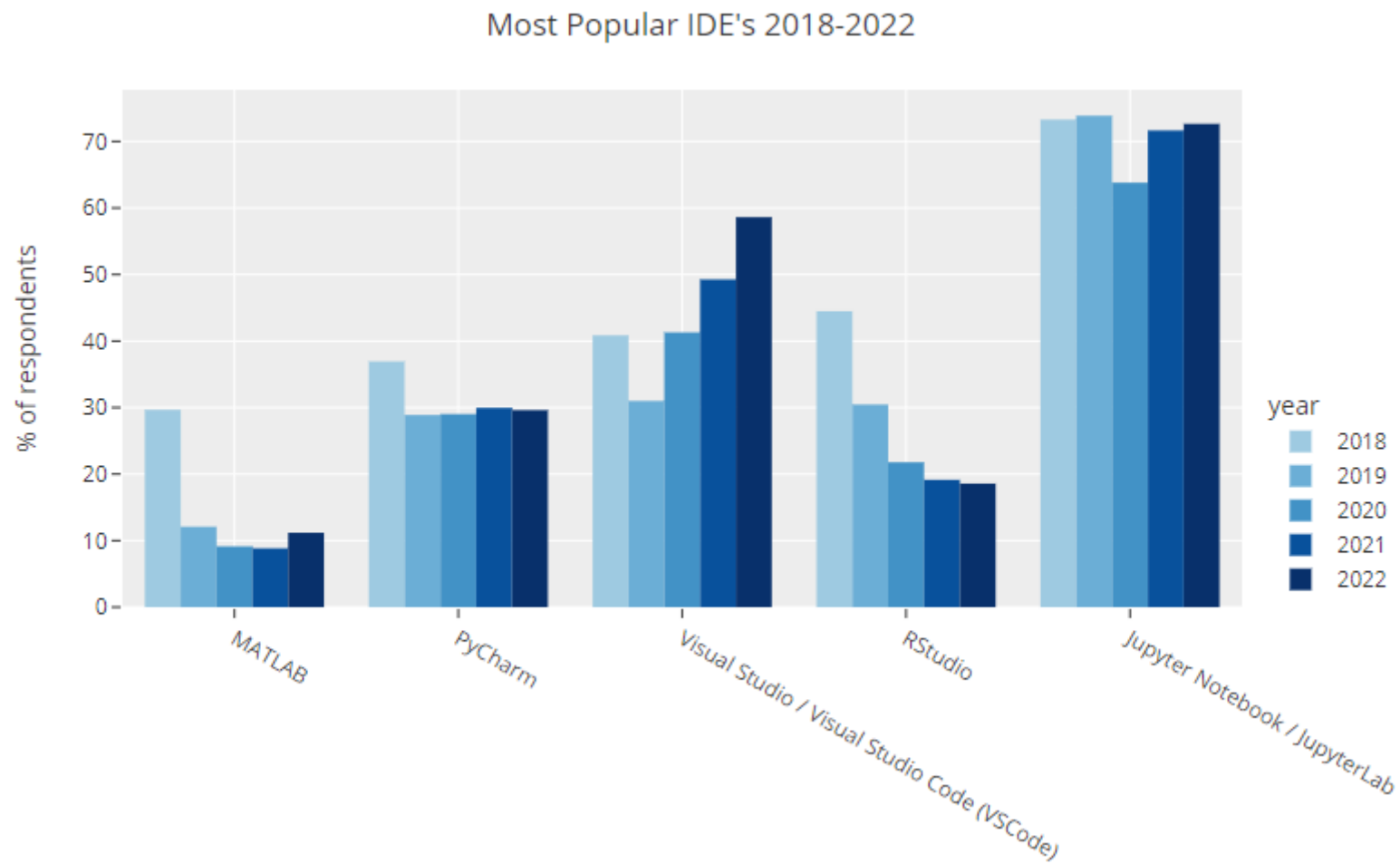
# Popular IDE's

Most Popular IDE's in 2022












# Popular IDE's





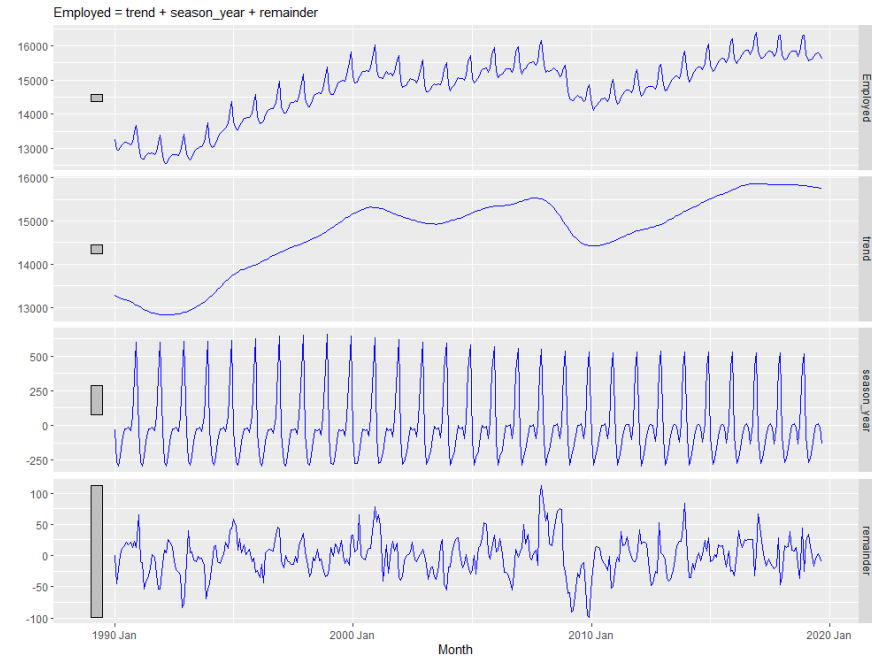
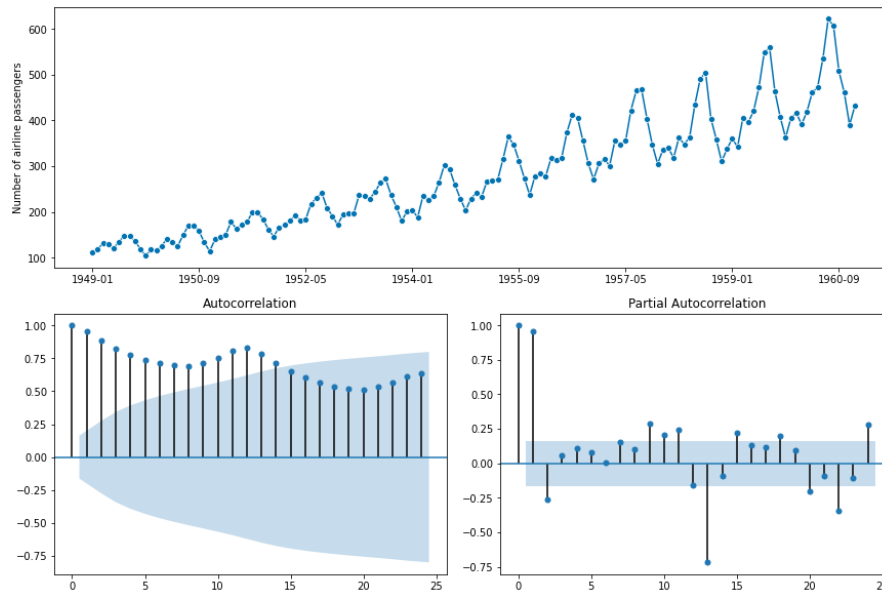
# Python packages

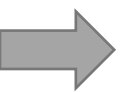
Package	Metrics	ML	DL	Focus
 statsmodels	✓✓✓			Statistics, Econometrics
 scikit-learn	✓	✓✓✓		General Machine Learning
 Keras			✓✓✓	General Deep Learning
 PYCARET	✓✓✓	✓✓✓		Auto ML, Rapid prototyping, Comparison
 Darts	✓✓✓	✓	✓✓	Advanced timeseries and forecasting
 GLUONS	✓	✓	✓✓	Large Scale probabilistic Models
 NIXTLA	✓✓✓	✓✓✓	✓✓✓	Workflows, SOTA methods



# Module 2 – Part II

## Setting up Deep Forecasting Environment (Time Series Basics in Python)





# Road map!

- Module 1- Demystifying Timeseries Data and Modeling
- **Module 2- Setting up Deep Forecasting Environment (Python)**
- Module 3- Exponential Smoothing
- Module 4- ARIMA models
- Module 5- Machine Learning for Time series Forecasting
- Module 6- Deep Neural Networks
- Module 7- Deep Sequence Modeling (RNN, LSTM)
- Module 8- Prophet and Neural Prophet





# ➔ Road map!

- ✓ Module 1- Demystifying Timeseries Data and Modeling
- ✓ Module 2- Setting up Deep Forecasting Environment (Python)
- Module 3- Exponential Smoothing
- Module 4- ARIMA models
- Module 5- Machine Learning for Time series Forecasting
- Module 6- Deep Neural Networks
- Module 7- Deep Sequence Modeling (RNN, LSTM)
- Module 8- Prophet and Neural Prophet

