If you're already familiar with Colab, check out this video to learn about interactive tables, the executed code history view, and the command palette.

#task 1  
import pandas as pd  
import numpy as np  
s1=pd.Series(np.random.rand(4))  
s2=pd.Series(np.random.rand(4))  
s3=pd.Series(np.random.rand(4))  
s4=pd.Series(np.random.rand(4))  
print(s1,s2,s3,s4)  
df=pd.DataFrame([s1,s2,s3,s4])

0 0.413989  
1 0.053215  
2 0.273290  
3 0.520557  
dtype: float64 0 0.620287  
1 0.726743  
2 0.575726  
3 0.077099  
dtype: float64 0 0.589224  
1 0.181918  
2 0.790616  
3 0.100420  
dtype: float64 0 0.614919  
1 0.965879  
2 0.937190  
3 0.745495  
dtype: float64

0 1 2 3  
0 0.413989 0.053215 0.273290 0.520557  
1 0.620287 0.726743 0.575726 0.077099  
2 0.589224 0.181918 0.790616 0.100420  
3 0.614919 0.965879 0.937190 0.745495

#task 2  
df.columns=['Random value 1','Random value 2','Random value 3','Random value 4']  
df

Random value 1 Random value 2 Random value 3 Random value 4  
0 0.413989 0.053215 0.273290 0.520557  
1 0.620287 0.726743 0.575726 0.077099  
2 0.589224 0.181918 0.790616 0.100420  
3 0.614919 0.965879 0.937190 0.745495

#task 3  
df.describe() #numeric

Random value 1 Random value 2 Random value 3 Random value 4  
count 4.000000 4.000000 4.000000 4.000000  
mean 0.559605 0.481939 0.644206 0.360893  
std 0.098019 0.435103 0.288412 0.327514  
min 0.413989 0.053215 0.273290 0.077099  
25% 0.545416 0.149743 0.500117 0.094590  
50% 0.602072 0.454331 0.683171 0.310488  
75% 0.616261 0.786527 0.827260 0.576791  
max 0.620287 0.965879 0.937190 0.745495

#task 3  
df.describe(include='all') #string

Random value 1 Random value 2 Random value 3 Random value 4  
count 4.000000 4.000000 4.000000 4.000000  
mean 0.559605 0.481939 0.644206 0.360893  
std 0.098019 0.435103 0.288412 0.327514  
min 0.413989 0.053215 0.273290 0.077099  
25% 0.545416 0.149743 0.500117 0.094590  
50% 0.602072 0.454331 0.683171 0.310488  
75% 0.616261 0.786527 0.827260 0.576791  
max 0.620287 0.965879 0.937190 0.745495

#task 4  
df.isnull().sum() #numeric

Random value 1 0  
Random value 2 0  
Random value 3 0  
Random value 4 0  
dtype: int64

#task 4  
df.isnull()#boolean

Random value 1 Random value 2 Random value 3 Random value 4  
0 False False False False  
1 False False False False  
2 False False False False  
3 False False False False

#task 4  
print('column1 datatype is',df['Random value 1'].dtype)  
print('column2 datatype is',df['Random value 2'].dtype)  
print('column3 datatype is',df['Random value 3'].dtype)  
print('column4 datatype is',df['Random value 4'].dtype)

column1 datatype is float64  
column2 datatype is float64  
column3 datatype is float64  
column4 datatype is float64

#task 5  
df.loc[0:4,'Random value 2':'Random value 3'] #location method

Random value 2 Random value 3  
0 0.053215 0.273290  
1 0.726743 0.575726  
2 0.181918 0.790616  
3 0.965879 0.937190

#task 5  
df.iloc[0:,1:3] #index location(0:index location elements,1:3 column range)

Random value 2 Random value 3  
0 0.053215 0.273290  
1 0.726743 0.575726  
2 0.181918 0.790616  
3 0.965879 0.937190

Colab, or "Colaboratory", allows you to write and execute Python in your browser, with

* Zero configuration required
* Access to GPUs free of charge
* Easy sharing

Whether you're a **student**, a **data scientist** or an **AI researcher**, Colab can make your work easier. Watch [Introduction to Colab](https://www.youtube.com/watch?v=inN8seMm7UI) to learn more, or just get started below!

The document you are reading is not a static web page, but an interactive environment called a **Colab notebook** that lets you write and execute code.

For example, here is a **code cell** with a short Python script that computes a value, stores it in a variable, and prints the result:

seconds\_in\_a\_day = 24 \* 60 \* 60  
seconds\_in\_a\_day

86400

To execute the code in the above cell, select it with a click and then either press the play button to the left of the code, or use the keyboard shortcut "Command/Ctrl+Enter". To edit the code, just click the cell and start editing.

Variables that you define in one cell can later be used in other cells:

seconds\_in\_a\_week = 7 \* seconds\_in\_a\_day  
seconds\_in\_a\_week

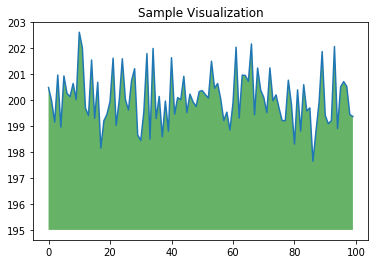
604800

Colab notebooks allow you to combine **executable code** and **rich text** in a single document, along with **images**, **HTML**, **LaTeX** and more. When you create your own Colab notebooks, they are stored in your Google Drive account. You can easily share your Colab notebooks with co-workers or friends, allowing them to comment on your notebooks or even edit them. To learn more, see [Overview of Colab](/notebooks/basic_features_overview.ipynb). To create a new Colab notebook you can use the File menu above, or use the following link: [create a new Colab notebook](http://colab.research.google.com#create=true).

Colab notebooks are Jupyter notebooks that are hosted by Colab. To learn more about the Jupyter project, see [jupyter.org](https://www.jupyter.org).

With Colab you can harness the full power of popular Python libraries to analyze and visualize data. The code cell below uses **numpy** to generate some random data, and uses **matplotlib** to visualize it. To edit the code, just click the cell and start editing.

import numpy as np  
from matplotlib import pyplot as plt  
  
ys = 200 + np.random.randn(100)  
x = [x for x in range(len(ys))]  
  
plt.plot(x, ys, '-')  
plt.fill\_between(x, ys, 195, where=(ys > 195), facecolor='g', alpha=0.6)  
  
plt.title("Sample Visualization")  
plt.show()



You can import your own data into Colab notebooks from your Google Drive account, including from spreadsheets, as well as from Github and many other sources. To learn more about importing data, and how Colab can be used for data science, see the links below under [Working with Data](#working-with-data).

With Colab you can import an image dataset, train an image classifier on it, and evaluate the model, all in just [a few lines of code](https://colab.research.google.com/github/tensorflow/docs/blob/master/site/en/tutorials/quickstart/beginner.ipynb). Colab notebooks execute code on Google's cloud servers, meaning you can leverage the power of Google hardware, including [GPUs and TPUs](#using-accelerated-hardware), regardless of the power of your machine. All you need is a browser.

Colab is used extensively in the machine learning community with applications including:

* Getting started with TensorFlow
* Developing and training neural networks
* Experimenting with TPUs
* Disseminating AI research
* Creating tutorials

To see sample Colab notebooks that demonstrate machine learning applications, see the [machine learning examples](#machine-learning-examples) below.

* [Overview of Colaboratory](/notebooks/basic_features_overview.ipynb)
* [Guide to Markdown](/notebooks/markdown_guide.ipynb)
* [Importing libraries and installing dependencies](/notebooks/snippets/importing_libraries.ipynb)
* [Saving and loading notebooks in GitHub](https://colab.research.google.com/github/googlecolab/colabtools/blob/main/notebooks/colab-github-demo.ipynb)
* [Interactive forms](/notebooks/forms.ipynb)
* [Interactive widgets](/notebooks/widgets.ipynb)
* [Loading data: Drive, Sheets, and Google Cloud Storage](/notebooks/io.ipynb)
* [Charts: visualizing data](/notebooks/charts.ipynb)
* [Getting started with BigQuery](/notebooks/bigquery.ipynb)

### Machine Learning Crash Course

These are a few of the notebooks from Google's online Machine Learning course. See the [full course website](https://developers.google.com/machine-learning/crash-course/) for more.

* [Intro to Pandas DataFrame](https://colab.research.google.com/github/google/eng-edu/blob/main/ml/cc/exercises/pandas_dataframe_ultraquick_tutorial.ipynb)
* [Linear regression with tf.keras using synthetic data](https://colab.research.google.com/github/google/eng-edu/blob/main/ml/cc/exercises/linear_regression_with_synthetic_data.ipynb)
* [TensorFlow with GPUs](/notebooks/gpu.ipynb)
* [TensorFlow with TPUs](/notebooks/tpu.ipynb)
* [NeMo Voice Swap](https://colab.research.google.com/github/NVIDIA/NeMo/blob/stable/tutorials/VoiceSwapSample.ipynb): Use Nvidia's NeMo conversational AI Toolkit to swap a voice in an audio fragment with a computer generated one.
* [Retraining an Image Classifier](https://tensorflow.org/hub/tutorials/tf2_image_retraining): Build a Keras model on top of a pre-trained image classifier to distinguish flowers.
* [Text Classification](https://tensorflow.org/hub/tutorials/tf2_text_classification): Classify IMDB movie reviews as either *positive* or *negative*.
* [Style Transfer](https://tensorflow.org/hub/tutorials/tf2_arbitrary_image_stylization): Use deep learning to transfer style between images.
* [Multilingual Universal Sentence Encoder Q&A](https://tensorflow.org/hub/tutorials/retrieval_with_tf_hub_universal_encoder_qa): Use a machine learning model to answer questions from the SQuAD dataset.
* [Video Interpolation](https://tensorflow.org/hub/tutorials/tweening_conv3d): Predict what happened in a video between the first and the last frame.