

# Welcome to Colab!

## (New) Try the Gemini API

- [Generate a Gemini API key](#)
- [Talk to Gemini with the Speech-to-Text API](#)
- [Gemini API: Quickstart with Python](#)
- [Gemini API code sample](#)
- [Compare Gemini with ChatGPT](#)
- [More notebooks](#)

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.



Start coding or [generate](#) with AI.

## What is Colab?

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](#) to find out more, or just get started below!

## ✓ Getting started

The document that 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 find out more, see [Overview of Colab](#). To create a new Colab notebook you can use the File menu above, or use the following link: [Create a new Colab notebook](#).

Colab notebooks are Jupyter notebooks that are hosted by Colab. To find out more about the Jupyter project, see [jupyter.org](#).

## ▼ Data science

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

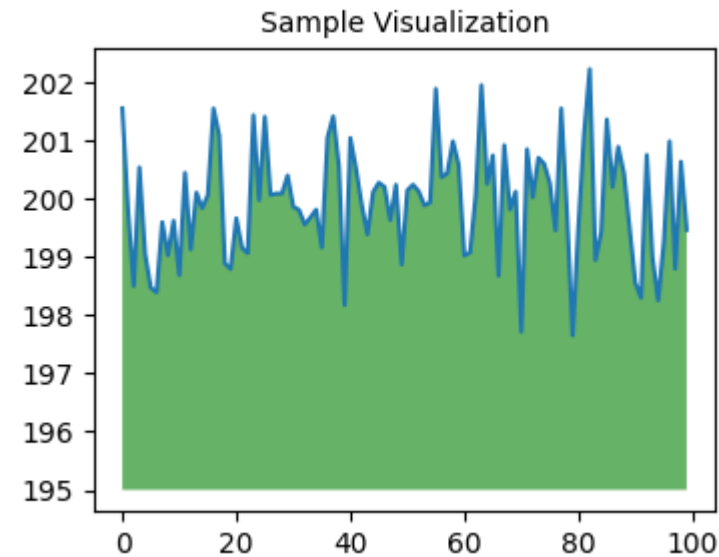
```
import numpy as np
import IPython.display as display
from matplotlib import pyplot as plt
import io
import base64
```

```
ys = 200 + np.random.randn(100)
x = [x for x in range(len(ys))]
```

```
fig = plt.figure(figsize=(4, 3), facecolor='w')
plt.plot(x, ys, '-')
plt.fill_between(x, ys, 195, where=(ys > 195), facecolor='g', alpha=0.6)
plt.title("Sample Visualization", fontsize=10)
```

```
data = io.BytesIO()
plt.savefig(data)
image = F"data:image/png;base64,{base64.b64encode(data.getvalue()).decode()}"
alt = "Sample Visualization"
```

```
display.display(display.Markdown(F"!!![{alt}]({image})"))  
plt.close(fig)
```



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 find out more about importing data, and how Colab can be used for data science, see the links below under [Working with data](#).

## ✓ Machine learning

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](#). Colab notebooks execute code on Google's cloud servers, meaning you can leverage the power of Google hardware, including [GPUs and TPUs](#), 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](#) below.

## ✓ More resources

Working with notebooks in Colab

- [Overview of Colaboratory](#)

- [Guide to markdown](#)
- [Importing libraries and installing dependencies](#)
- [Saving and loading notebooks in GitHub](#)
- [Interactive forms](#)
- [Interactive widgets](#)

## Working with data

- [Loading data: Drive, Sheets and Google Cloud Storage](#)
- [Charts: visualising data](#)
- [Getting started with BigQuery](#)

## Machine learning crash course

These are a few of the notebooks from Google's online machine learning course. See the [full course website](#) for more.

- [Intro to Pandas DataFrame](#)
- [Linear regression with tf.keras using synthetic data](#)

## Using accelerated hardware

- [TensorFlow with GPUs](#)
- [TensorFlow with TPUs](#)

## ▼ Featured examples

- [NeMo voice swap](#): Use Nvidia NeMo conversational AI toolkit to swap a voice in an audio fragment with a computer-generated one.
- [Retraining an Image Classifier](#): Build a Keras model on top of a pre-trained image classifier to distinguish flowers.
- [Text Classification](#): Classify IMDB film reviews as either *positive* or *negative*.
- [Style Transfer](#): Use deep learning to transfer style between images.
- [Multilingual Universal Sentence Encoder Q&A](#): Use a machine-learning model to answer questions from the SQuAD dataset.
- [Video Interpolation](#): Predict what happened in a video between the first and the last frame.

```
import pandas as pd
```

```
df=pd.read_csv("/content/advertising.csv")  
df
```



	TV	Radio	Newspaper	Sales
0	230.1	37.8	69.2	22.1
1	44.5	39.3	45.1	10.4
2	17.2	45.9	69.3	12.0
3	151.5	41.3	58.5	16.5
4	180.8	10.8	58.4	17.9
...	...	...	...	...
195	38.2	3.7	13.8	7.6
196	94.2	4.9	8.1	14.0
197	177.0	9.3	6.4	14.8
198	283.6	42.0	66.2	25.5
199	232.1	8.6	8.7	18.4

200 rows × 4 columns

```
df.info()
```



```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 4 columns):
#   Column      Non-Null Count  Dtype
---  -
0    TV          200 non-null    float64
1    Radio        200 non-null    float64
2    Newspaper    200 non-null    float64
3    Sales        200 non-null    float64
dtypes: float64(4)
memory usage: 6.4 KB
```

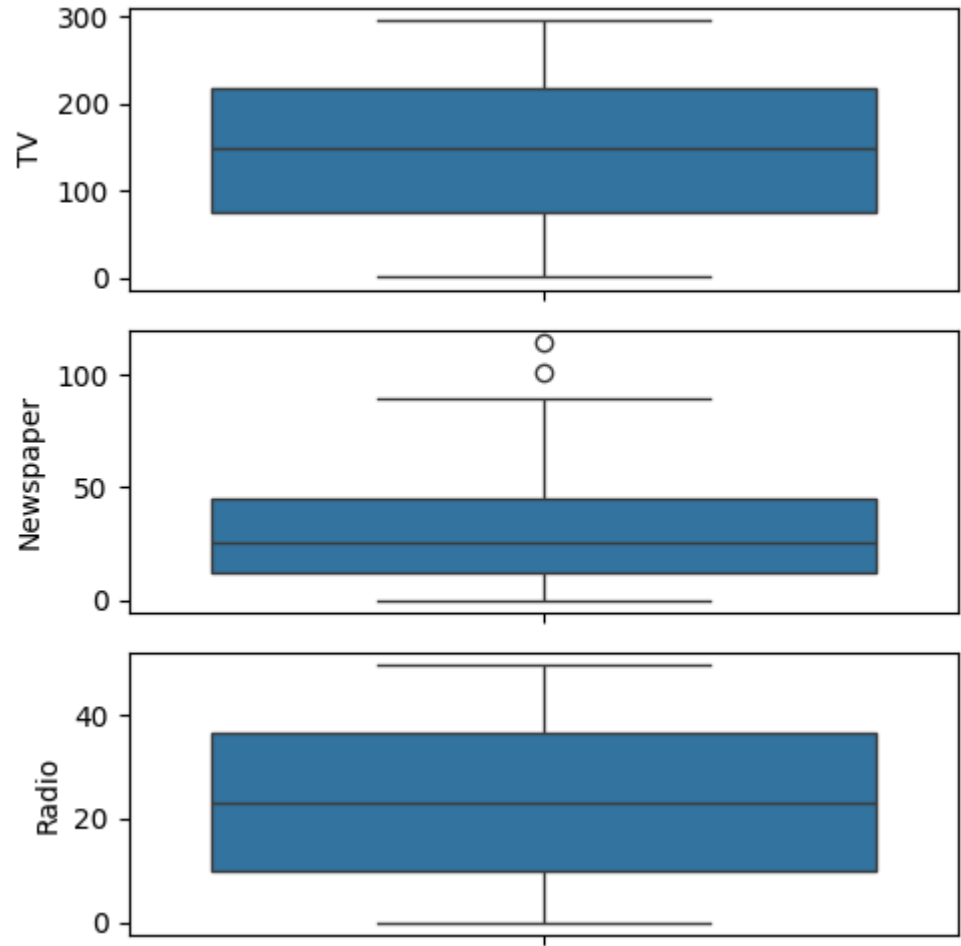
```
df.describe()
```



	TV	Radio	Newspaper	Sales
count	200.000000	200.000000	200.000000	200.000000
mean	147.042500	23.264000	30.554000	15.130500
std	85.854236	14.846809	21.778621	5.283892
min	0.700000	0.000000	0.300000	1.600000
25%	74.375000	9.975000	12.750000	11.000000
50%	149.750000	22.900000	25.750000	16.000000
75%	218.825000	36.525000	45.100000	19.050000
max	296.400000	49.600000	114.000000	27.000000

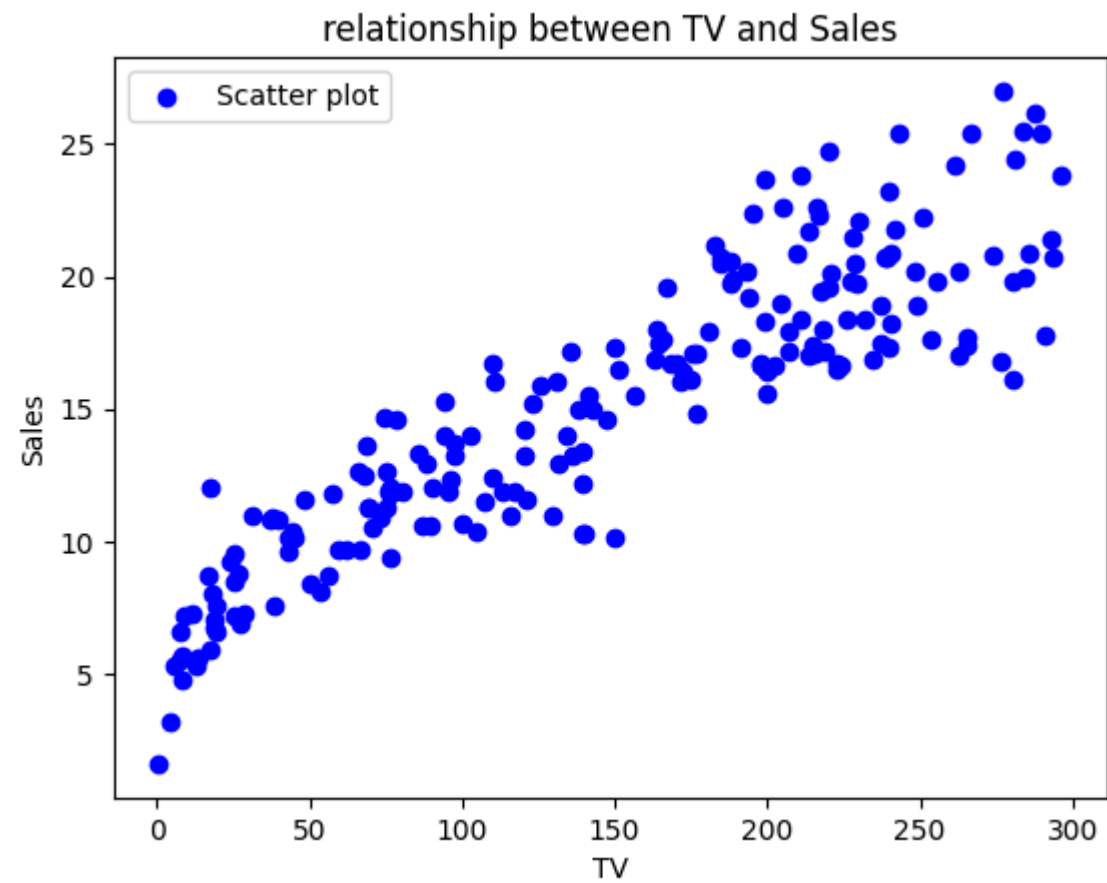
```
import matplotlib.pyplot as plt
import seaborn as sns

fig, axs=plt.subplots(3,figsize=(5,5))
plt1=sns.boxplot(df['TV'],ax=axs[0])
plt2=sns.boxplot(df['Newspaper'],ax=axs[1])
plt3=sns.boxplot(df['Radio'],ax=axs[2])
plt.tight_layout()
```



```
x=df['TV']  
y=df['Sales']
```

```
plt.scatter(x,y,color='blue',label='Scatter plot')  
plt.title('relationship between TV and Sales')  
plt.xlabel('TV')  
plt.ylabel('Sales')  
plt.legend()  
plt.show()
```



```
import numpy as np  
X=np.array(x)  
y=np.array(y)  
X=X.reshape(-1,1)  
y=y.reshape(-1,1)  
  
from sklearn.model_selection import train_test_split  
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.30,random_state=42)  
  
print(X_train.shape)  
print(y_train.shape)  
print(X_test.shape)  
print(y_test.shape)
```



```
(140, 1)  
(140, 1)  
(60, 1)  
(60, 1)
```

```
from sklearn.linear_model import LinearRegression
lm = LinearRegression()
lm.fit(X_train,y_train)
y_pred=lm.predict(X_test)
```

```
plt.scatter(X_train,y_train,color='red')
plt.plot(X_train,lm.predict(X_train),color='blue')
```

↗ [matplotlib.lines.Line2D at 0x79b1f22750d0]

