Welcome to Colab!

Access Popular LLMs via Google-Colab-Al Without an API Key

Users with Colab's paid plans have free access to most popular LLMs via google-colab-ai Python library. For more details, refer to the getting started with google colab ai.

```
from google.colab import ai
response = ai.generate_text("What is the capital of France?")
print(response)
```

Explore the Gemini API

The Gemini API gives you access to Gemini models created by Google DeepMind. Gemini models are built from the ground up to be multimodal, so you can reason seamlessly across text, images, code, and audio.

How to get started?

- Go to Google Al Studio and log in with your Google account.
- · Create an API key.
- Use a quickstart for Python, or call the REST API using curl.

Discover Gemini's advanced capabilities

- Play with Gemini multimodal outputs, mixing text and images in an iterative way.
- Discover the multimodal Live API (demo here).
- · Learn how to analyze images and detect items in your pictures using Gemini (bonus, there's a 3D version as well!).
- · Unlock the power of Gemini thinking model, capable of solving complex task with its inner thoughts.

Explore complex use cases

- Use Gemini grounding capabilities to create a report on a company based on what the model can find on internet.
- · Extract invoices and form data from PDF in a structured way.
- Create illustrations based on a whole book using Gemini large context window and Imagen.

To learn more, check out the Gemini cookbook or visit the Gemini API documentation.

Colab now has AI features powered by <u>Gemini</u>. The video below provides information on how to use these features, whether you're new to Python, or a seasoned veteran.



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 <u>Introduction to Colab</u> or <u>Colab</u> <u>Features You May Have Missed</u> to learn more, or just get started below!

Getting started

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
```



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 <u>Overview of Colab</u>. To create a new Colab notebook you can use the File menu above, or use the following link: <u>create a new Colab notebook</u>.

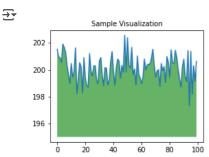
Colab notebooks are Jupyter notebooks that are hosted by Colab. To learn more about the Jupyter project, see jupyter.org.

Data science

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.

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.

```
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 \text{ for } x \text{ 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)
```



Colab notebooks execute code on Google's cloud servers, meaning you can leverage the power of Google hardware, including <u>GPUs and TPUs</u>, regardless of the power of your machine. All you need is a browser.

For example, if you find yourself waiting for **pandas** code to finish running and want to go faster, you can switch to a GPU Runtime and use libraries like <u>RAPIDS cuDF</u> that provide zero-code-change acceleration.

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To learn more about accelerating pandas on Colab, see the 10 minute guide or US stock market data analysis demo.

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 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 Colab
- · 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: visualizing data
- Getting started with BigQuery

Machine Learning

These are a few of the notebooks related to Machine Learning, including Google's online Machine Learning course. See the <u>full course</u> website for more.

- Intro to Pandas DataFrame
- Intro to RAPIDS cuDF to accelerate pandas
- Getting Started with cuML's accelerator mode
- Linear regression with tf.keras using synthetic data

Using Accelerated Hardware

- TensorFlow with GPUs
- TPUs in Colab

Featured examples

- Retraining an Image Classifier: Build a Keras model on top of a pre-trained image classifier to distinguish flowers.
- Text Classification: Classify IMDB movie 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.
- <u>Video Interpolation</u>: Predict what happened in a video between the first and the last frame.

Start coding or generate with AI.

◆ Gemini

```
```python
import pandas as pd
df = pd.read_csv('/content/StudentsPerformance.csv')
df.head()
```

₹		gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score	writing score	
	0	female	group B	bachelor's degree	standard	none	72	72	74	115
	1	female	group C	some college	standard	completed	69	90	88	
	2	female	group B	master's degree	standard	none	90	95	93	
	3	male	group A	associate's degree	free/reduced	none	47	57	44	
	4	male	group C	some college	standard	none	76	78	75	
Next steps: Generate code with df View recommended plots New interactive sheet										

Start coding or generate with AI.

print(df.shape)
df.info()
df.describe()

**→** (1000, 8)

<class 'pandas.core.frame.DataFrame'> RangeIndex: 1000 entries, 0 to 999 Data columns (total 8 columns):

#	Column	Non-Null Count	Dtype			
0	gender	1000 non-null	object			
1	race/ethnicity	1000 non-null	object			
2	parental level of education	1000 non-null	object			
3	lunch	1000 non-null	object			
4	test preparation course	1000 non-null	object			
5	math score	1000 non-null	int64			
6	reading score	1000 non-null	int64			
7	writing score	1000 non-null	int64			
dtypes: int64(3). object(5)						

dtypes: int64(3), object(5)
memory usage: 62.6+ KB

	math score	reading score	writing score	$\blacksquare$
count	1000.00000	1000.000000	1000.000000	ıl.
mean	66.08900	69.169000	68.054000	
std	15.16308	14.600192	15.195657	
min	0.00000	17.000000	10.000000	
25%	57.00000	59.000000	57.750000	
50%	66.00000	70.000000	69.000000	
75%	77.00000	79.000000	79.000000	
max	100.00000	100.000000	100.000000	

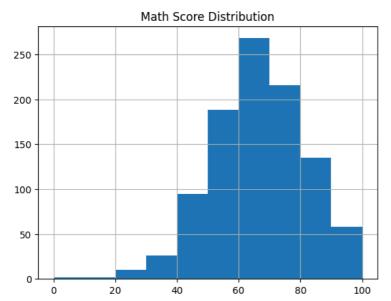
df.columns = [col.strip().lower().replace(' ', '\_') for col in df.columns]
df.head()

<del>_</del>		gender	race/ethnicity	<pre>parental_level_of_education</pre>	lunch	test_preparation_course	math_score	reading_score	
	0	female	group B	bachelor's degree	standard	none	72	72	
	1	female	group C	some college	standard	completed	69	90	
	2	female	group B	master's degree	standard	none	90	95	
	3	male	group A	associate's degree	free/reduced	none	47	57	
	4	male	group C	some college	standard	none	76	78	

Next steps: Generate code with df View recommended plots New interactive sheet

import matplotlib.pyplot as plt
df['math\_score'].hist(bins=10)
plt.title('Math Score Distribution')
plt.show()





```
df['total_score'] = df['math_score'] + df['reading_score'] + df['writing_score']
top_student = df[df['total_score'] == df['total_score'].max()]
high_achievers = df[(df['math_score'] > 90) & (df['reading_score'] > 90) & (df['writing_score'] > 90)]
gender_scores = df.groupby('gender')[['math_score', 'reading_score', 'writing_score']].mean()
```

print("Average Scores by Gender:")
print(gender\_scores)

# → Average Scores by Gender:

 math\_score
 reading\_score
 writing\_score

 gender
 female
 63.633205
 72.608108
 72.467181

 male
 68.728216
 65.473029
 63.311203

import seaborn as sns
plt.figure(figsize=(10, 5))
sns.boxplot(data=df[['math\_score', 'reading\_score', 'writing\_score']])
plt.title("Score Distribution by Subject")
plt.show()



# Score Distribution by Subject 100 80 60 40 8 8 20 0 Ö 0 0 0 math\_score reading\_score writing\_score

prep\_scores = df.groupby('test\_preparation\_course')[['math\_score', 'reading\_score', 'writing\_score']].mean()
print("\nAverage Scores based on Test Preparation Course:")
print(prep\_scores)



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 completed
 69.695531
 73.893855
 74.418994

 none
 64.077882
 66.534268
 64.504673