



```
from google.colab import drive
drive.mount('/content/drive')
```

 Mounted at /content/drive

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import math
import random
```

```
df = pd.read_csv('/content/drive/MyDrive/Colab Notebooks/adult.csv')
```

df



	age	workclass	fnlwgt	education	education_num	marital_status	occupation	relationship	race	sex	capital_gain	capit
0	39	State-gov	77516	Bachelors	13	Never-married	Adm-clerical	Not-in-family	White	Male	2174	
1	50	Self-emp-not-inc	83311	Bachelors	13	Married-civ-spouse	Exec-managerial	Husband	White	Male	0	
2	38	Private	215646	HS-grad	9	Divorced	Handlers-cleaners	Not-in-family	White	Male	0	
3	53	Private	234721	11th	7	Married-civ-spouse	Handlers-cleaners	Husband	Black	Male	0	
4	28	Private	338409	Bachelors	13	Married-civ-spouse	Prof-specialty	Wife	Black	Female	0	
...	...	...	...	...	...	...	...	...	...	...	...	...
32556	27	Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	Wife	White	Female	0	
32557	40	Private	154374	HS-grad	9	Married-civ-spouse	Machine-op-inspct	Husband	White	Male	0	
32558	58	Private	151910	HS-grad	9	Widowed	Adm-clerical	Unmarried	White	Female	0	
32559	22	Private	201490	HS-grad	9	Never-married	Adm-clerical	Own-child	White	Male	0	
32560	52	Self-emp-inc	287927	HS-grad	9	Married-civ-spouse	Exec-managerial	Wife	White	Female	15024	

32561 rows x 15 columns

Next steps:

Generate code with df

 View recommended plots


New interactive sheet

```
quasi_identifiers = ['age', 'education', 'relationship', 'sex', 'race']

df['income'] = df['income'].apply(lambda x: random.randint(20000, 70000) if x == "<50k"
                                  else random.randint(40000, 120000))

sensitive_attribute = 'income'

df
```



workclass	fnlwgt	education	education_num	marital_status	occupation	relationship	race	sex	capital_gain	capital_loss
State-gov	77516	Bachelors	13	Never-married	Adm-clerical	Not-in-family	White	Male	2174	
Self-emp-not-inc	83311	Bachelors	13	Married-civ-spouse	Exec-managerial	Husband	White	Male	0	
Private	215646	HS-grad	9	Divorced	Handlers-cleaners	Not-in-family	White	Male	0	
Private	234721	11th	7	Married-civ-spouse	Handlers-cleaners	Husband	Black	Male	0	
Private	338409	Bachelors	13	Married-civ-spouse	Prof-specialty	Wife	Black	Female	0	
...	...	...	...	...	...	...	...	...	...	...
Private	257302	Assoc-acdm	12	Married-civ-spouse	Tech-support	Wife	White	Female	0	
Private	154374	HS-grad	9	Married-civ-spouse	Machine-op-inspct	Husband	White	Male	0	
Private	151910	HS-grad	9	Widowed	Adm-clerical	Unmarried	White	Female	0	
Private	201490	HS-grad	9	Never-married	Adm-clerical	Own-child	White	Male	0	
Self-emp-inc	287927	HS-grad	9	Married-civ-spouse	Exec-managerial	Wife	White	Female	15024	

3 columns

Next steps:

[Generate code with df](#)[View recommended plots](#)[New interactive sheet](#)


```
sensitivity = df[sensitive_attribute].max() - df[sensitive_attribute].min()
epsilon_used = 1.0
```

```
def my_laplace_noise(mu, b):
    u = random.uniform(-0.5, 0.5)
    noise = mu - b * math.copysign(1, u) * math.log(1 - 2 * abs(u))
    return noise
```

```
df['noisy_income_custom'] = df[sensitive_attribute].apply(
    lambda x: x + my_laplace_noise(0, sensitivity / epsilon_used)
)
```

```
noise = df['noisy_income_custom'] - df[sensitive_attribute]
avg_noise = np.abs(noise).mean()
estimated_epsilon = sensitivity / avg_noise
```

```
print(f"Estimated  $\epsilon$  (epsilon) for Differential Privacy: {estimated_epsilon:.6f}")
```

 Estimated  $\epsilon$  (epsilon) for Differential Privacy: 0.998109

```
nums_trials = 1000
def simulate_dp_income_custom(data, epsilon, sensitivity, num_trials):
    dp_outputs = []
    for _ in range(num_trials):
        # For each trial, add custom Laplace noise to every income value
        noisy_income = data[sensitive_attribute].apply(
            lambda x: x + my_laplace_noise(0, sensitivity / epsilon)
        )
        dp_outputs.append(np.mean(noisy_income))
    return np.array(dp_outputs)

dp_full = simulate_dp_income_custom(df, epsilon_used, sensitivity, num_trials)
```

```
df_neighbor = df.drop(df.sample(1, random_state=42).index)
dp_neighbor = simulate_dp_income_custom(df_neighbor, epsilon_used, sensitivity, num_trials)
```

```
print("\nFull dataset DP mean:")
print(f" Mean: {dp_full.mean():.4f}, Std: {dp_full.std():.4f}")
print("Neighboring dataset DP mean:")
print(f" Mean: {dp_neighbor.mean():.4f}, Std: {dp_neighbor.std():.4f}")
```



Full dataset DP mean:  
Mean: 79957.4620, Std: 634.2368  
Neighboring dataset DP mean:  
Mean: 79990.5513, Std: 626.9773

```
plt.figure(figsize=(10,6))
plt.hist(dp_full, bins=30, alpha=0.6, label="Full Dataset DP Mean")
plt.hist(dp_neighbor, bins=30, alpha=0.6, label="Neighboring Dataset DP Mean")
plt.xlabel("DP Mean of Income")
plt.ylabel("Frequency")
plt.title("Empirical Distributions of DP Mean for Income (Custom Laplace Noise)")
plt.legend()
plt.show()
```

