

# Youtube Data Presentation 1

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#### **OVERVIEW**

OUR DATA	Youtube v3 API				
Research Question 1	Are music videos getting higher view counts than educational videos?				
Research Question 2	Are fitness videos getting higher like counts than vlogging videos?				
Research Question 3	Does upload timing affect the number of likes a video receives?				
Conclusion	Brief restatement of the Hypothesis 1, 2 and 3				
References	https://developers.google.com/youtube/v3				











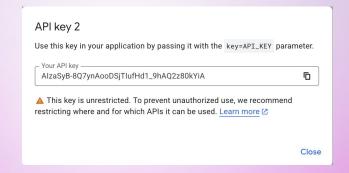
### How We Collected Our Data

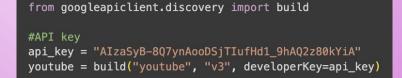
- Used YouTube Data API v3 via Google Cloud Console
- Generated API key and connected it to Python
- Collected data from YouTube for selected topics related to our hypotheses
- Extracted:
  - Video title, published date
  - View count, like count, comment count
  - Duration and other metadata
- Stored results in Pandas DataFrames for analysis



#### YouTube Data API

With the YouTube Data API, you can add a variety of YouTube features to your application. Use the API to upload videos, manage playlists and subscriptions.







## Research Question & Hypothesis 1

**Research Question:** Do music videos get higher view counts than educational videos on YouTube?

**Hypothesis:** Classical music videos have higher average view counts than math tutorials.

**Null Hypothesis:** There is no significant difference between the two categories.







### **Data Preparation**

- Retrieved data on two topics: "classical music" and "math tutorial"
- Fetched 400 videos (200 each)
- Labeled each video with a Category column (Music / Math)
- Stored results in Pandas DataFrames

	title	publishedAt	viewCount	likeCount	commentCount	duration	category
0	Peaceful Classical Music I Bach, Mozart, Vival	2024-05-06T11:00:12Z	6279598	37528	1322	PT1H40M37S	Music
1	50 Most Beautiful Classical Music Pieces	2024-06-24T11:00:28Z	5495447	37317	1132	PT3H43M55S	Music
2	8 Hours The Best of Classical Music: Mozart, B	2015-11-04T21:54:02Z	11747452	77759	2713	PT7H25M	Music
3	Dramatic Classical Music	2024-06-28T11:00:41Z	1259352	22363	736	PT2H23M44S	Music
4	15 Most Listened To Classical Masterpieces of	2025-04-14T14:30:28Z	3744992	62550	1549	PT2H14M39S	Music
5	Classical Music for Brain Power I Mozart, Beet	2023-03-08T12:00:00Z	13762160	116141	2127	PT3H15M53S	Music
6	Deep Focus - Classical Music for Thinking	2025-05-11T23:07:38Z	266147	4148	68	PT2H25M39S	Music
7	10 Hours Classical Music I Mozart, Bach, Chop	2021-12-20T12:00:25Z	1323637	9091	249	PT10H4M49S	Music
8	Timeless Classical Music You Should Listen to	2025-10-11T16:37:51Z	1548	86		PT3H8M2S	Music
9	Classical Music for Studying	2023-05-08T11:00:37Z	3905274	35733	789	PT2H27M57S	Music

0	ma <sup>-</sup>	th_df.head(10)							
<del>∑</del>		title	publishedAt	viewCount	likeCount	commentCount	duration	category	
	0	Algebra Basics: What Is Algebra? - Math Antics	2015-05-22T17:18:33Z	9857361	148619		PT12M7S	Math	11.
		How to Actually Get Better at Math	2025-07-21T15:46:02Z	285332	16941	263	PT10M37S	Math	
	2	Math Antics - What Are Percentages?	2012-10-31T01:35:12Z	10164563	119139		PT8M53S	Math	
	3	Learn to Add! 💥 Easy Math's Addition for Kinde	2025-07-07T21:00:35Z	1536807	4499		PT11M14S	Math	
	4	Math Antics - Order Of Operations	2012-04-16T07:45:10Z	9078944	116836		PT9M40S	Math	
	5	The Key to Understanding Math (with apples)	2024-07-21T13:30:06Z	83498	4301	168	PT3M32S	Math	
	6	Cool Multiplication hack that will blow your m	2024-04-06T17:00:16Z	1838408	20878		PT31S	Math	
	7	Fractions Made EASY!	2022-08-24T18:24:24Z	1160898	24102		PT21M4S	Math	
	8	Easy Mathtrick ₹ ¶ #maths #mathematics #study #k	2025-10-09T16:40:27Z	3044	22		PT25S	Math	
	9	Grade 2 Math: Addition Solution	2019-06-25T12:53:21Z	1262037	5031	0	PT1M10S	Math	



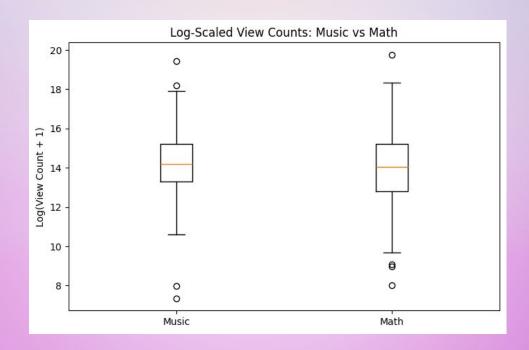




#### **Visual Analysis**

#### **Boxplot:**

- Compared view count distributions for classical music and math tutorials
- Used logarithmic scale to handle large differences in view counts
- Similar medians → average view count levels are close
- Similar spread of views
- Some outliers on both





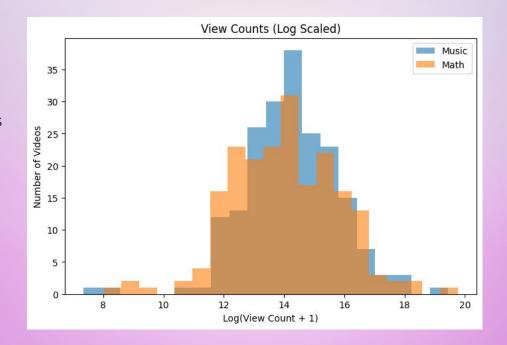




#### **Visual Analysis**

#### **Histogram:**

- Displays distribution of log scaled view counts for both topics
- Overlapping shapes → similar range of views
- Overall similar pattern, peaking at mid-range views
- Suggests that classical music videos generally attract more viewers, supporting the hypothesis





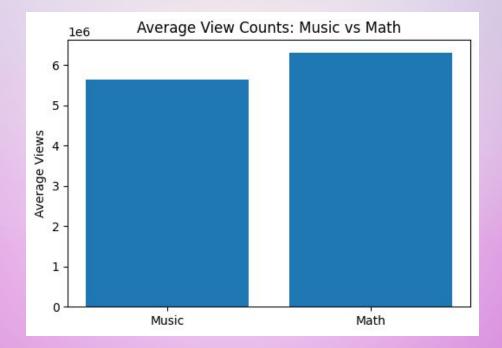






#### Bar chart:

- Compared average view counts between classical music and math tutorial videos
- Math videos slightly higher average views
- Small difference → perform similarly









#### Statistical Test & Results

- Used t-test to compare both groups
- T-statistic: -0.26
- P-value: 0.6028
- Significance level (α): 0.05
- Since p > α → Fail to reject the null hypothesis
- No significant difference between view counts

```
#t-test
from scipy.stats import ttest_ind

t_stat, p_value = ttest_ind(music_views, math_views, alternative='greater', equal_var=False)

print(f"T-statistic: {t_stat:.5f}")
print(f"P-value: {p_value:.5f}")

alpha = 0.05
if p_value < alpha:
    print("Reject the null hypothesis. Classical music videos have significantly higher view counts")
else:
    print("Fail to reject the null hypothesis. There is no significant difference")</pre>
```

T-statistic: -0.26078
P-value: 0.60280
Fail to reject the null hypothesis. There is no significant difference



## Research Question & Hypothesis 2

Research Question: Are fitness videos getting higher like counts than vlogging videos?

**Hypothesis:** Travel vlog videos get higher average like counts than Pilates Fitness videos.

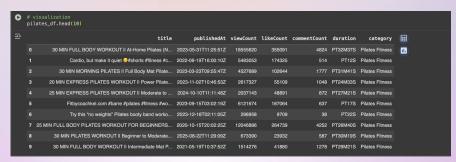
**Null Hypothesis:** There is no significant difference between the two categories.





#### **Data Preparation (2)**

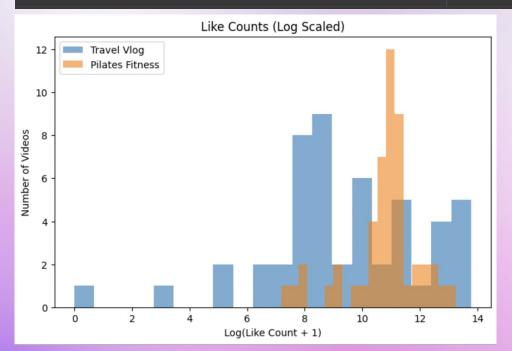
- Searched and fetched data for two topics -
  - 1. Travel vlog videos
  - 2. Pilates fitness videos
- Labeled categories and showed them in data frames for each categories.
- Visualised summary statistics for each categories.
- Stored the results.



0	vlo	og_df.head(10)							
₹		title	publishedAt	viewCount	likeCount	commentCount	duration	category	
		Food Trip sa UST by Alex Gonzaga	2025-10-12T04:00:58Z	694150	18990	658	PT26M26S	Travel Vlog	11.
		Night life in World's Richest City I NEW YORK 💴 😉	2025-10-12T04:41:00Z	114382	6383	562	PT27M15S	Travel Vlog	
		PACK, PREP AND TRAVEL W ME TO HAWAII FOR A MON	2025-09-28T19:41:46Z	54604	2188	60	PT16M54S	Travel Vlog	
		Italy Road Trip I Travel Guide to Puglia I Ita	2025-10-12T09:31:53Z	3128	198		PT26M9S	Travel Vlog	
		Guess where I ammm #travelday #travelvlog	2025-08-22T12:49:20Z	3383297	162600	353	PT1M38S	Travel Vlog	
	5	KYOTO in Autumn 🍁 quiet corners in busy Arashi	2025-10-11T11:00:59Z	25891	1386	168	PT14M37S	Travel Vlog	
		ultimate *PACK + PREP* guide for vacation I tr	2025-09-22T19:25:01Z	46875	2309	63	PT10M58S	Travel Vlog	
		Travel day as a mom of 3 X#minivlog #travelvl	2023-03-25T19:15:23Z	6598856			PT1M	Travel Vlog	
	8	Sabji bajar #shorts #minivlog #comedy #funny	2025-10-11T05:00:04Z	1075			PT44S	Travel Vlog	
	9	first time ever in JAPAN I shopping, eating an	2025-06-08T01:36:11Z	980421	26147	986	PT45M44S	Travel Vlog	



```
#Histogram - spread/shape for each category
plt.figure(figsize=(8,5))
plt.hist(np.log1p(vlog_df["likeCount"]), bins=20, alpha=0.6, label="Travel Vlog")
plt.hist(np.log1p(pilates_df["likeCount"]), bins=20, alpha=0.6, label="Pilates Fitness")
plt.title("Like Counts (Log Scaled)")
plt.xlabel("Log(Like Count + 1)")
plt.ylabel("Number of Videos")
plt.legend()
plt.show()
```



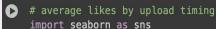
#### Visual Analysis (2)

**Histogram:** Travel vlog videos have higher clusters(for no. of likes) while Pilates fitness videos are more spread out. Thus, travel vlog videos have more likes.

Pilates fitness videos have taller bars for number of videos which doesn't indicate more videos have higher likes.

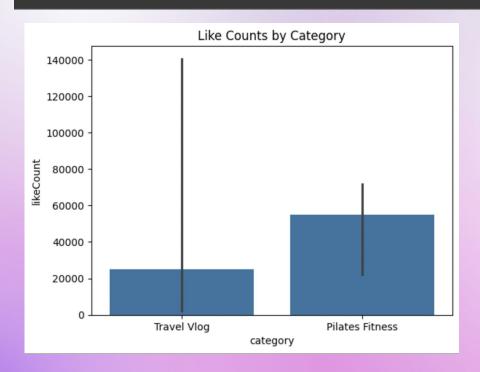
Travel vlog videos are more spread out which indicates less number of videos but has more likes than Pilates Fitness videos.





# Combine the two dataframes
df = pd.concat([vlog\_df, pilates\_df])

sns.barplot(x='category', y='likeCount', data=df, estimator=np.median, ci=95)
plt.title('Like Counts by Category')
plt.show()



#### Visual Analysis (2)

**Bar chart:** The black line for travel vlog videos is much higher than that of pilates fitness videos.

Travel vlog avg ~ 20,000 Pilates fitness avg ~ 58,000

- Observations: Huge overlap
- When confidence intervals overlap, it suggests that the difference in medians might not be statistically significant.
- Even though the Pilates Fitness bar is higher (≈ 58,000 than that of Travel vlog videos ≈ 20,000),
- The large overlap tells us that there's still a high chance they're not significantly different in reality.





### Statistical Test & Results (2)

T-test has been carried out -

1. P-value: 0.03011

2. **T-statistics:** 1.90922

Significance level is 0.05 and since

P-value > 0.05 it rejects the null hypothesis

So, we have enough evidence to support the hypothesis.

```
#t-test
from scipy.stats import ttest_ind

print(f"Samples - Travel Vlog: {len(vlog_likes)}, Pilates Fitness: {len(pilates_likes)}")

t_stat, p_value = ttest_ind(vlog_likes, pilates_likes, alternative="greater", equal_var=False)

print(f"T-statistic: {t_stat:.5f}")
print(f"P-value: {p_value:.5f}")

alpha = 0.05
if p_value < alpha:
    print("Reject the null hypothesis. Travel vlogging videos have significantly else:
    print("Fail to reject the null hypothesis. There is no significant evidence

Samples - Travel Vlog: 50, Pilates Fitness: 50
T-statistic: 1.90922
P-value: 0.03011
Reject the null hypothesis. Travel vlogging videos have significantly higher like counts.
```



## Research Question & Hypothesis 3

**Research Question:** Does upload timing affect the number of likes a video receives?

**Hypothesis (null/Ho):** There is no difference in average like counts between videos uploaded on weekdays and weekends.

**Hypothesis (alternative/H1):** Videos uploaded during the weekend get higher average like counts in comparison to videos uploaded on weekdays.





#### **Data Preparation (3)**

- Fetched 100 pages of data, and stored into a dataframe.
- Confirmed variables: 'publishedAt' and 'likeCount' to measure publish timing and interaction values.

```
# pandas dataframe creation

df = get_video_data("videos", max_pages=100) # Fetching data for 100 pages as an example

# confirm variables
df['publishedAt'] = pd.to_datetime(df['publishedAt'])
df['likeCount'] = pd.to_numeric(df['likeCount'], errors='coerce')
```

```
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 620 entries, 0 to 619
Data columns (total 8 columns):
    Column
                  Non-Null Count Dtype
    title
                  620 non-null
                                 object
    publishedAt
                  620 non-null
                                 datetime64[ns, UTC]
    viewCount
                  620 non-null
                                 int64
    likeCount
                                 int64
                  620 non-null
    commentCount 620 non-null
                                 int64
    duration
                                 object
                  620 non-null
    upload day
                  620 non-null
                                 int32
    upload timing 620 non-null
                                 object
dtypes: datetime64[ns, UTC](1), int32(1), int64(3), object(3)
memory usage: 36.5+ KB
```

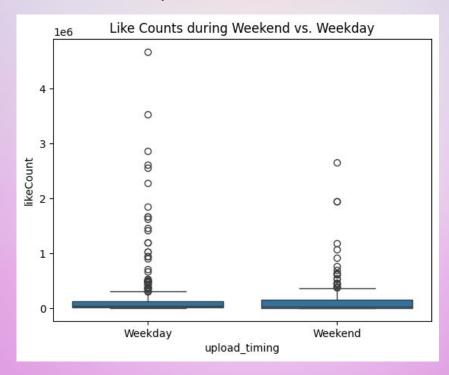




### Visual Analysis (3)

**Boxplot:** Comparison of like counts and upload timing between the weekend and weekday.

- X-Axis: upload timing (weekday vs. weekend)
- Y-Axis: like count (# likes per video)
- Interpretation:
  - Median like counts are about the same
  - Outliers are the "viral" videos
  - No significant differences between both groups
- Conclusion: Fail to reject the null hypothesis



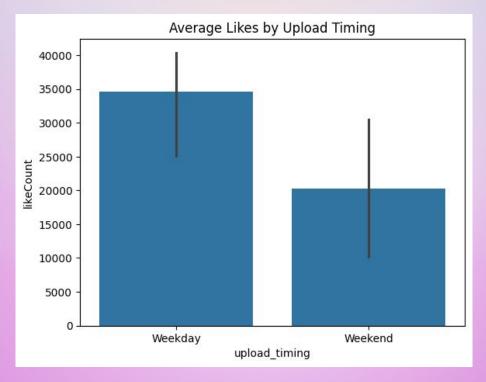




### Visual Analysis (3)

Barplot: comparing the average likes of videos uploaded on the weekday vs. the weekend.

- X-Axis: upload timing (weekday and weekend)
- Y-Axis: like count (avg. # of likes per video)
- Weekday Avg: ~ 35,000 likes
- Weekend Avg: ~ 20,000 likes
- Interpretation:
  - Black lines are 95% confidence intervals around mean (uncertainty)
  - Overlap of errors means the difference between weekend and weekday averages are not significantly different
  - Mean like count for weekday visibility higher, BUT visual evidence doesn't confirm significant difference.
  - Fail to reject null hypothesis







## Statistical Test & Results (3)

- One-tailed t-test was executed
- T-test = -1.1089
- **P-value** = 0.8660
- Seeing as the p-value

   (0.8660) is greater than the
   alpha value of 0.05, meaning
   we fail to reject the null
   hypothesis.

```
from scipy.stats import ttest_ind

# Categorize videos by upload timing (weekday vs. weekend)

#f['upload_day'] = df['publishedAt'].dt.dayofweek # Monday-0, sunday-6

#f['upload_timing'] = df['upload_day'].apply(lambda x: 'Weekend' if x >= 5 else

# Separate like counts for weekday and weekend uploads

weekday_likes = df[df['upload_timing'] == 'Weekeday']['likeCount'].dropna()

# t-test

# We use equal_var=False because we don't assume equal variances

# alternative='greater' for the alternative hypothesis: weekend likes are greater

# than weekday likes

# alternative='greater' for the alternative hypothesis: weekend likes are greater

# than weekday likes

# alternative='greater')

# Print results

print(f"Independent Samples T-Test Results:")

print(f" T-statistic: (t_statistic:.4f]")

# Interpret the results

alpha = 0.65

print("Minterpretation:")

if p_value < alpha:
    print(f" Since the p-value ((p_value:.4f)) is less than the significance level ((alpha)), we reject the null hypothesis.")

print(" Since the p-value ((p_value:.4f)) is greater than the significance level ((alpha)), we fail to reject the null hypothesis.")

print(" Since the p-value ((p_value:.4f)) is greater than the significance level ((alpha)), we fail to reject the null hypothesis.")

print(" There is not enough statistically significant evidence to suggest that videos uploaded on weekends have a higher average like count compared to videos uploaded on weekends have a higher average like count compared to videos uploaded on weekends have a higher average like count compared to videos uploaded on weekends have a higher average like count compared to videos uploaded on weekends have a higher average like count compared to videos uploaded on weekends have a higher average like count compared to videos uploaded on weekends have a higher average like count compared to videos uploaded on weekends have a higher average like count compared to videos uploaded on weekends have a higher average like count compared to videos uploaded on weekends
```







Hypothesis 1: Classical music videos have higher average view counts than math tutorials.

Null Hypothesis: There is no significant difference between the two categories.

(Failed to reject the null hypothesis)

Hypothesis 2: Travel vlogging videos get higher average like counts than pilates fitness videos.

**Null Hypothesis:** There is no significant difference between the two categories

(rejects the null hypothesis)

**Hypothesis 3:** Videos uploaded during the weekend get higher average like counts in comparison to videos uploaded on weekdays.

**Null Hypothesis :** There is no difference in average like counts between videos uploaded on weekdays and weekends.

(Fail to reject the null hypothesis)



### 1

#### References

https://developers.google.com/youtube/v3



# THANK YOU

