





SANDBOX

Minerva University

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Team click.ai



Top row (from left to right)

Saad Bin Ihsan, Nico Gankhuyag, Chris Fok, Eisha

Bottom row

Steven H. Yang

Not in the picture

Hugo Siu



Project Focus

Thumbnail

- Project significance
- Approach
- Successes and improvements
- Deliverable
- Team reflection

Video

- Time Stamp of Popular Comments
- Sound analysis
- **Learning Outcome**
- **Improvements**

Thumbnail Analysis

Project significance

Why is Thumbnail significant in the creator economy?

- Human eyes analyze an image in less than 13 milliseconds
- User spend only 1.8s looking once
- Thumbnail is the focus of 82% of browsing and
- The **biggest influence** to watch content as per Netflix

What are in Thumbnails?

- Color
- Text
- Faces
- Optional: localization



Approach - Data Science Project Life Cycle

- 1. Business Understanding
 - a. Netflix case study 🗸
 - b. Project plan ✓
- 2. Data understanding
 - a. Data collection 🗸
 - b. Data analysis 🚧
- 3. Data preparation (feature engineering + selection)
- 4. Modelling (training)
- 5. Evaluation
- 6. Deployment

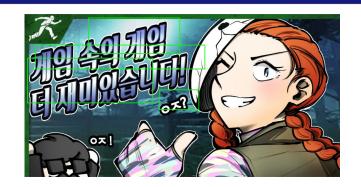
Data Collection (success)

Dependent variable

- YouTube API
- View counts, like/dislike counts, favourite counts, comment counts

Colors - independent variable

- OpenCV
- Average hue, saturation, Value



Text - independent variable

- EAST Algorithm
- Scene text percentage cover

Faces - independent variable

- Haar Cascade Classifier
- Number of faces

Data Collection (improvement)

Dependent variable

- Cannot simply compare all videos' thumbnail
- Need to control for extraneous variables

Colors - independent variable

- Mean value not conclusive → standard deviation
- Need number of colors and color distribution



Text - independent variable

- Bounding boxes are accurate estimates
- Need fine-tuning

Faces - independent variable

- Not successful
- Need more accurate algorithms like YOLOv5

Deliverable (For more detailed information, please refer to the GitHub)

```
from googleapiclient.discovery import build
def get_stats(video_id):
   # create voutube resource object
   youtube = build("youtube", "v3", developerKey = "AIzaSyDsD5jELu-4jyFRYpeUfOiueSuuBMXz7aA")
   # get the video statistics
   request = youtube.videos().list(part='statistics', id=video id)
   response = request.execute()
   # return None if request has no result, e.g. private video
   if not response['items']:
       return None
   items = response['items'][0]
   viewCount = items['statistics']['viewCount']
   likeCount = items['statistics']['likeCount']
   dislikeCount = items['statistics']['dislikeCount']
   favoriteCount = items['statistics']['favoriteCount']
   commentCount = items['statistics']['commentCount']
   return viewCount, likeCount, dislikeCount, favoriteCount, commentCount
def add apidata(df):
   for index, row in df.iterrows():
       stats = get stats(row["video id"])
       if stats is None:
            df.loc[index, 'view count'] = np.nan
           df.loc[index, 'like count'] = np.nan
           df.loc[index, 'dislike count'] = np.nan
           df.loc[index, 'favorite_count'] = np.nan
           df.loc[index, 'comment count'] = np.nan
       else:
           df.loc[index, 'view count'] = stats[0]
           df.loc[index, 'like_count'] = stats[1]
           df.loc[index, 'dislike count'] = stats[2]
            df.loc(index, 'favorite count') = stats[3]
            df.loc[index, 'comment_count'] = stats[4]
   return df
df = add apidata(df)
df.head()
```

December Color
2020-
1 08- 20 _NoMi5pp0 https://iytimg.com/vi/NoMi5pp0/sddefault.jpg 33654 1004 18 0 141 52.15 45.87 2 2021- 07-27 _le1QJ8-y8 https://iytimg.com/vi/1e1QJ8-y8/maxresdefau 6645 260 10 0 60 89.38 170.28 3 2020- 05-14 _3fHmFbnhU https://i.ytimg.com/vi/3fHmFbnhU/maxresdefau 44244 396 17 0 109 56.60 84.42 2020- 06- 06 _4sPuqw6s0 https://i.ytimg.com/vi/4sPuqw6s0/maxresdefau 111873 1250 69 0 2 59.55 13795 5 2021- 02-18 _566nRGAt4 https://i.ytimg.com/vi/566nRGAt4/maxresdefau NaN
2020- 05-14 https://i.ytimg.com/vi/_5Negroun
2020- 2020- 2020- 2020- 2020- 2020- 2020- 2020- 2020- 2020- 2021-
4 05-
2021SV-J5M/RW https://i.ytimg.com/vi/_5V-J5M/Rw/maxresdefau 274835 8994 174 0 561 30.75 48.92 7 2021SV-J5M/RW https://i.ytimg.com/vi/_6NIUmDAMA/maxresdefau 48019 729 7 0 206 90.20 58.65
04-119V-IoMirkv Intups://i-ytimg.com/vi/6NIUmDAMA/maxresdefau
7 07-31UNIONIDAMA III.Ups./ji.yuing.conjujiyiONIONIDAMAJinaxiesuelau 40019 729 7 0 200 90.20 50.00
2020-
8 2020- 05-147ba5-FNAc https://i.ytimg.com/vi/7ba5-FNAc/maxresdefau 108761 1192 54 0 480 78.09 141.49
2020- 9 06arsuxE_P8 https://i.ytimg.com/vi/arsuxE_P8/maxresdefau 28586 222 12 0 59 50.72 86.79 09
10 2020- 05-14 _bV2M-ZoqE https://i.ytimg.com/vi/_bV2M-ZoqE/maxresdefau 240531 6247 79 0 106 15.24 91.36
11 2021- 06-11 _CcmYGPCXY https://i.ytimg.com/vi/_CcmYGPCXY/maxresdefau 186861 2304 51 0 291 85.12 97.43
12 2021- 02-27cPg_qlA8c https://i.ytimg.com/vi/_cPg_qlA8c/maxresdefau 89349 940 24 0 151 62.73 102.19
2020- 13 05cwYjeVyRA https://i.ytimg.com/vi/_cwYjeVyRA/maxresdefau 20346 156 9 0 42 92.02 53.55 26
2021- 14 02d1g67sGd4 https://i.ytimg.com/vi/_d1g67sGd4/maxresdefau 22070 525 2 0 85 83.38 102.02 04

Reflection - Challenges

Steep learning curve

- Data science project life cycle
- New module: YouTube API + OpenCV
- New algorithms: text detection + face detection

Constraints

- Assignment deadlines + different workload
- Accurate data collection is crucial before moving on

Needed more support

- concrete expectation + directions
- support for technical onboarding

Reflection

Learnings

- Opportunity to work with Sandbox
- Code collaboration on GitHub and feedback

Opportunity

- Continue working on the project to *refine the accuracy of data* (priority) + include more image-related data
- Possibly work with Sandbox to train a thumbnail-processing model and deploy in the business

Video Analysis

Project significance

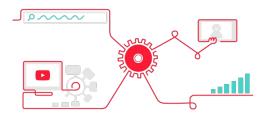
- Feedback is a major aspect in improving future video performance
- Thus, the analysis of feedback data available through Youtube api is extremely important.

Comment Analysis

- Using timestamps as a measure for traction at different points in a video, we can try to analyse the more popular portions.
- This helps us know which parts are better received by the audience



Approach



Data Collection using the Youtube Api

- Extracting all comments using the Youtube api
- Parsing the data to collect all the timestamps
- Using a min-heap to find the 5 ranges of 10 seconds which have the most traction within the video

```
class Ranges():
   def __init__(self,ranges,amount):
        self.ranges = ranges
        self.amount = amount
   def __lt__(self,other):
       return self.amount<other.amount
def clip parts(times,tophowmany,gif size):
   start = 0
   end = gif size
   ranges = []
   if len(times) == 0:
        return ranges
   for a in range(0, max(times)+1,gif size):
       current_range = Ranges((start,end),len([i for i in times if i >= start and i <= end]))</pre>
       if len(ranges) == tophowmany:
           heapq.heappush(ranges,current_range)
           heapq.heappop(ranges)
        else:
           heapq.heappush(ranges,current_range)
        start += gif size
        end += gif size
   return ranges
```



Gif creation using moviepy

- Downloading the video using the yt-dlp module
- Storing the video as a clip object and using the previously calculated ranges to make the required gifs

```
def clipper(ranges, title = ''):
    for i in ranges:
        test = VideoFileClip("test.mp4")
        if i.ranges[0] < test.duration:
            test = (VideoFileClip("test.mp4").subclip((i.ranges[0]),(i.ranges[1])).resize(0.3))
            test.write_gif(f'{title}test{i.ranges}.gif',program = 'ffmpeg')
            test.close()
        test.close()</pre>
```

Conclusions

Usefulness of Gif Creation

Highly useful with a large data set, pointed out areas well received by the audience

For example:



Conclusions (continued)

Useful... to an extent

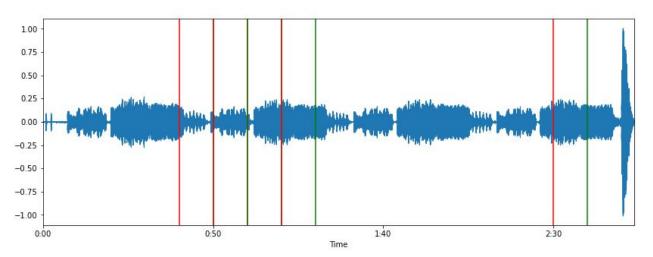
- Even though we get a good amount of information with which areas perform well, it's
 difficult to put it into place without a large enough dataset
- Smaller youtubers with not as many timestamps will not benefit from this

How it could be used

- By looking at larger more popular videos within the same genre and identifying the best parts, we can help smaller youtubers know which areas are better received by the audience.

Initial exploration

To see whether there's correlation between the audio levels of the video and the portions
of the video that are more popular



Approach

Using the librosa module to read audio data (TensorFlow)

- Firstly, the audios for the corresponding videos were downloaded using yt-dlp and converted to .wav format manually.
- Then the read functionality for wav files with librosa was used
- Alongside the waveplot to show the amplitude graph, the ranges were highlighted using matplotlib's vertical line feature

```
for i in range(10):
    import librosa
    x , sr = librosa.load(f"{i}.wav")
    print(type(x), type(sr))#<class 'numpy.ndarray'> <class 'int'>print(x.shape, sr)#(94316,) 22050

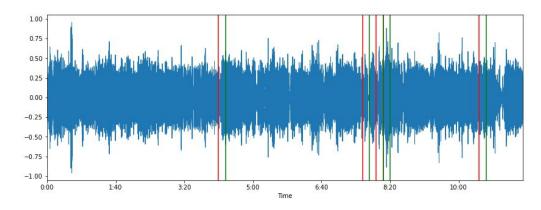
%matplotlib inline
    import matplotlib.pyplot as plt
    import librosa.display
    plt.figure(figsize=(14, 5))
    librosa.display.waveplot(x, sr=sr)
    for j in ranges[i]:
        plt.axvline(x=j[0][0], color = 'red')
        plt.axvline(x=j[0][1], color = 'green')
    plt.show()
    print(x)
```

Exploration in volume analysis

- Using runningman video as an example

Findings

- results obtained are not significant



Exploration in sound pitch analysis + confidence

- crepe module
- processes frequency, pitch and confidence

Findings

- useful in evaluating whether youtuber is confident during speeches
- overlapping between background and foreground (speech) sounds

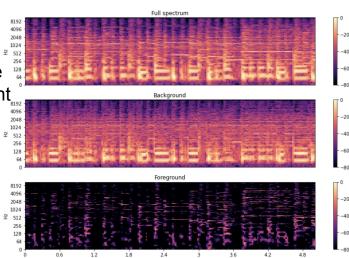
time		frequency	confidence
	0	0	nan
0	.01	1964.225	0.057465
0	.02	1961.278	0.075938
0	.03	35.996	0.062172
0	.04	145.578	0.113675
0	.05	163.862	0.10396
0	.06	103.483	0.215041
0	.07	105.621	0.179718
0	.08	161.095	0.286962
0	.09	157.39	0.188892
	0.1	338.356	0.12647
0	.11	173.464	0.133835
0	.12	351.153	0.265045
0	.13	348.9	0.241365
0	.14	347.037	0.256739
0	.15	201.131	0.456456
0	.16	206.721	0.554744
0	.17	209.429	0.201914
0	.18	233.71	0.611275
0	.19	237.796	0.75827
	0.2	241.967	0.665832
0	.21	246.281	0.604705
0	.22	1024.513	0.520397
0	.23	1040.344	0.500229

Exploration in vocal separation for sound pitch analysis

- sound pitch and effect on popularity
- overlapping between background and foreground (speech) sounds

Findings

- librosa separates the two spectrums
- however, there are many pitches occurring at the same time so results obtained still cannot pinpoint one certain sound



Conclusions

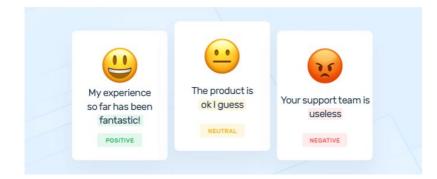
Problems

- 1. ability to separate vocal and background music
 - pitch and volume

Further possible exploration

Comment analysis

One thing that extracting gifs doesn't tell us is the exact response of the viewers towards the video which we can possibly do using a sentiment analysis on the comments to see the general response and by filtering out the negative responses, see what exactly was criticized and what could have been improved more!



https://monkeylearn.com/sentiment-analysis/

Wrap-Up



Special Thanks to Joseph, Sungsan, and Jisu from Sandbox Network, Inc. Dahyun from Minerva University.