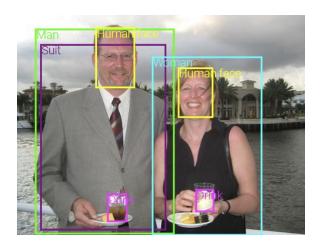
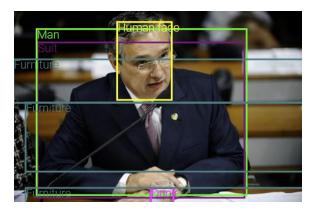
PROJECT #1

By: Nathan, Ashhad, Will

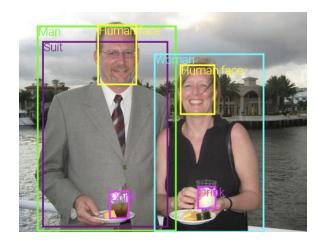
- ~9 million images, ~20 TB
- 61.4 million labels across 20638 classes
- Creative Commons 2.0 License

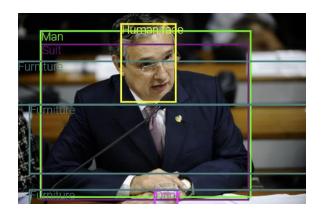
| 14364 | /m/0g5rsyg | Pistol |
|-------|------------|-------------|
| 14365 | /m/0836fh | Piston ring |
| 14366 | /m/0gw9c2 | Pistou |
| 14367 | /m/0h5xg | Pit bull |
| 14368 | /m/0gwlg1 | Pit cave |
| 14369 | /m/03mgd6 | Pit stop |
| 14370 | /m/02s6fs | Pit viper |
| | | 1 - |



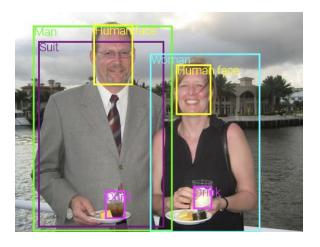


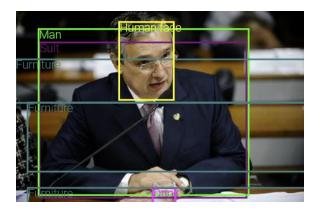
- Motivation:
 - Represents 'real life'
 - Applications in advertising, security, healthcare, accessibility
 - Predefined training, validation, test sets
 - Compare with Google Cloud Vision



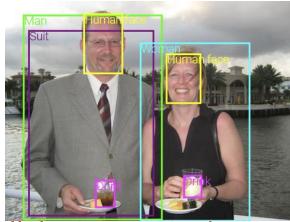


- Challenges:
 - Large size
 - Slow processing
 - Higher complexity



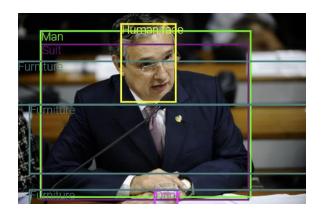


• Text Corpus :-



Collection of Human-verified and Machine-generated labels of all the images in the dataset along with their IDs

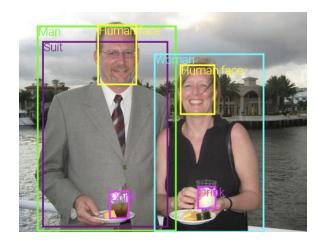
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| 14370 | /m/02s6fs | Pit viper |
| | | 1. |

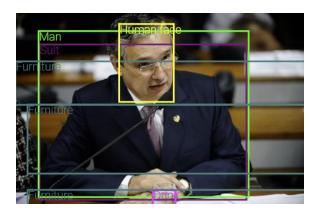


Objective:-

Gives us an idea of class distribution across the dataset

Can downscale the dataset to the classes relevant to our topic for faster processing





Defining Classes

- 'Person', 'Sky', 'Tree'
- High frequency
- Avoid "descriptions"e.g. 'Photograph', 'Beauty','Design'
- Subcategories
 'Male person' within 'person'
- Broad categories
 'Plant' vs 'Tree'

| Sky | 1 | 1072403 |
|--------------------------------------|---|---------|
| Infrastructure | 1 | 1096442 |
| Line | 1 | 1106937 |
| Photography | 1 | 1128545 |
| Building | 1 | 1132423 |
| Town | 1 | 1135208 |
| Male person | 1 | 1147919 |
| Morning | 1 | 1155162 |
| Tree | 1 | 1180918 |
| Green | 1 | 1191279 |
| Organ (Biology) | 1 | 1225968 |
| Human | 1 | 1242846 |
| Plant | 1 | 1269549 |
| Lighting | 1 | 1298671 |
| Beauty | 1 | 1334229 |
| Architecture | 1 | 1393856 |
| Nature | 1 | 1414263 |
| White | 1 | 1421473 |
| Light | 1 | 1540318 |
| Black | 1 | 1563390 |
| Design | 1 | 1806441 |
| Snapshot | 1 | 1924042 |
| Photograph | I | 2196983 |
| Person | I | 3032788 |
| + | + | + |
| 15137 rows in set (2 min 17.911 sec) | | |
| | | |

Software

- Spark/Pyspark
- MariaDB/SQL
- Google Cloud API





Setup

- Metadata stored separately from image data
- Concatenate image data with metadata to search for keywords

```
DROP TABLE IF EXISTS HumanClassDesc;
CREATE TABLE HumanClassDesc AS
SELECT Human.ImageID, Human.Source, Human.LabelName, Human.Confidence, ClassDesc.DisplayName
FROM Human
INNER JOIN ClassDesc
ON Human.LabelName = ClassDesc.LabelName;
DROP TABLE IF EXISTS MachineClassDesc;
CREATE TABLE MachineClassDesc AS
SELECT Machine.ImageID, Machine.Source, Machine.LabelName, Machine.Confidence, ClassDesc.DisplayName
FROM Machine
INNER JOIN ClassDesc
ON Machine.LabelName = ClassDesc.LabelName;
```

Maria DB

• Person Count: 4015096

• Sky Count: 2150621

• Tree Count: 1913744

• Time: 2 min 52.911 sec

Spark

• Person Count: 4015096

• Sky Count: 2150621

• Tree Count: 1913744

• Time: 7 min 34 sec

```
# Create SparkSession
from pyspark.sql import SparkSession
from pyspark.sql.functions import when
from pyspark.sql.functions import sum
import time

start = time.time()
sparkSession = SparkSession.builder.master("local[1]").appName("ParseData").getOrCreate()

dataframe = spark.read.csv("/mnt/disks/Disk1/sql/tables/TotalSet.csv")
dataframe.createOrReplaceTempView("dataset")
spark.sql("SELECT sum(case when _c4='Person' then 1 else 0 end) as PersonCount,sum(case when _c4='Sky' then 1 e
lse 0 end) as SkyCount,sum(case when _c4='Tree' then 1 else 0 end) as TreeCount FROM dataset;").show()

print("Program Elapsed Time")
print(time.time() - start, "seconds")
~
```

Conclusion

- MariaDB faster by 4 min 41 sec
- Spark better with larger datasets