

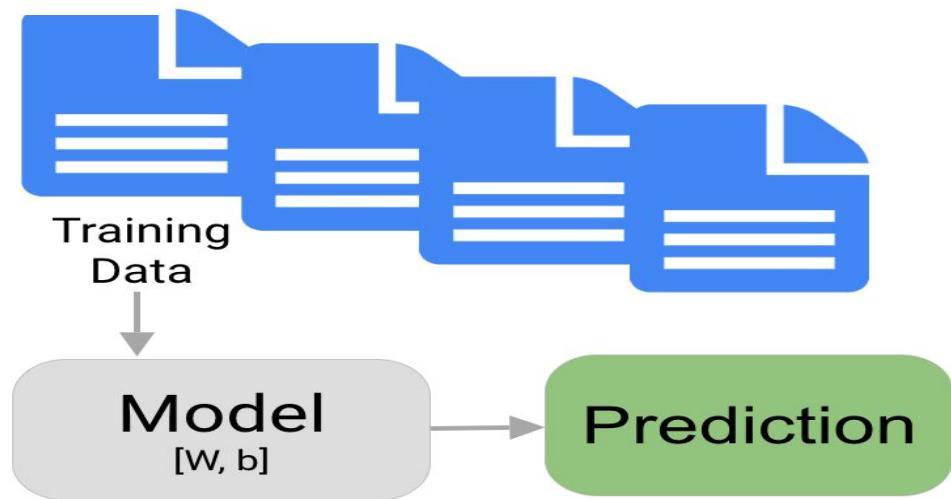
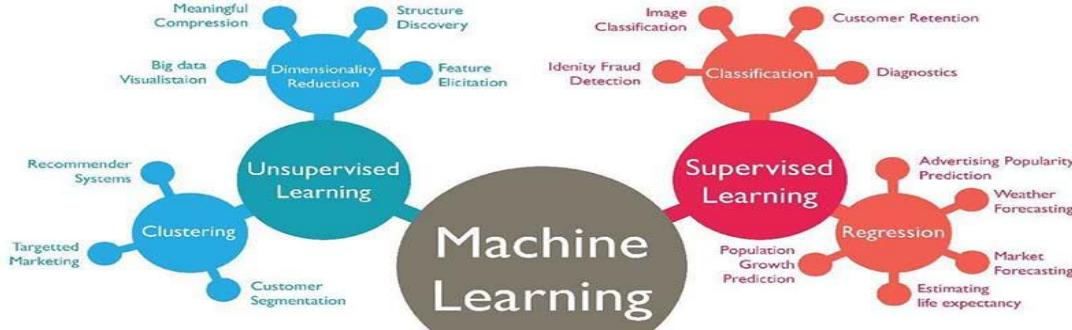
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# Analyzing and Modeling Forensic Data

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## What if:

- Unclean data
- Few labeled
  - 2699 vs 164k images
- Not precise labeling
  - Terminology
  - Bounding boxes
- Uniform textures



# What is the data?



Image Dataset Examples

An imagery collected from Anthropology Research Facility (ARF) at the University of Tennessee

- Multiple images from multiple donors
- Image sampling not uniform
  - Time of day
  - Weather conditions
  - Angles
  - Lighting
- Image content not easily comparable to other image datasets
- One million images adding up to 4TB of disk space

# What do we want to achieve?



# Why do we want to achieve this?

- Facilitate utilization of the collection for:
  - Law enforcement officers
  - Forensic researchers
  - Education
  - Statistical analysis
  - Studying the decomposition process
    - What leads to some forensic features
    - Estimating the duration of death and the decom. stage by analyzing images

# How?

- Automating the labelling process:
  - Reduce expenses and time used to label images
  - Deep Learning: Mask RCNN
- A platform that provides:
  - Usability of existing data and labels
    - Easy browsing
  - Collecting more data

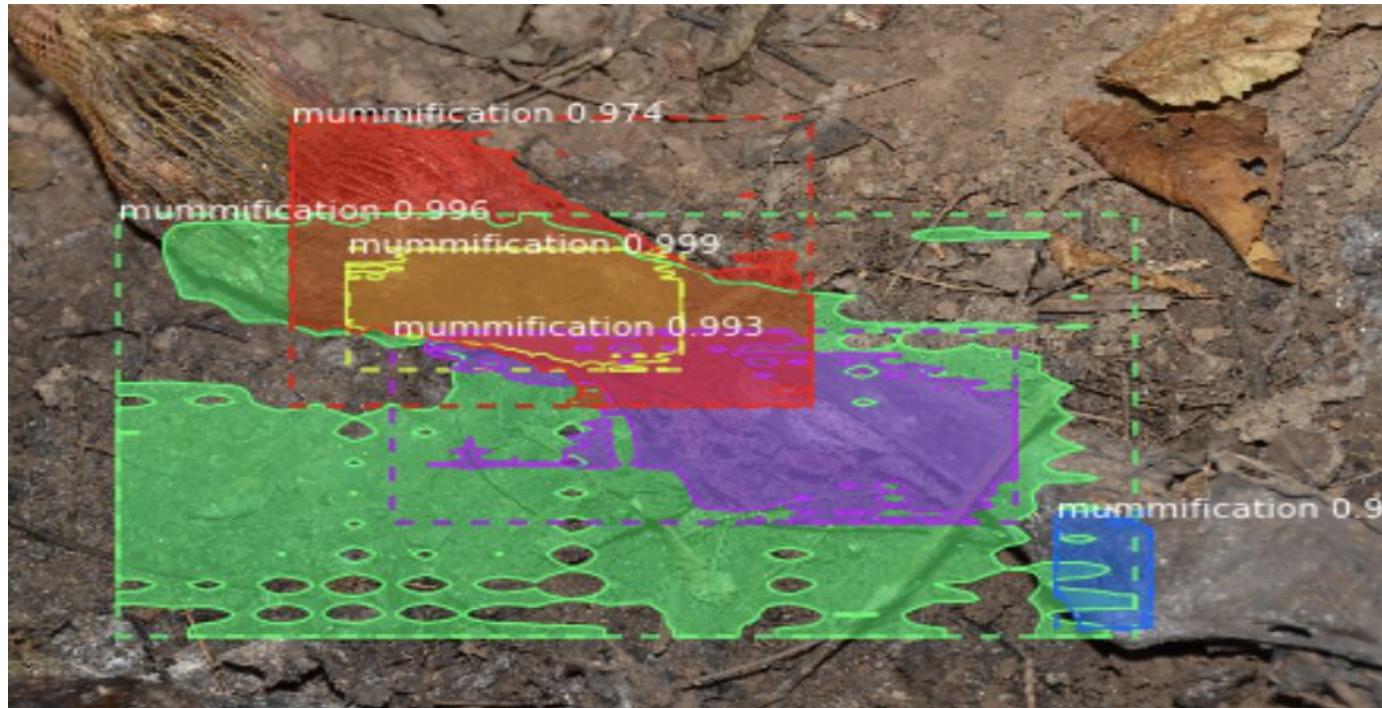
# Mask RCNN

- Deep neural network
  - Used for instance segmentation
- Input:
  - An image
- Output:
  - Bounding box
  - Class
  - Mask

# Example of Mask RCNN



# Example of Mask RCNN on our data



# Issues Our Team Investigated

**#1 Rosemary** - Improving the online platform interface and functionality

**#2 Zach** - Cleaning the input image data

**#3 Tasmia** - Adding weather data to the current system

**#4 Sara** - Expanding the dataset artificially to tune the model

# ICPUTRD: Image Cloud Platform for Use In Tagging and Research on Decomposition

- Facilitate browsing through images
- Provides tag-based categories
- Provides a tagging environment for domain exports
- Enables multiple user to access and label images



“I see putrid”

# What is a MEAN stack?



**MongoDB:** database

**ExpressJS:** server-side JS framework

**AngularJS:** frontend web framework

**NodeJS:** backend runtime environment

# Online Platform - Objectives

- Improve tag searching view
- Add carousel component to image browsing
- Add a feature to mean.js web application that tracks when an image is loaded and viewed by user
  - Currently can only track users if they tag an image
- Search tag includes duplicate images
- Fix the CSS file error, some of the pages do not load the css

# Online Platform - Approaches

- Outline issues to be improved on
- Gain access
- Become familiar with existing online platform
- Begin add Javascript code:
  - Client side
  - Server side
- Time permitting, test code
  - challenging because of some limitations of access to certain files
  - Some of the files were read only

# Results - Online Platform Tag Search Views

ICPUTRD Demography ▾ Demographics & Images ▾ Donors ▾ Tags ▾ Utags ▾ Sign Up Sign In

## Basic Tag Search

Select Choice type. "AND" : "OR"

AND  OR

fluid  purge  mummification  adult fly  eggs  larvae  discoloration  adipocere  scavenging  skin slippage  lividity  bloat  
 marbling  tattoo  mold  insect

Total "" images

Before

# Results - Online Platform Tag Search Views

ICPUTRD

Demography ▾

Demographics & Images ▾

Donors ▾

Tags ▾

Utags ▾

Sign Up

Sign In

## Basic Tag Search

Enter Tag Search Keywords

Select Choice type:

AND  OR

Search

Clear

fluid  purge  momification  adult fly  eggs  larvae  discoloration  adipocere  scavenging  skin slippage  lividity  bloat  marbling  tattoo  mold  insect

Total "0" images

After

# Data Cleaning - Objectives

- Remove duplicate and similar images
  - Processing this type of data is expensive
  - Eliminate unnecessary use of resources
- Separate background and foreground
  - Objects surrounding the target can distort results
  - Improve accuracy of labeling and analysis



# Data Cleaning - Approach

- Imagehash
  - Image Hash - constructed value based on the contents of an image
    - Average - based on average grayscale value
    - Difference - calculates the difference between adjacent pixels
    - Perceptive - depends of frequency of pixels instead of color

# Imagehash approach

- Find hashes corresponding to each image
- Stored in a file with this format:
  - Images name; hash

```
UT108-11D_01_08_12 (10).JPG;00031f7f7f7f0000
UT108-11D_01_08_12 (11).JPG;00031f7f7f7f0000
UT108-11D_01_08_12 (12).JPG;010307762-0-0492
```



```
17 00031f7f7f7f0000
15 003cfeffffffe1c00
12 70f0f0fcfcfc8f0f0
7 60f0f0f0f8f0f030
7 0038ff7f3e272000
6 e0f8fcfcfc8e0c0c0
6 e0f0f8fcfcfc0e000
6 003cfeffffffc3c00
5 7070f0f8fcfc8f0f0
5 60f0f0f0f0f0f030
5 00040e0f1f3f3000
5 00031f7f3f7f0000
5 00031f7f3f6f0000
5 00010f7f7f7f0000
5 00000c7eff1c0000
4 e0f8fcfc8f8e0c0c0
4 e0e0f8fcfcfcf0e0
4 d8fcfcfcfcfb88000
4 c0f8fcfcfcfc9c80
4 c0f8fcfcfcfc8f0c0
4 c0c0fcfcfcfcfc8c0
4 c0c0f8fcfcfc8f0c0
4 70f0f8fcfcfc8f0f0
4 60f0f0f0f0d0f030
4 180cfcc7ffe1e0000
4 040f2f1f0f0f0c0c
4 04001c3e7e060c08
4 007cfccfcfc0c0c30
4 003cfeffffffe3c00
```

# Imagehash stats

- From 26000 images
  - We found 2300 images that had at least one duplicate
- One million (989838)
  - Took 5 days for computation
  - 63468 images with one or more duplicates

# Background Removal

- What approaches
  - OpenCV
  - LabelMe
- How it works:
  - Mask created to create a general outline of objects in the image
  - Mask and original image multiplied together to generate new image

# Background Removal



# Weather Data - Objectives

- Finding the weather pattern prior to occurrence forensic features
  - Input: Image Name
  - Output: Weather data prior to that image
- Using the weather data to provide uniform sampling for the labeling
- Improve the prediction
  - Create categories of the images based on the likelihood of the occurrence of the features in specific weather
  - Doing prediction based on both images and text (weather data)

# Weather Data - Analysis

- Data Sets:
  - Image Names
  - Weather Data
  - Images with tag data
- Inconsistencies fixed
  - Image names format
  - Duplication of Image
- Data generation:
  - Donor id, max-min date
    - Per donor
    - Per image

**Image Name**

UT01-13D_03_04_13 (29).JPG
UT14-12D_04_23_!2 (1).JPG
UT14-12D_04_23_!2 (10).JPG
UT36-16D_06_30_2016 (9).JPG
UT36-16D_07_01_0216 (1).JPG

image name	ut_id	date_from_image/max_date	min_date/first date
UT-010-15D_09-21-2015.JPG	UT-010-15D	2015-09-21	2015-09-21
UT01-13D_01_23_13.JPG	UT01-13D	2013-01-23	2013-01-23
UT01-13D_01_24_13.JPG	UT01-13D	2013-01-24	2013-01-23
UT01-13D_01_26_13.JPG	UT01-13D	2013-01-26	2013-01-23
UT01-13D_01_27_13.JPG	UT01-13D	2013-01-27	2013-01-23

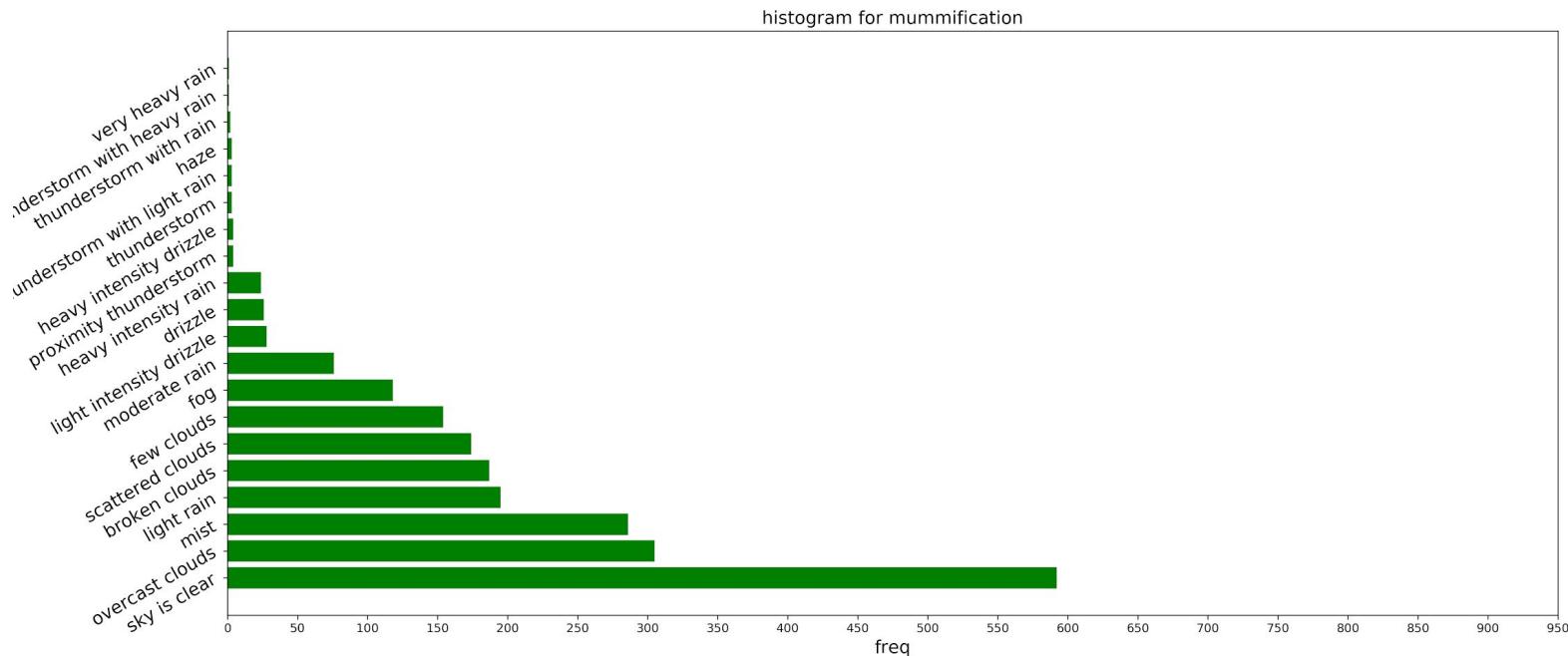
# Weather Data - Approach

- Calculating Accumulated degree date or ADDs
  - Data Analysis
    - Mapping image names to the donors
  - Finding the first date for each image name
  - Extract the dates and weather data in between minimum/first date and current date
- Weather pattern
  - Compare the weather distribution of a specific feature against that of a general feature
    - Mummification against leg

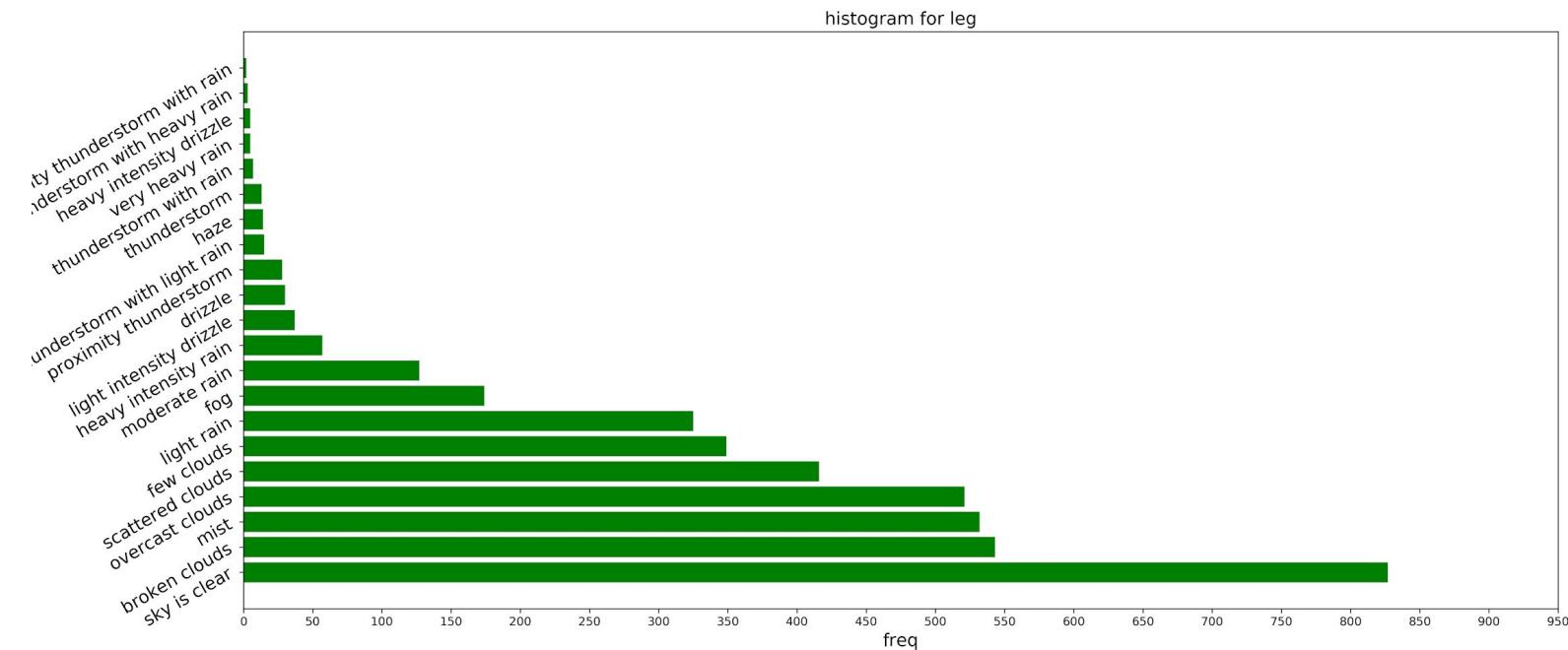
# Tagged Data - Data selection

- Mummification Data:
  - Donor: UT35-15D
  - Min Date: 2015-09-09
  - Max Date: 2016-01-09
  - Total in between days: 123
- Leg Data:
  - Donor: UT35-15D
  - Min Date: 2015-06-29
  - Max Date: 2016-01-09
  - Total in between days: 195

# Weather Data - Mummification



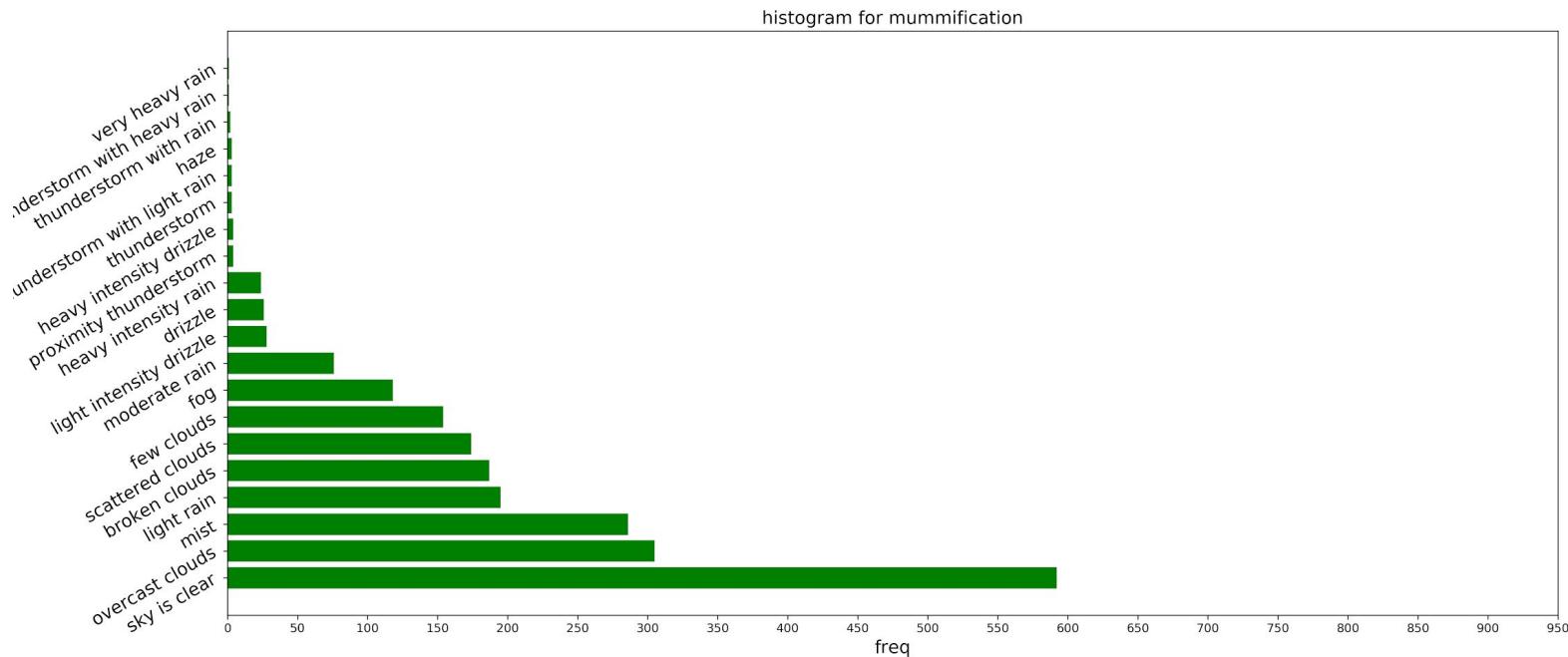
# Weather Data - Leg



# Tagged Data - Data selection: Two different donors

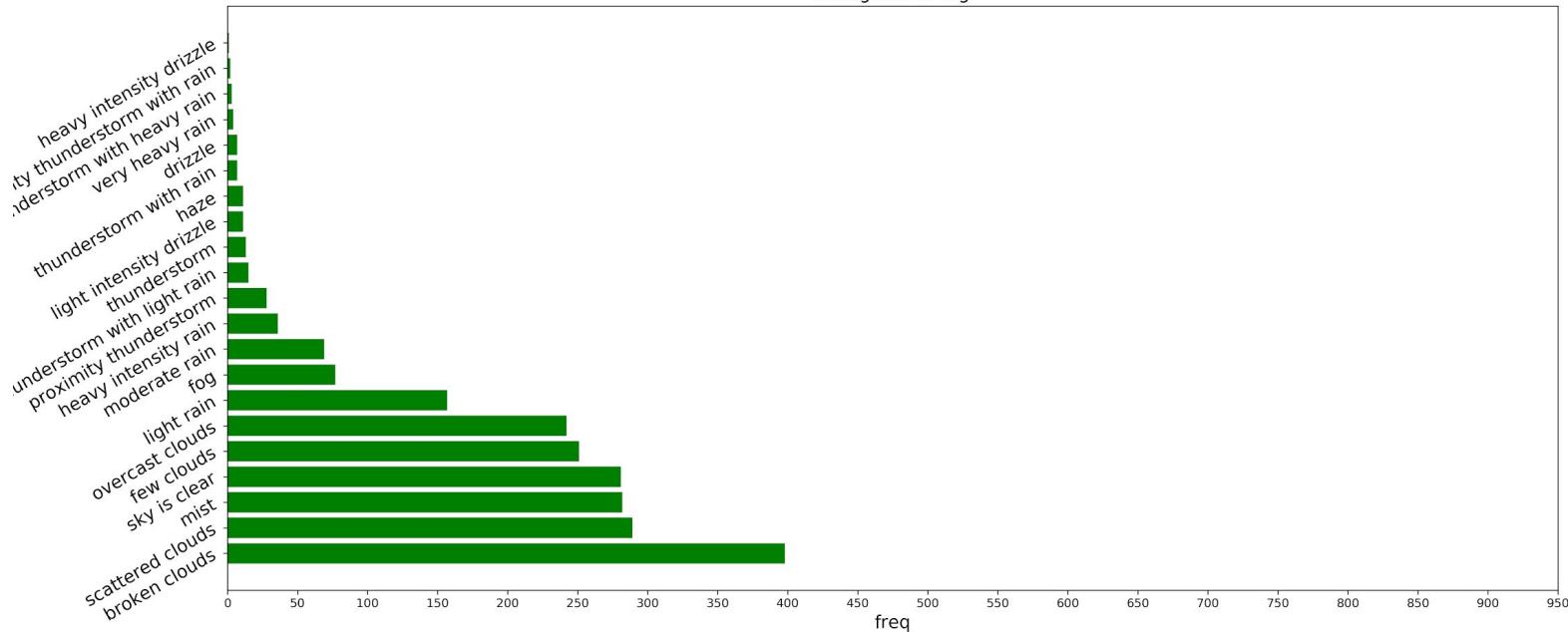
- Mummification Data:
  - Donor: UT35-15D
  - Min Date: 2015-09-09
  - Max Date: 2016-01-09
  - Total in between days: 123
- Leg Data:
  - Donor: UT06-15D
  - Min Date: 2015-04-07
  - Max Date: 2015-09-17
  - Total in between days: 195

# Weather Data - Mummification



# Weather Data - Leg

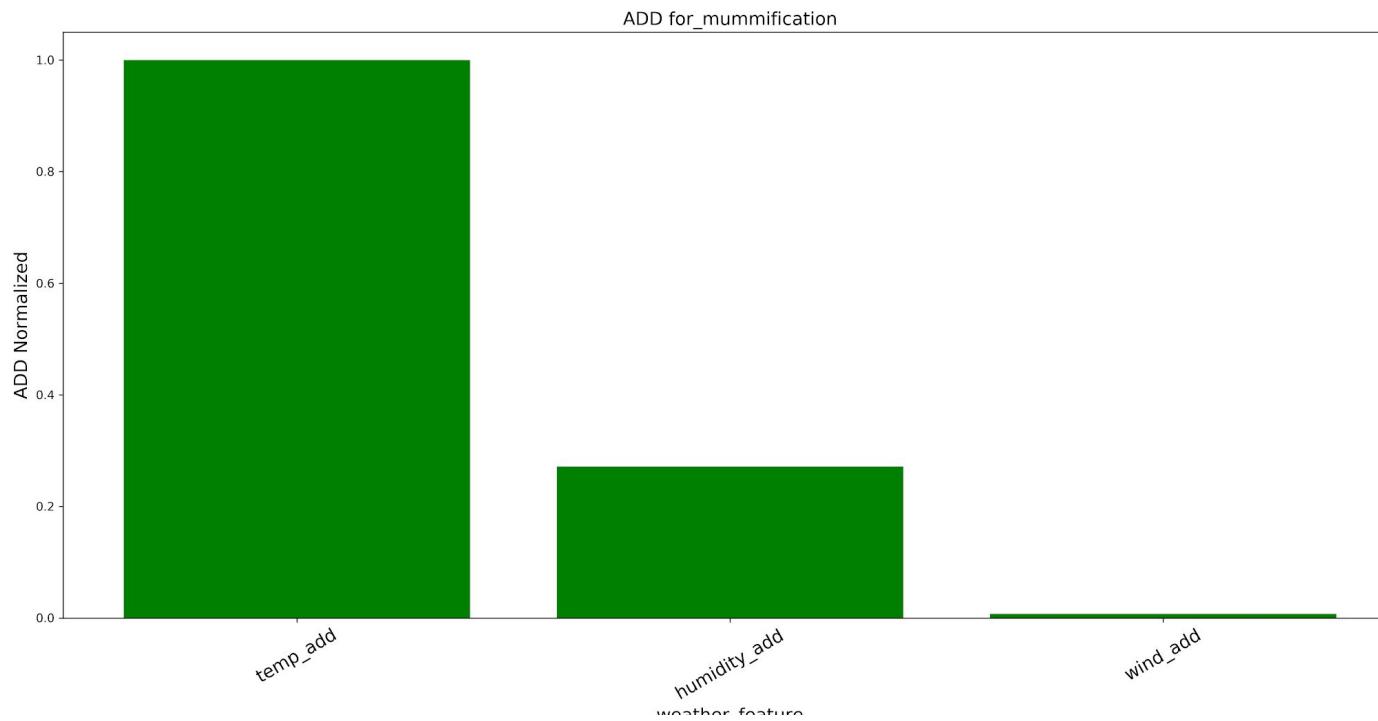
histogram for leg



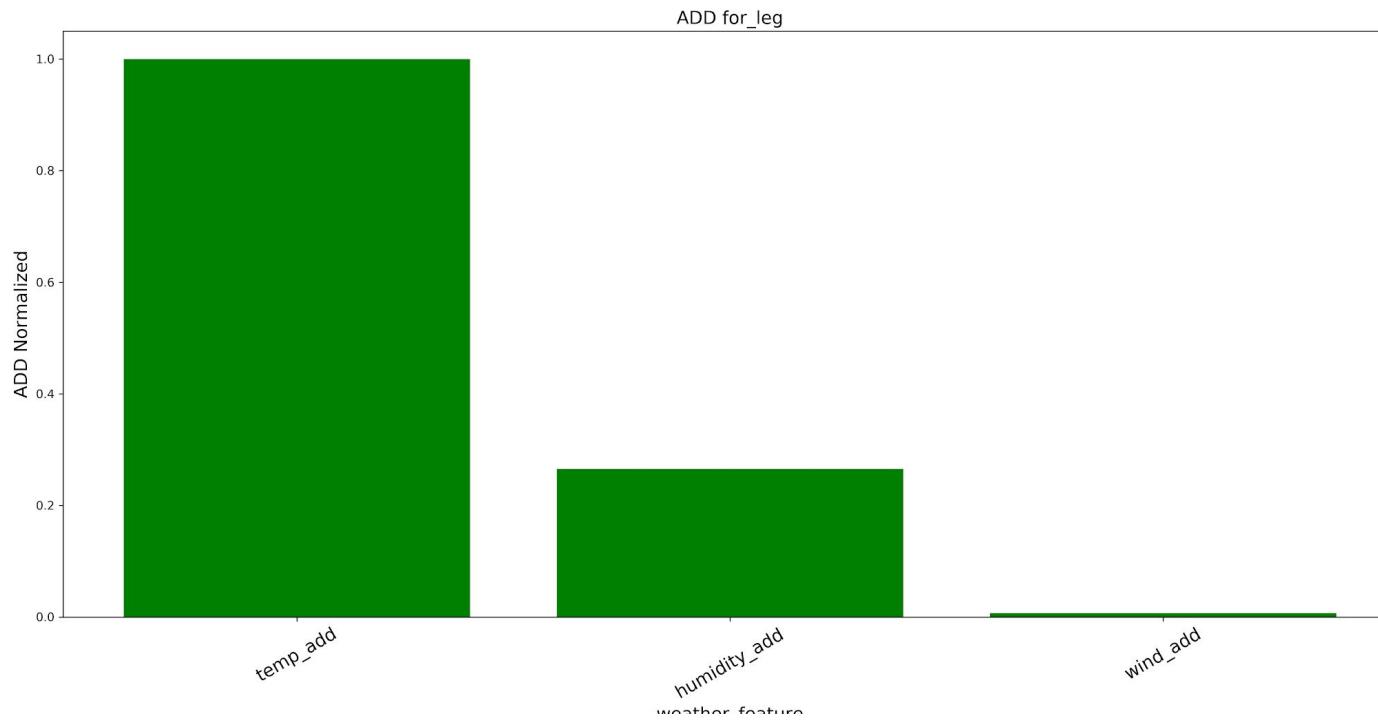
# ADD Calculation - Data Selection

- Mummification Data:
  - Donor: UT35-15D
  - Min Date: 2015-09-09
  - Max Date: 2016-01-09
  - Total in between days: 123
- Leg Data:
  - Donor: UT35-15D
  - Min Date: 2015-06-29
  - Max Date: 2016-01-09
  - Total in between days: 195

# ADD - Mummification



# ADD - Leg



# Increasing the Size of the Training Set

- Image augmentation
  - Scale
  - Flip
- Iterative validation
  - Predicting labels on unlabeled images
  - Asking a knowledgeable in forensic area person for a manual evaluation

# Augmentation: Scale

- Cropping different ratios of images with respect to each tag
  - Check if any of the other tags will be in the cropped part
  - Resize the cropped part to the original image size
  - Recalculate the location of the tag in the new image

# Augmentation: Flip image

- 0: flip around x axis
- 1: flip around y axis
- -1: flip around x and y axis

$$dest_{ij} = \begin{cases} src_{row-i-1,j} & \text{if } flipCode = 0 \\ src_{i,col-j-1} & \text{if } flipCode = 1 \\ src_{row-i-1,col-j-1} & \text{if } flipCode = -1 \end{cases}$$

# Augmentation: Flip tag

$$bbox\_flipped\_start_{ij} = \begin{cases} bbox_{i, row-j-h-1} & \text{if } flipCode = 0 \\ bbox_{col-i-w-1, j} & \text{if } flipCode = 1 \\ bbox_{col-i-w-1, row-j-h-1} & \text{if } flipCode = -1 \end{cases}$$

$$bbox\_flipped\_end_{ij} = \begin{cases} bbox_{i, row-j+h-1} & \text{if } flipCode = 0 \\ bbox_{col-i+w-1, j} & \text{if } flipCode = 1 \\ bbox_{col-i+w-1, row-j+h-1} & \text{if } flipCode = -1 \end{cases}$$

# Augmentation: image example



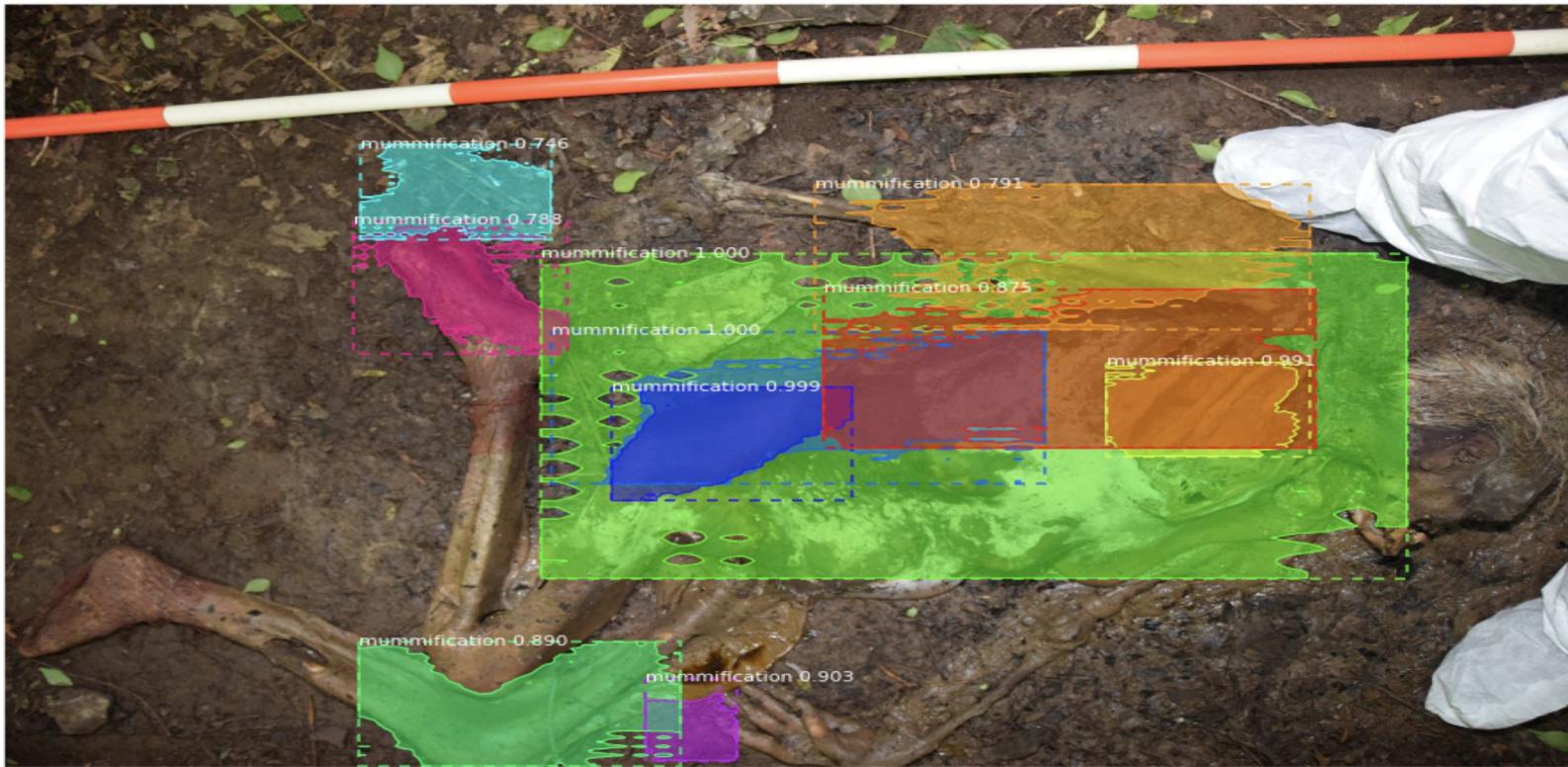
# Augmentation: csv example

row\_number,\_id,user,location,image,tag,created,\_v,image1  
205,objectID(5b97516b1025fcfc18ae1b55),objectID(5984676aea94fb4bc9c04bc64),"[{"type":"rect","geometry":{"x":0.6141666666666666,"y":0.5245283018867924,"width":0.1158333333333333,"height":0.1761006289308176}],"style":{}},http://localhost:3000/2015/UT01-15D/Daily Photos/UT01-15D\_01\_02\_2016 (3.JPG,mummification,2017-12-18T17:48:55.827Z,0,UT01-15D\_01\_02\_2016 (3.JPG  
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row\_number,\_id,user,location,image,tag,created,\_v,image1  
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205,objectID(5b97516b1025fcfc18ae1b58),objectID(5984676aea94fb4bc9c04bc64),"[{"geometry":{"x":0.1759882312986534}],"style":{},rect}],http://localhost:3000/2015/UT01-15D/Daily Photos/UT01-15D\_01\_02\_2016 (3.JPG,mummification,2017-12-18T17:49:19.081Z,0,UT01-15D\_01\_02\_2016 (3.mummification.0.5.flipped.xy.jpg  
207,objectID(5b97516b1025fcfc18ae1b58),objectID(5984676aea94fb4bc9c04bc64),"[{"geometry":{"x":0.151957844201025185}],"style":{},rect}],http://localhost:3000/2015/UT01-15D/Daily Photos/UT01-15D\_01\_02\_2016 (3.JPG,mummification,2017-12-18T17:49:19.081Z,0,UT01-15D\_01\_02\_2016 (3.mummification.0.5.flipped.x.jpg  
206,objectID(5b97516b1025fcfc18ae1b58),objectID(5984676aea94fb4bc9c04bc64),"[{"geometry":{"x":0.4381463040980467}],"style":{},rect}],http://localhost:3000/2015/UT01-15D/Daily Photos/UT01-15D\_01\_02\_2016 (3.JPG,mummification,2017-12-18T17:49:19.081Z,0,UT01-15D\_01\_02\_2016 (3.mummification.0.5.flipped.y.jpg

# Example



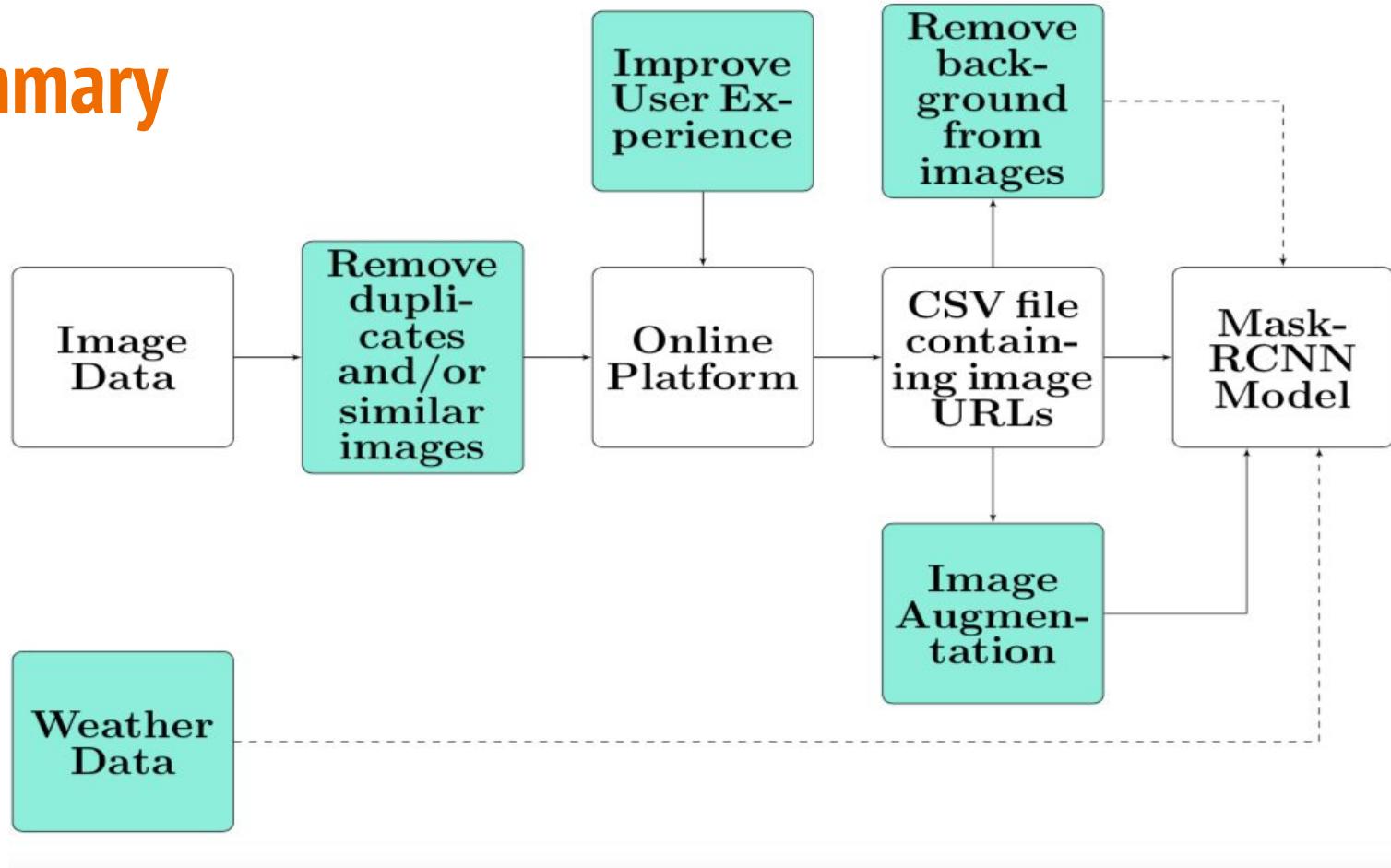
# Old model's prediction for Mummification



# New model's prediction for Mummification



# Summary



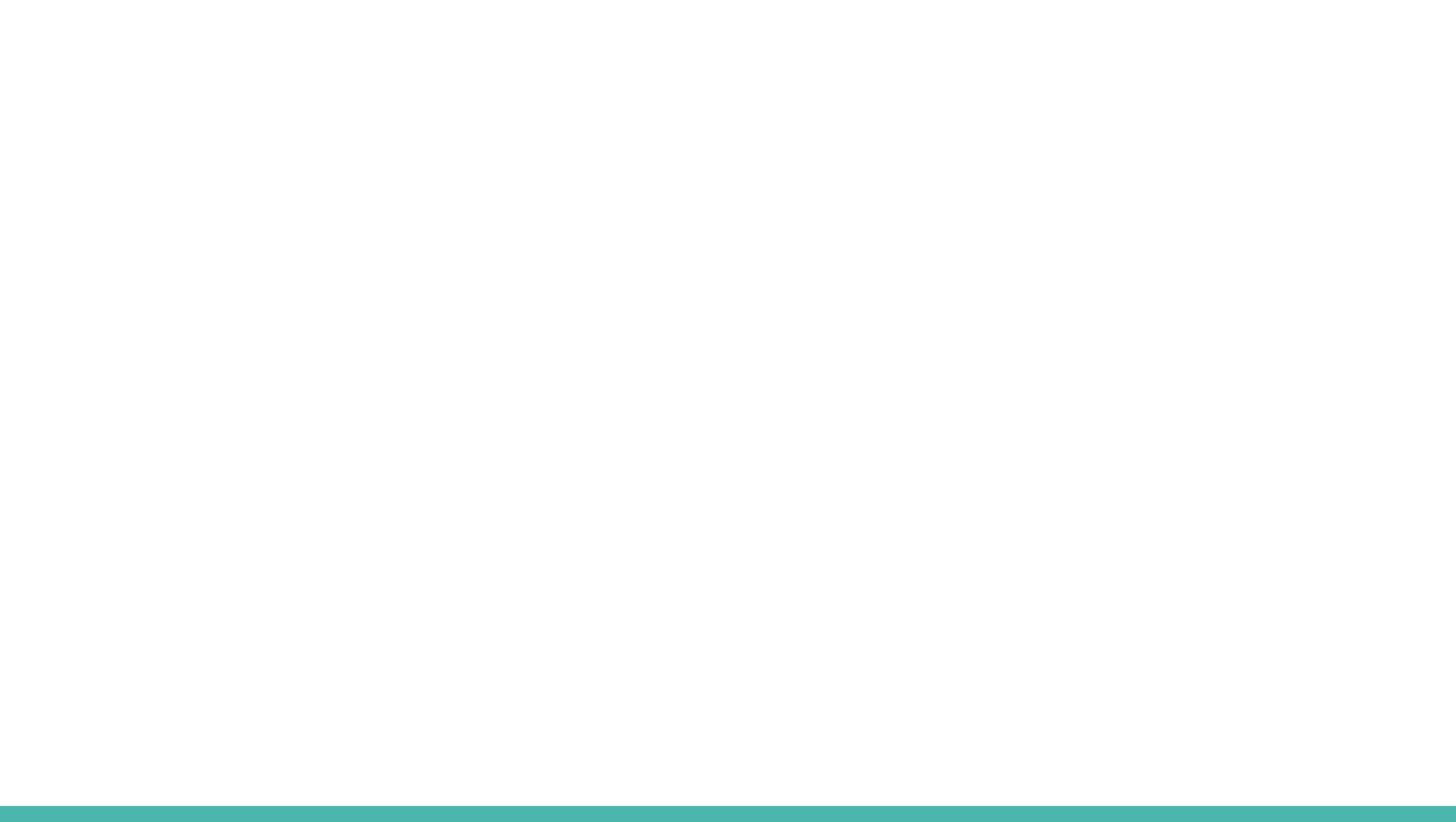
# Lessons learned

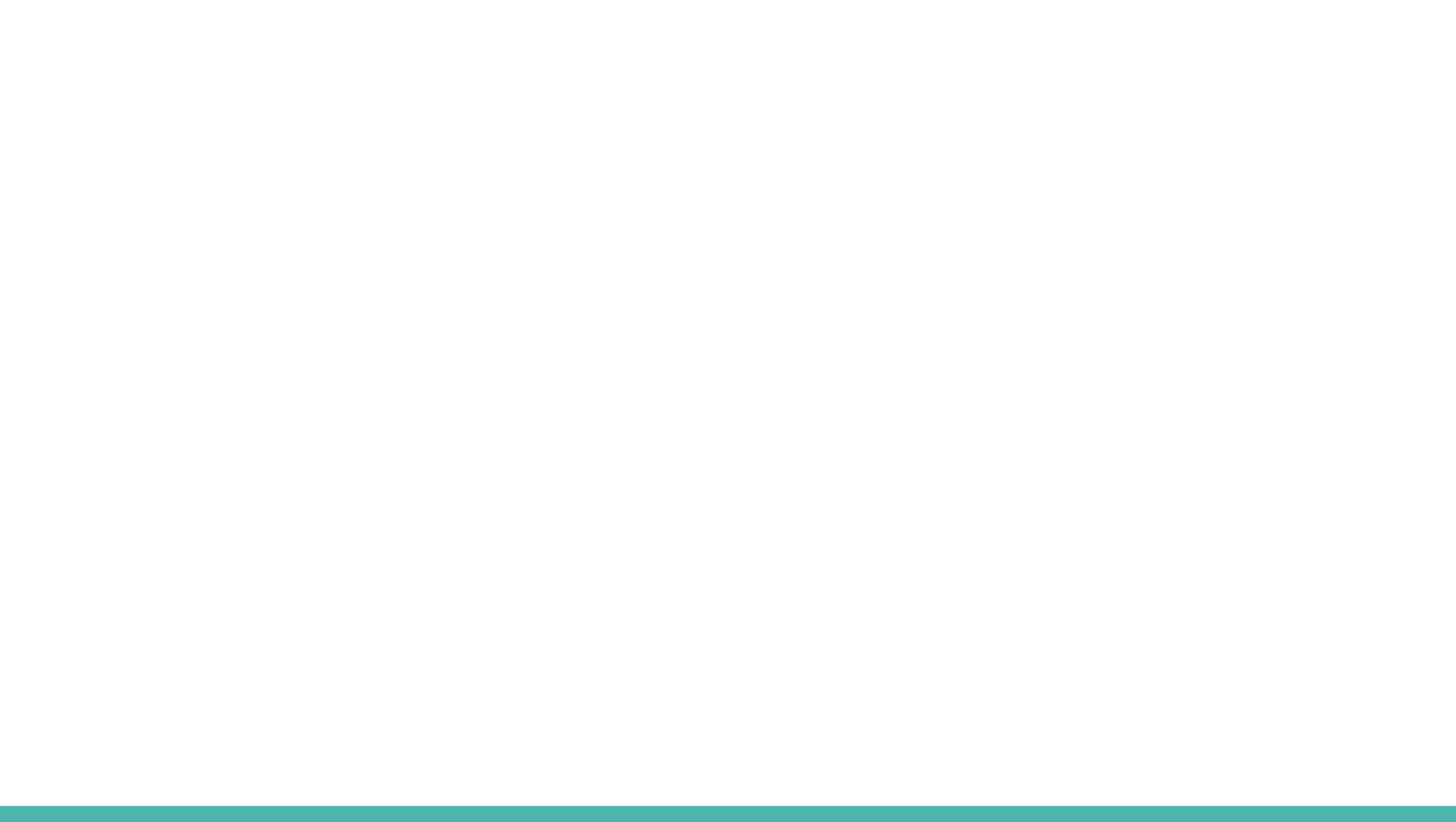
- Using docker containers slows down the development process
  - Specifically when accesses are limited

# Future Work

- Integrating the works together
- Use a simpler approach than Mask RCNN
- Train on multiple tags together

**Thank you**

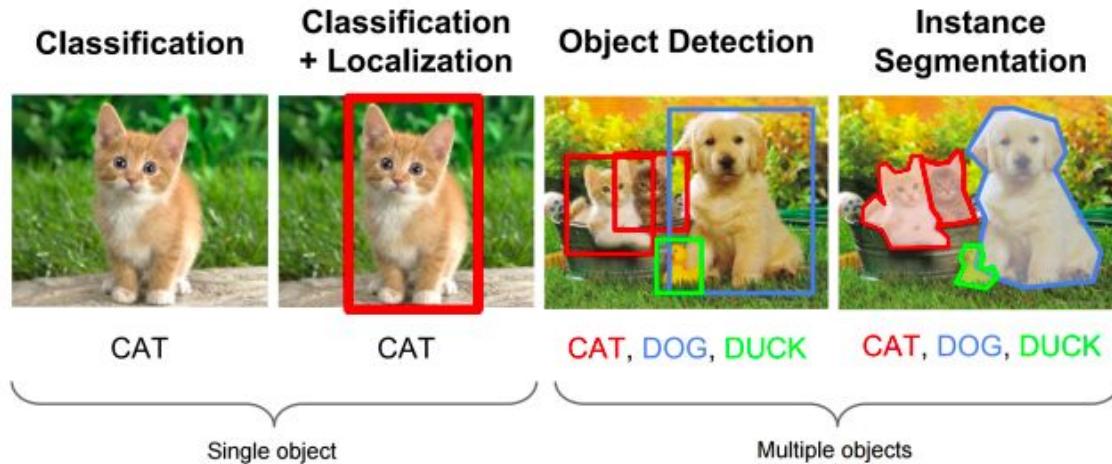




# EXTRA SLIDES:

# What is Mask R-CNN?

Object instance segmentation: detecting “objects of interest” in a particular image



Source: [https://leonardoaraujosantos.gitbooks.io/artificial-intelligence/content/object\\_localization\\_and\\_detection.html](https://leonardoaraujosantos.gitbooks.io/artificial-intelligence/content/object_localization_and_detection.html)

# ARF

- Anthropology Research Facility (ARF) at the University of Tennessee
  - Aka the “Body Farm”
  - Studying decomposition
  - Founded in 1987 by Dr. William M. Bass



# Tasmia

- A function that get the image name and gives weather data prior in a matrix format
  - col = wind, humidity, rain...
  - row = date
  - input= image name
  - Output = ..up
- Sets of images with either leg or mummification
- Plot ADD for wind, humidity, temp-ADD, rain,..
  - Distribution that we get is the stat. Part
  - Dist. of images with mumm
  - Dist. of images with either leg or mumm

# Cleaning the Data

- Remove the duplicate and similar images
  - Image hash
  - Images with identical hashes marginally different are removed
  - Any other approach than can be done fast
- Removing background
  - Improves the accuracy of the labeling algorithms
  - OpenCV library in python
- What is image hash
- We want to remove them because of expensive
- ML methods
- How many images
- How many matches
- Show numbers and results



# Incorporate Weather Data

- Weather pattern prior to each image
  - How weather affects the decom. process
- ADD value corresponding to each image
  - Image\_name components:
    - Donor ID
    - Date of image being taken
  - So calculate ADD for an image for that donor and for that date
- ADD corresponding to each donor
- Minimum date for each image
  - Each image's date is its current date
  - List all the prior dates for that image(the day the current image was taken for that donor)
  - Return the minimum date
- All the prior dates for each image

# Weather Data Purpose

ADD - uniform distribution for tagging?

What causes mummification, rain?

Implement in deep learning model

Ex. weather data plus object mask => forensic feature detection

# What I learned?

For weather data:

In addition to the images and the demography we would like to use the weather data

The primary reason is the see the effect of the decom.

Weather data is complicated and people do the ADD which is...

Show the result

A function that can return..

# Challenges

Challenge 1

## Online Platform Usability

Interface and functionality improvements.

Challenge 2

## Uncleaned Data

Duplicate images  
Having Misleading objects in images

Challenge 3

## Absence of Weather Data

Uniform sampling for different ADDs  
Do some features appear only under specific weather patterns?

Challenge 4

## Small Training Dataset

Lack of training data for model

# ADD Calculation - Data Selection

- Mummification Data:
  - Donor: UT35-15D
  - Min Date: 2015-09-09
  - Max Date: 2016-01-09
  - Total in between days: 123
- Scavenging Data:
  - Donor: UT01-16D
  - Min Date: 2016-02-04
  - Max Date: 2016-04-23
  - Total in between days: 80

# ADD - Scavenging

