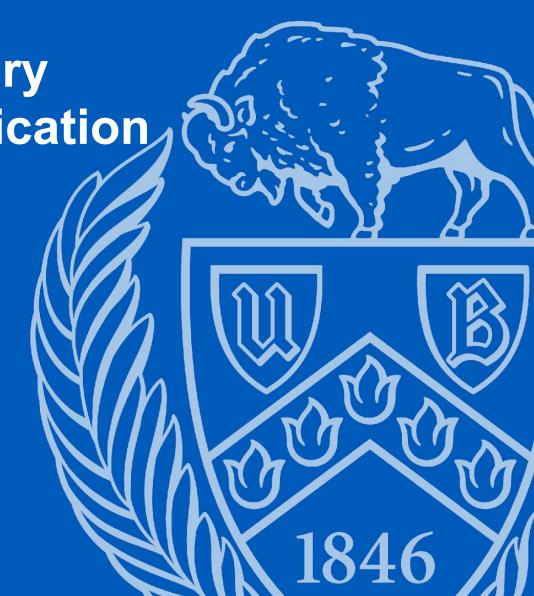
Identifying products in grocery catalogs using image classification

Team 129: akshobhy, ialtun, nsaquib2 CSE 676 Final Project 04/03/2024





Agenda

- Project Description (1 min)
- Background (2 mins)
- Dataset (3 mins)
- Methods (3 mins)
- Results (3 mins)
- Key Observations (2 mins)
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Project Description

Identifying products in grocery catalogs using image classification

- Outputs acts as labels for input images
- Web Scraped data using beautifulsoup as there was no complete dataset
- Multiclass classification on 10 classes
- Used 6 Deep Learning models and achieved highest of 85.19% accuracy

Background

- Makes the products more accessible to the customers by making the process of searching for the product more convenient.
- Easiest way for users to search for products using VR and AR technologies.
- It is going to be a necessary technology in near future.

Dataset

- No publicly available dataset that fits to our purpose.
- Collecting data using Python web scrapers.
- We expect around 1500 images per class (15000 images in total).
- Downloading 10 images in a single machine takes 5 mins. (total 125 hrs required for the download which is not feasible).

Dataset — Solution!

- We utilize CCR in a two phase approach.
- 1) Get the URL links of each image without downloading it using BeautifulSoup and store those links.
- 2) Generate 750 batch scripts where each batch corresponds to a download set of 20.
 - Using CCR, we collected required image dataset (around 1GB) for our models.

```
for cls in classes:

links = []

for i in range(5):

htmldata = getdata("search URL")

soup = BeautifulSoup(htmldata, 'html.parser')

links.append(soup.find_all('img'))

file.write(links)
```

Methods

Data Preprocessing

- We used calculated normalization parameters from our own images.
- Heavy data augmentation including resize, random rotation, cropping, horizontal flip, etc.
- 70 15 15 split for training, testing and validation.
- Used a very big size for data loaders so that we could train our model quicker.



Model architectures

- 1. Resnet with dropout layer
- 2. SE blocks with Resnet
- 3. Resnet with transformer
- 4. Resnet + SE block + transformer layer
- 5. Resnet 50 with transformer
- 6. Resnet 50



Results

Our models gave following testing accuracy:

- 1. Resnet 18 with batch norm 85.19
- 2. Resnet 18 with SE modules 84.56
- 3. Resnet 18 with transformer layer 78.859
- 4. Resnet 18 + SE block + transformer 77.65
- 5. Resnet 50 84.61
- 6. Resnet 50 + transformer 77.25

*Different results across various iteration, for transformer based models.



Key Observations

- 1. Importance of data collection scheme for model performance.
- 2. Importance of label selection.
- 3. Focus on training efficiency.
- 4. Preprocessing trade off.



Contribution Table

Name	UBIT	Contribution
Akshobhya Sharma	akshobhy	33%
Ibrahim Bahadir Altun	ialtun	33%
Nazmus Saquib	nsaguib2	33%

Thank you!



Q & A

