## EE 569 Discussion 14



**Yao Zhu** 04/16/2021



### Contents

- Announcements
- Homework 6: Successive Subspace Learning
  - Will focus more on implementation details for P2

#### Announcements

#### HW6

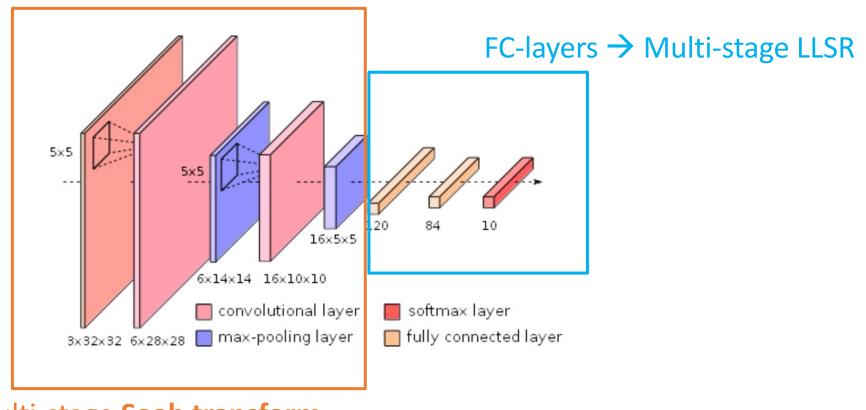
- Issued: 04/14/2021
- Due: 04/30/21 11:59pm
- Start early!!

#### Summer Intern application

- Due 4/18/2021
- We welcome all applications, send us yours!

## HW6 P1: origin of green learning

HW6 P1(a): FF-CNN and Saab Transform Feedforward-designed Convolutional Neural Networks (FF-CNNs) [2]



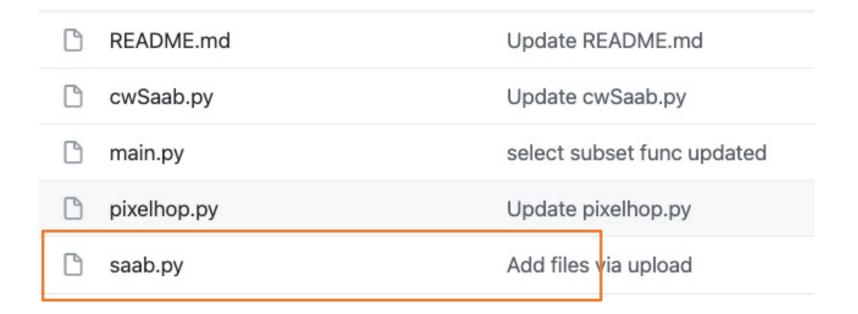
Conv layers → Multi-stage Saab transform

## HW6 P1: origin of green learning

HW6 P1(a): FF-CNN and Saab Transform

Code for Saab Transform:

https://github.com/USC-MCL/Channelwise-Saab-Transform



## HW6 P1(b): SSL Methods

#### 1. PixelHop [3]

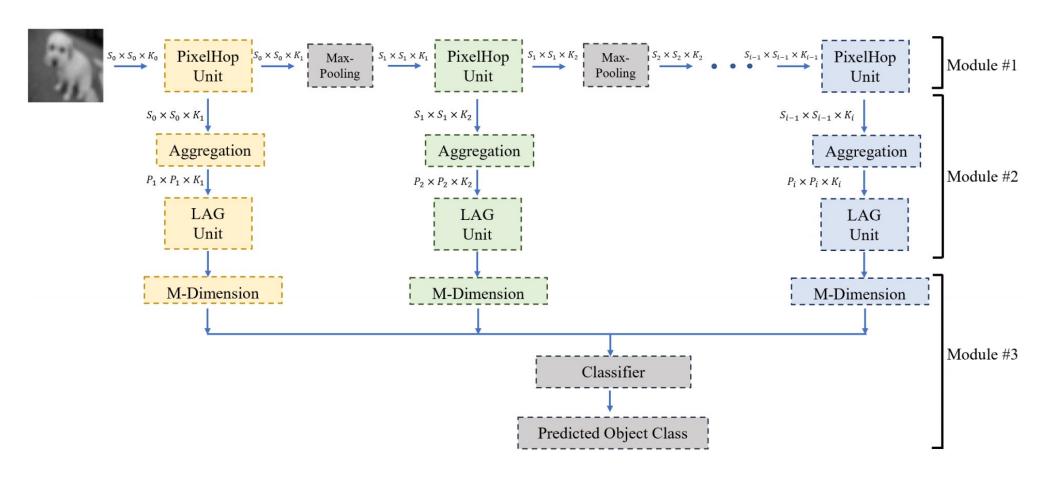


Figure 1: The block diagram of the PixelHop method.

## HW6 P1(b): SSL Methods

#### 2. PixelHop++ diagram in paper[4]

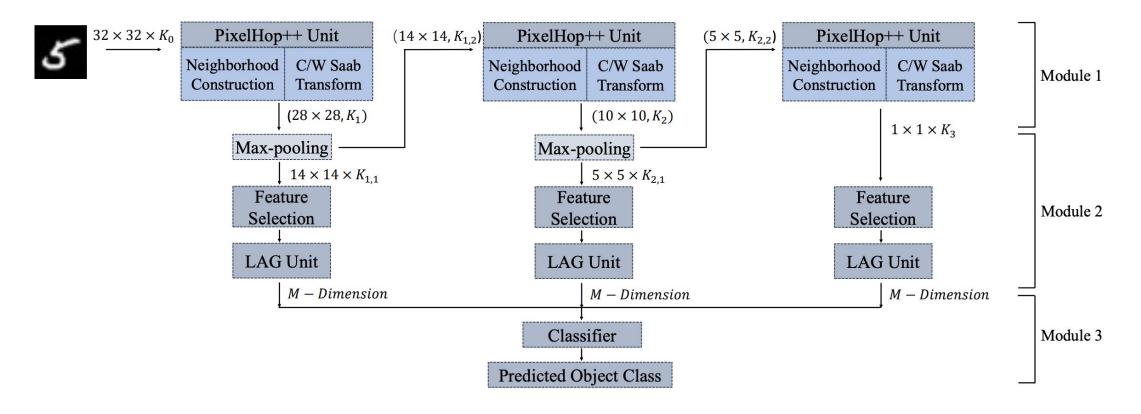


Fig. 1. The block diagram of the PixelHop++ method that contains three PixelHop++ Units in cascade.

#### Things to be covered later:

- General guidance
- Explain the source code from high level
- Explain how to use the source code
- Implementation steps

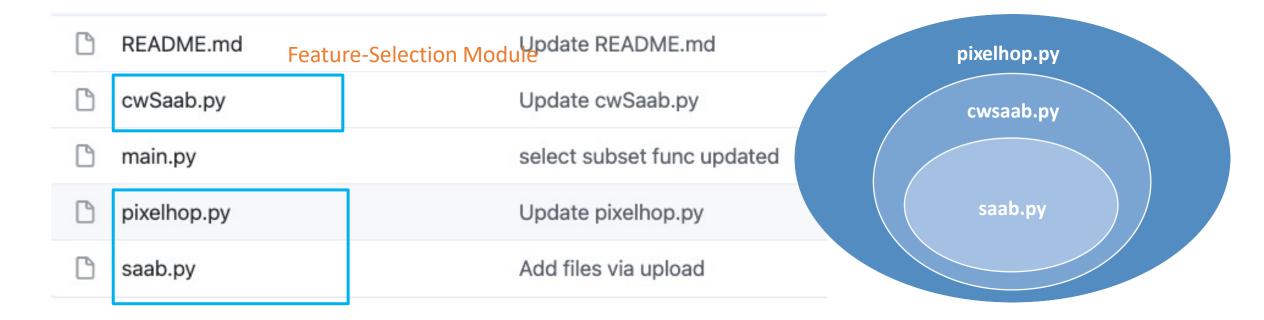
## SSL

#### **General guidance**

- Source code for key modules are provided in the GitHub link
- Need to complete main.py file by yourself
- The link is not the official implementation for the PixelHop++ paper
- Your data should be channel-last,
   i.e. shape = (N\_image, Height, Width, N\_channel)
   instead of
   (N image, N\_channel, Height, Width)

## SSL

#### More about the source code



- Main.py directly uses pixelhop.py as feature extraction
- Module 2 and Module 3 should be constructed in main.py after feature extraction using pixelhop

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#### More about the source code

About Neighborhood Construction (write your own Shrink function):

- Pooling
  - Question: why pooling is ahead of neighborhood construction in Shrink function?
- Collect 5x5 (spatial) patches

#### Example:

#### 2 Hop unit shrink argument settings

```
def Shrink(X, shrinkArg):
    #---- max pooling----
    pool = shrinkArg['pool']
    # TODO: fill in the rest of max pooling
    #--- neighborhood construction
    win = shrinkArg['win']
    stride = shrinkArg['stride']
    pad = shrinkArg['pad']
    ch = X.shape[-1]
    # TODO: fill in the rest of neighborhood construction
    if pad>0:
        X = np.pad(X,((0,0),(pad,pad),(pad,pad),(0,0)), 'reflect')
    X = view_as_windows(X, (1,win,win,ch), (1,stride,stride,ch))
    return X.reshape(X.shape[0], X.shape[1], X.shape[2], -1)
```

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### SSL

#### More about the source code

About Saab arguments:

- using Channel-wise structure or not
   PixelHop++: no channel-wise in Hop1, channel-wise for rest Hops
   PixelHop: no channel-wise for all Hops
- Need Bias or not (Hop1 no bias, rest Hops need bias)

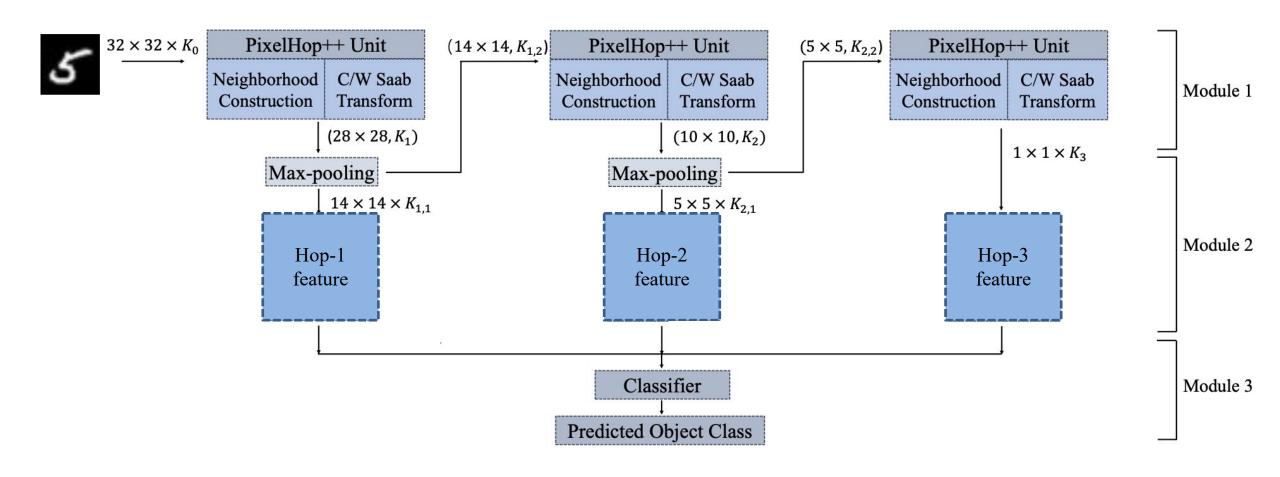
Example of Saab arguments for 2 Hop PixelHop++:

## SSL

#### More about the source code

An example about how to import key modules and how to use it:

## SSL



PixelHop++ diagram used in HW6

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## SSL



#### **Implementation Steps**

- 1. Train Module 1:
  - i. (Pre-process data): channel-last, int → float, rescale to 0-1, etc.
  - ii. Randomly select 10K training images: need to be balanced (1000 per class) if your memory can hold 50k training images, ignore this step
  - iii. Define your Neighborhood Construction function, SaabArgs, ShrinkArgs, ConcatArg
  - iv. Use pixelhop.py to train (pixelhop.fit) your PixelHop model of a certain depth
  - v. Save the PixelHop model



#### **Implementation Steps**

- 2. Extract only Hop3 features from Module 1:
  - Use all the 50K training images, do pixelhop.transform\_singleHop to extract features only from hop 3
  - Get 10K testing images Hop3 feature
  - May need batch processing to save memory. Try different programming methods on your own



#### **Implementation Steps**

- 4. Module 3:
  - i. For each image, the features from Hop3 is a long feature vector with shape 1\*1\*K3
  - ii. Choose a classifier (e.g. XGBoost)
  - iii. Train the classifier on the Hop3 feature vector of training set
  - iv. Test on the testing set

#### **Parameters**

**Table 1** Choice of hyper-parameters of PixelHop++ model for MNIST dataset

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Spatial Neighborhood size in all PixelHop++ units	5x5
Stride	1
Max-pooling	(2x2) -to- (1x1)
Energy threshold for intermediate nodes (TH1)	0.005
Energy threshold for discarded nodes (TH2)	0.001
Classifier	XGBoost
Number of estimators in classifier	100

TH1 and TH2 for **PixelHop++**For **PixelHop**, only need to
finetune TH2

## using SSL

#### Dependencies and their recommended versions for main.py

- Sklearn 0.24.1
- Scikit-image 0.18.1
- xgboost 1.4.0
- Tensorflow.keras 2.4.0

