# Plant Seedlings Classification

### Outline

- Environment
- Data Introduction
- Preprocessing
- Method
- Neural Network
- Evaluation
- Result

### Environment

- Cuda 11.6
- Numpy 1.21.6
- Pandas 1.1.5
- Tensorflow 2.10.0
- Matplotlib 3.5.3
- Scikit-learn 1.0.2

### Data Introduction - 12 Classes

- Black-grass(263 picture)
- Charlock (390 picture)
- Cleavers (287 picture)
- Common Chickweed (611 picture)
- Common wheat (221 picture)
- Fat Hen (475 picture)
- Loose Silky-bent (654 picture)
- Maize (221 picture)
- Scentless Mayweed (516 picture)
- Shepherds Purse (231 picture)

- Small-flowered Cranesbill (496 picture)
- Sugar beet (385 picture)
- The total number of training data is
   4750 images.



### Preprocessing

- Resize data into same size (299,299)
- Shuffle
- One-hot encoding
- Data augmentation
  - rotation\_range
  - width\_shift\_range
  - height\_shift\_range
  - shear\_range
  - zoom\_range
  - horizontal\_flip
  - vertical\_flip
  - brightness\_range



### Method

- Due to the limited dataset size, it may be challenging to effectively split the data into training and validation sets. In such cases, using all the data for training maximizes the utilization of the limited data resources.
- To increase data diversity and enhance model generalization, transformations such as rotation, translation, and scaling were applied to the training data. This allows the model to learn robust features that are invariant to different transformations, thereby addressing the issue of invariance.

### **Network Architecture**

- DenseNet121
- DenseNet169
- DenseNet201
- Efficientnetb0
- Efficientnetb7

- MobileNetV3 Small
- MobileNetV3 Large
- Restnet50
- Restnet101
- InceptionResNetV2

### Evaluation

- Submissions are evaluated on MeanFScore,
- which at Kaggle is actually a micro-averaged F1-score.
- Given positive/negative rates for each class k, the resulting score is computed this way:

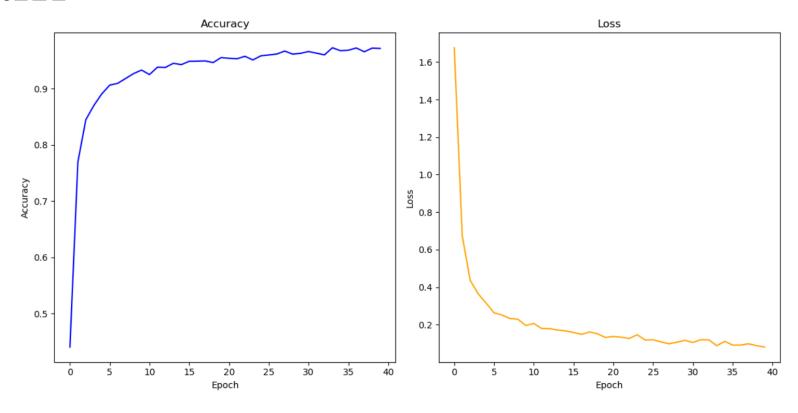
$$Precision_{micro} = rac{\sum_{k \in C} TP_k}{\sum_{k \in C} TP_k + FP_k}$$

$$Recall_{micro} = rac{\sum_{k \in C} TP_k}{\sum_{k \in C} TP_k + FN_k}$$

• F1-score is the harmonic mean of precision and recall.

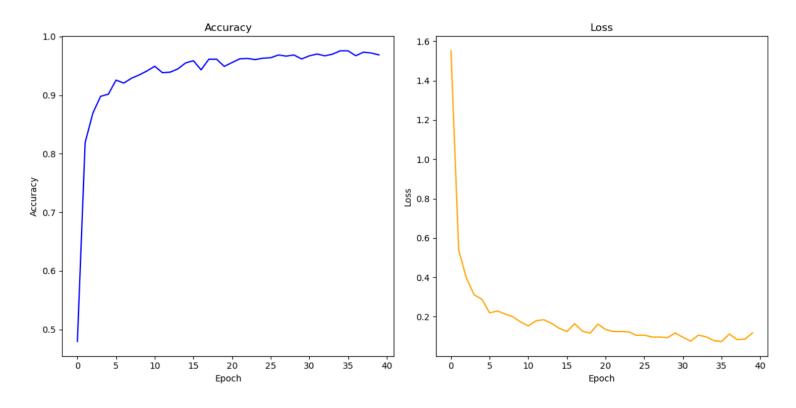
$$MeanFScore = F1_{micro} = rac{2Precision_{micro}Recall_{micro}}{Precision_{micro} + Recall_{micro}}$$

DenseNet121



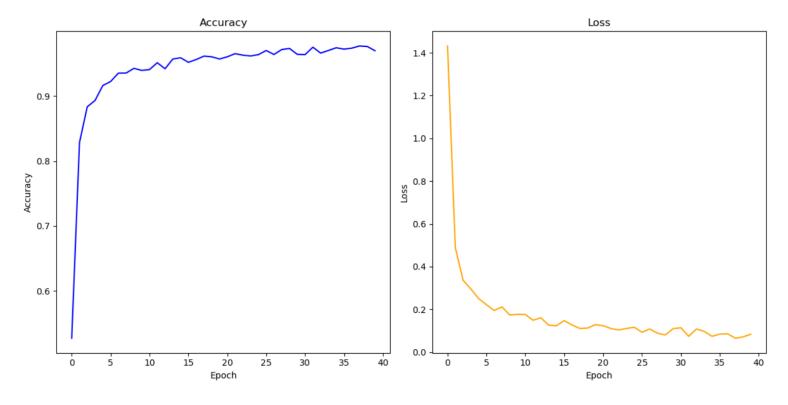


#### DenseNet169



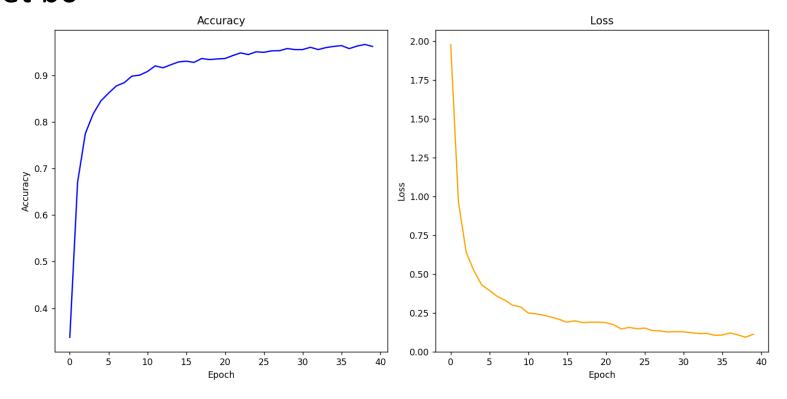


#### DenseNet201





#### • Efficientnet b0



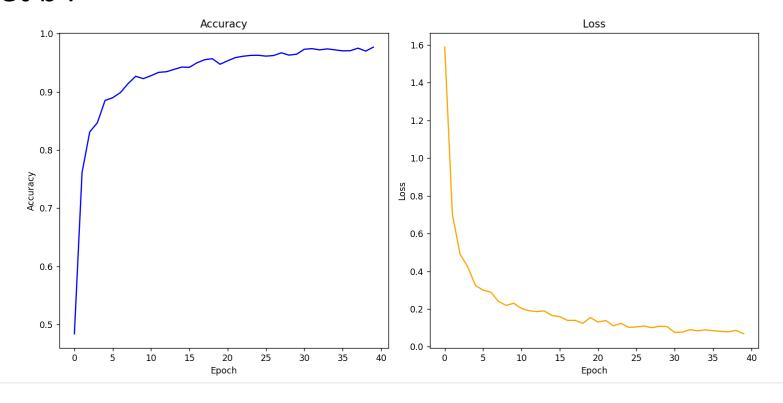


submission\_efficientnetb0.csv

0.97984

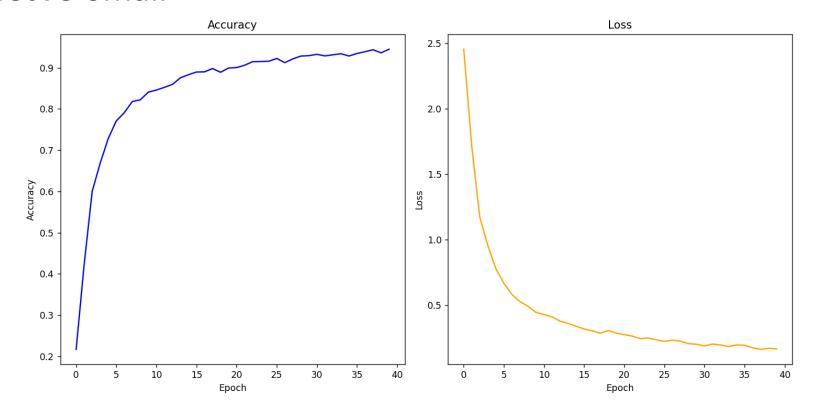
0.97984

#### • Efficientnet b4



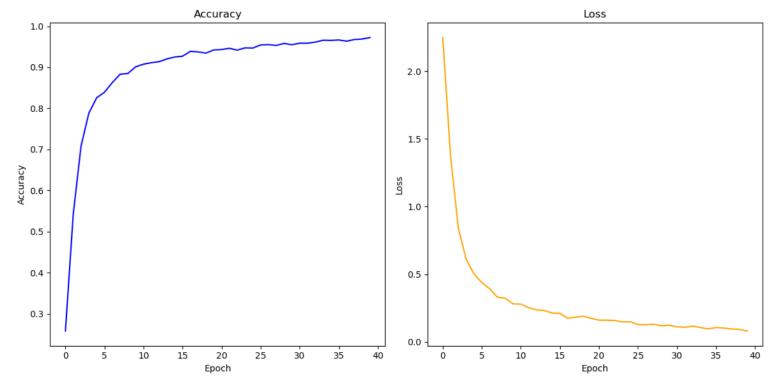


MobileNetV3 Small



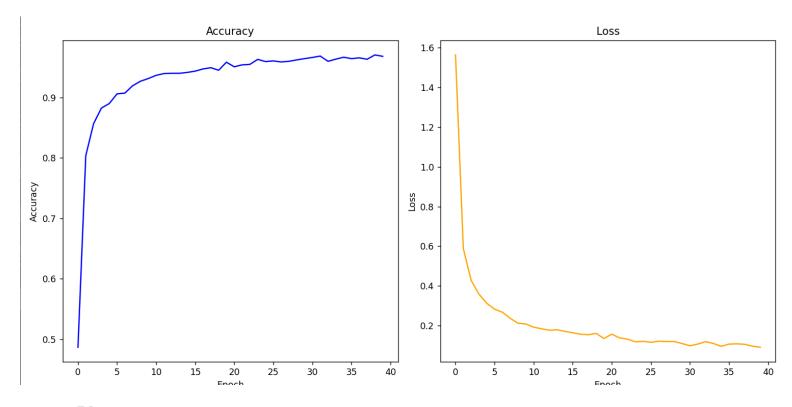


MobileNetV3 Large



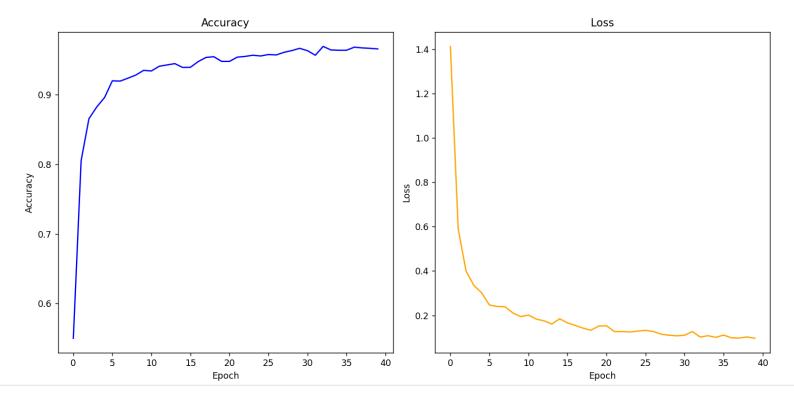


#### Restnet50

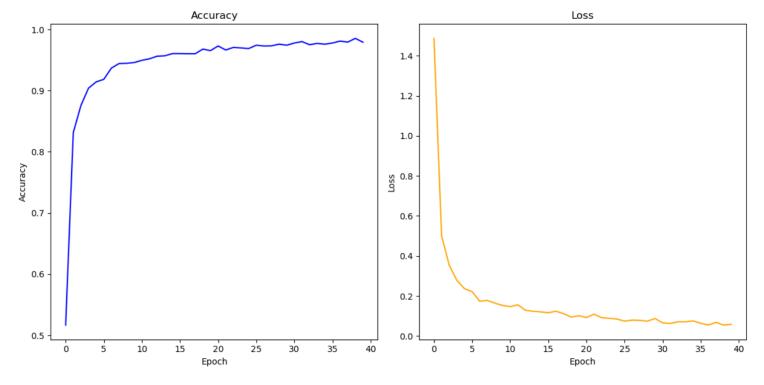




#### • Restnet101



InceptionResNetV2





submission\_ori.csv

0.97481

0.97481

### Result - Model Comparison

	DenseNet121	DenseNet169	DenseNet201	Efficientnet b0	Efficientnet b4
F1-score	0.9785	0.9798	0.9811	0.9798	0.9785

	MobileNetV3 Small	MobileNetV3 Large	Restnet50	Restnet101	InceptionResN etV2
F1-score	0.9521	0.9760	0.9710	0.9496	0.9748

- After comparing the performance of the listed models, it is found that DenseNet201 achieves the highest F1 score.
- This may be due to its larger model capacity, dense connectivity, specific architectural features, and adaptability to the characteristics of the evaluated dataset.

### DenseNet201 Architecture

Model: "model"						
Layer (type)	Output Shape	Param #				
image_input (InputLayer)	[(None, 299, 299, 3)]	0				
densenet201 (Functional)	(None, None, None, 1920)	18321984				
flatten (Flatten)	(None, 155520)	0				
dense (Dense)	(None, 128)	19906688				
dense_1 (Dense)	(None, 12)	1548 				
Total params: 38,230,220 Trainable params: 38,001,164 Non-trainable params: 229,056						