

Taiwan Deep Travel - Deep Learning-Based Recommendation APP for Taiwan Tourist Attractions

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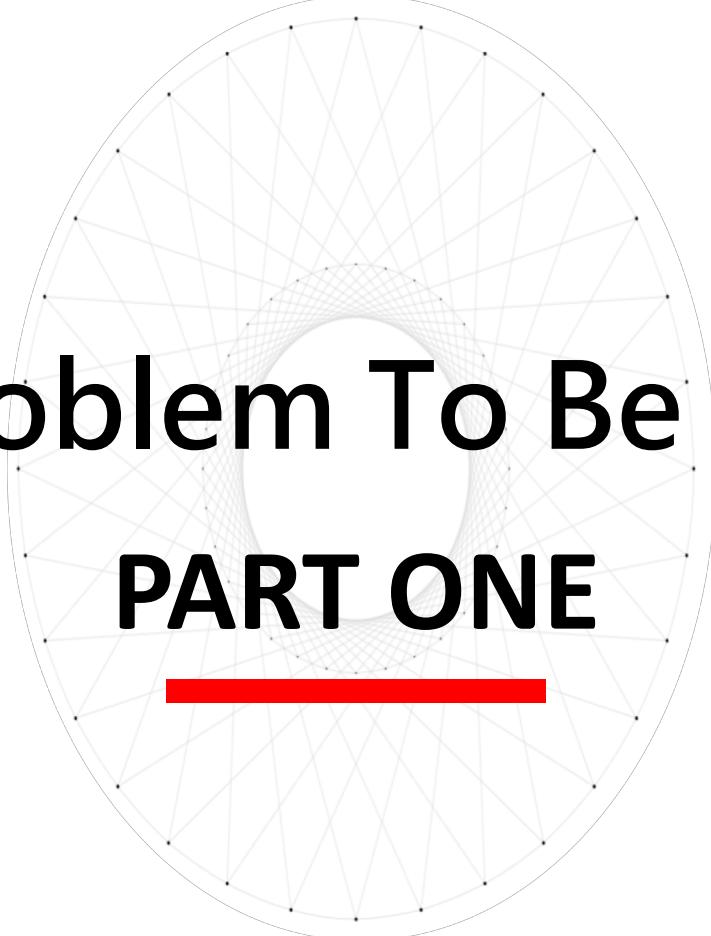
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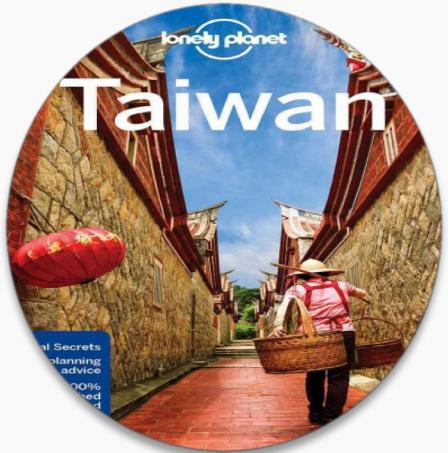




The Problem To Be Solved

PART ONE

Research Motivation



Taiwan is rich in beautiful natural and cultural scenes



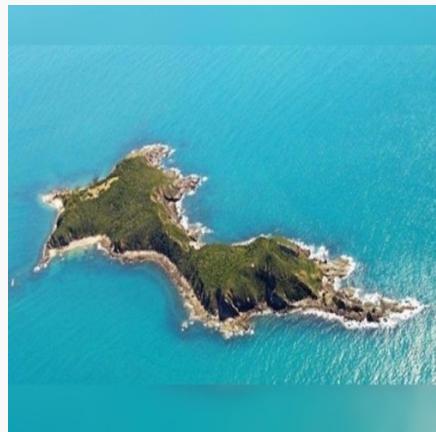
Modern people live a busy life and have little time to search for travel information



The popularity of smartphones

Operating Mode

— Import Image



Island



TOP 1
Danger rock island



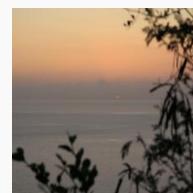
TOP 2
Beinan Creek
Rafting



TOP 3
Waisanding
Sand Bar



TOP 4
Wangankou
Beach



TOP 5
Guanshan
Xizhao

Operating Mode

— Take Photos



Xiaonanmen



Building Façade
Temple



TOP 1
Amitabha
Temple



TOP 2
Tower of
Light



TOP 3
Chiang Kai-shek
Memorial Hall



TOP 4
Tree-footed
church



TOP 5
Ci En Pagoda

Accurately meet the needs of users

Mountain
Mountain Path Forest
Sunset Canyon
Lawn



Focus on the recommendation of tourist attractions



Convenient user experience





Model Dataset— Source

MIT CSAIL 「Place - 365 Standard」 dataset

- 365 scenes
- 256x256x3 pixels

Training Dataset — Each scene has 5,000 images

Validation Dataset — Each scene has 100 images



Model Dataset— Filtered

- 20 scenes
- 256x256x3 pixels

Training Dataset — Each scene has 5,000 images

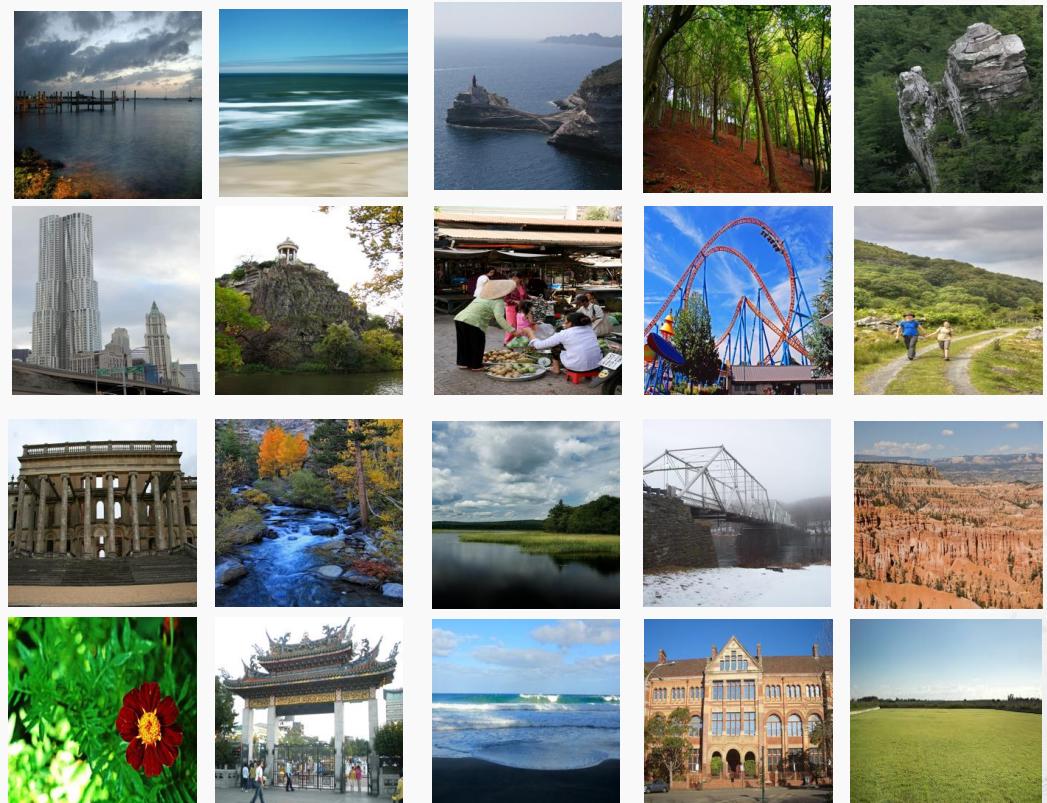
Validation Dataset — Each scene has 100 images



Model Dataset— 20 Scenes

Amusement Park
Botanical Garden
Building Façade
Mountain Path
Skyscraper
Canyon
Cliff
Creek
Forest
Islet

Pier
Temple
Lawn
Market
Mountain
Museum
Lake
Beach
Bridge
Ocean



Taiwan recommended tourist attractions dataset— Source

Tourism Bureau, Ministry of Transportation and Communications Government Information Open Platform : “Attractions - Sightseeing Information Dataset” - covers 4,809 scenic spots

Taipei City Government Information Open Platform : “Taipei City, Taipei Tourism Website - Attractions Data” - covers 200 scenic spots

- Locations all over Taiwan
- Each scenic spot has an image and a text introduction about the scenic spot



Recommended tourist attractions dataset— Filtered

Recommended tourist attractions dataset

- Covers 895 scenic spots
- Locations all over Taiwan
- Each scenic spot has an image and a text introduction about the scenic spot



Recommended tourist attractions dataset— Filter Condition

Artwork



People



Tourist center



Experience in nature



Data usage

1. Input 20 different scenes to the model for training

- Each scene has 5,000 images

2. Input the attraction recommendation data set into the model

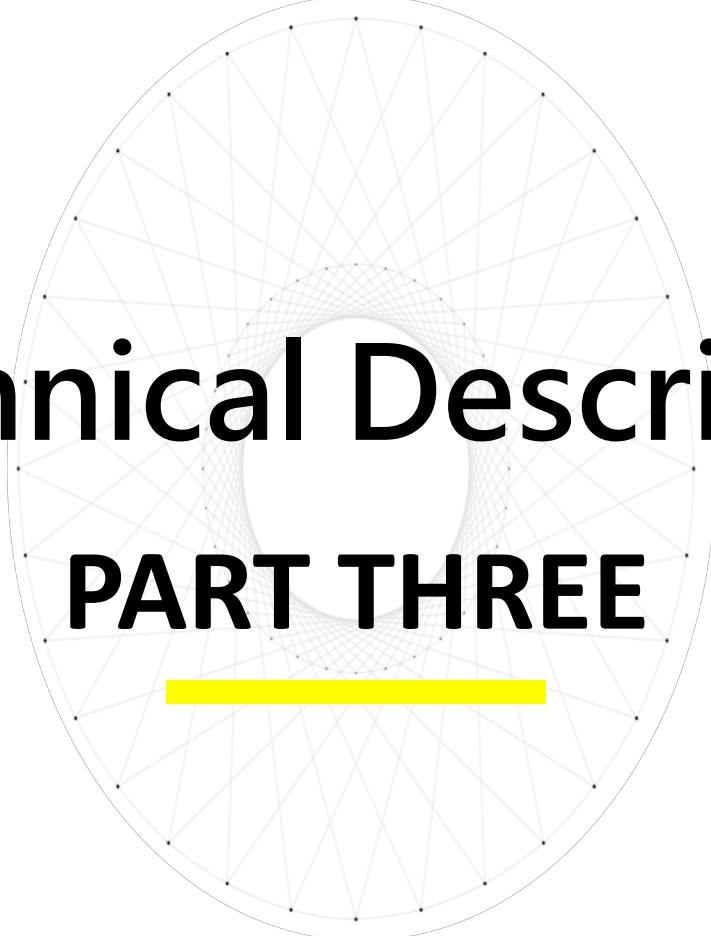
- Produces 895 1x20 vectors
- Each value in the vector represents the probability of each scene in the scenic spot predicted by the model

3. The user inputs an image of the attraction into the model

- Produces a 1x20 vectors

4. Recommend top five scenic spots with highest similarity to users

- Cosine similarity was compared 895 times between the vector generated by the image input by the user and the vector generated by the scenic spots



Technical Description

PART THREE

[Yellow horizontal bar]

Deep Learning Tools

Keras



- Written in Python code
- Open source application running on TensorFlow
- Various neural network layers are built in, such as CNN and RNN
- Through the back end of the TensorFlow engine, can be run on the GPU
- Easier to understand and more modular than TensorFlow

Google Cloud Platform

Compute Engine



Google Cloud Platform

Create virtual machine instances



Remote control virtual machines



Install CUDA and TensorFlow, Keras libraries

Virtual Machine standard

- Two vCPU, 26 GB Memory
- Operating System: Windows Server 2016
256GB SSD
- GPU standard : NVIDIA Tesla K80
24GB Memory

Augment Original Validation Dataset

Keras Image Data Augmentation



Original Validation Data

- 20 scenes
 - 100 images each scene
- 
- 20 scenes
 - **200** images each scene

Augmented Validation Data

- 20 scenes
- 100 images each scenes

Augmented Test Data

- 20 scenes
- 100 images each scenes

Modeling Procedure

- 
1. Determine suitable Transfer Learning model
 2. Determine Optimizer
 3. Determine Batch Normalization (BN) & Activation Function
 4. Determine to use dynamic Learning Rate or static Learning Rate
 5. Adjust Hyperparameter (Avoid Overfitting)
 6. Test model and evaluates the final results

Determine suitable Transfer Learning model

- Use **Transfer Learning** to train each model
- The result of training all layers **5 epochs**

Model	Loss	Top-1 Acc	Top-3 Acc	Val loss	Top-1 Val_acc	Top-3 Val_acc	Size	Parameters	Depth	Time
VGG16 (2014 ILSVRC)	0.8538	0.7390	0.9250	0.9885	0.6760	0.9160	266 MB	23,238,356	27	47 ms/step
ResNet50 (2015 ILSVRC)	0.3632	0.8760	0.9800	1.0926	0.7060	0.9140	417 MB	36,435,476	180	40 ms/step
MobileNet (2017 Google)	0.9317	0.7000	0.9170	1.0584	0.6940	0.9030	110 MB	9,654,100	92	17 ms/step
DenseNet121 (2017 CVPR)	0.7228	0.7650	0.9350	0.9701	0.7030	0.9150	154 MB	13,462,740	431	38 ms/step
NASNetMobile (2018)	0.7656	0.7650	0.9320	0.8591	0.7320	0.9250	128 MB	10,895,656	774	34 ms/step
MobileNetV2 (2018 Google)	0.8226	0.7480	0.9240	0.9828	0.6780	0.9120	118 MB	10,288,852	160	20 ms/step

Determine Optimizer

- Compare Adam & Adafactor
- The result of training all layers 5 epochs
- The validation data is 2000 images of augmented validation data

Model	Loss	Top-1 Acc	Top-3 Acc	Val loss	Top-1 Val_acc	Top-3 Val_acc	Size	Parameters	Depth	Time
MobileNet with Adam (2015)	0.9624	0.6760	0.9030	1.5239	0.5565	0.8020	110 MB	9,654,100	92	22 ms/step
MobileNet with Adafactor (2018)	3.2579	0.0510	0.1300	3.4619	0.0500	0.1500	37 MB	9,654,100	92	21 ms/step

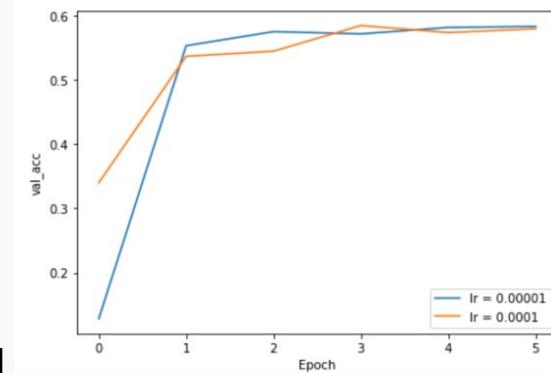
Determine Batch Normalization (BN) & Activation Function

- Compare ReLU with BN, ReLU without BN and SeLU with BN
- The result of training all layers 5 epochs
- The validation data is 2000 images of augmented validation data

Model	Loss	Top-1 Acc	Top-3 Acc	Val loss	Top-1 Val_acc	Top-3 Val_acc	Size	Parameters	Depth	Time
ReLU with BN	0.8410	0.7250	0.9230	1.3066	0.5835	0.8315	117.7 MB	9,855,316	95	21 ms/step
ReLU without BN	0.9624	0.6760	0.9030	1.5239	0.5565	0.8020	110 MB	9,654,100	92	22 ms/step
SeLU with BN (2015)	0.7213	0.7840	0.9390	1.3839	0.5695	0.8215	117.7 MB	9,855,316	94	20 ms/step

Determine to use dynamic Learning Rate or static Learning Rate

- Compare dynamic and static Learning Rate (LR) results
- The result of training all layers 5 epochs
- The validation data is 2000 images of augmented validation data



Model	Loss	Top-1 Acc	Top-3 Acc	Val loss	Top-1 Val_acc	Top-3 Val_acc	Size	Parameters	Depth	Time
MobileNet (LR = 0.00002)	0.8410	0.7250	0.9230	1.3066	0.5835	0.8315	117.7 MB	9,855,316	95	21 ms/step
MobileNet (LR = 0.0002)	0.7651	0.7460	0.9360	1.3139	0.5800	0.8400	112 MB	9,855,316	95	21 ms/step
MobileNet (LR = 0.002)	1.1672	0.6360	0.8590	1.7032	0.4620	0.7455	112 MB	9,855,316	95	22 ms/step
MobileNet (LR = 0.002 _ 0.00002)	1.0588	0.6670	0.8770	1.4206	0.5540	0.7965	112 MB	9,855,316	95	21 ms/step

Adjust Hyperparameter

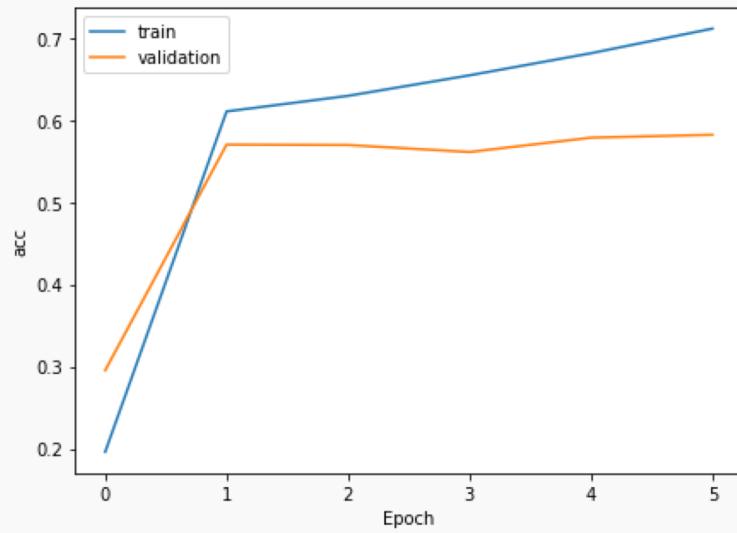
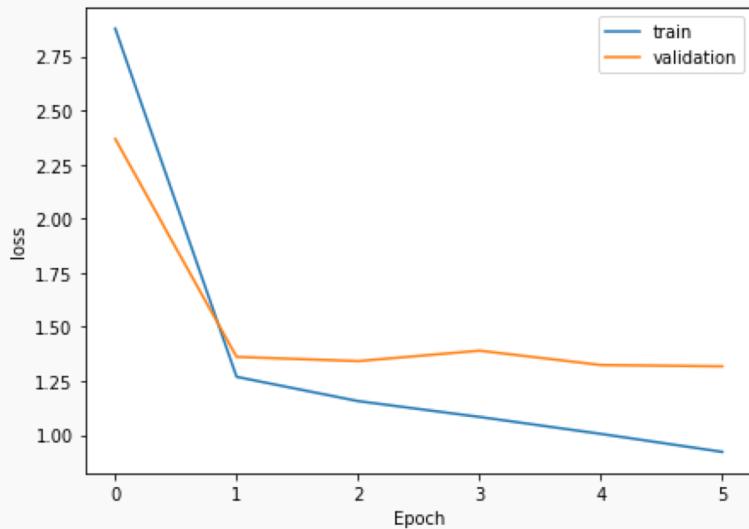
- The result of training all layers 5 epochs
- The validation data is 2000 images of augmented validation data

Depth	Time
94	21 ms/step
94	22 ms/step
94	21 ms/step
94	21 ms/step
94	21 ms/step
94	22 ms/step
94	21 ms/step
94	22 ms/step

kernel_regularizer activity_regularizer	Dropout	Dense	Loss	Top-1 Acc	Val loss	Top-1 Val_acc	Size	Parameters
0	0.5	128	0.9438	0.6880	1.2661	0.5840	112 MB	9,855,316
0	0.5	256	0.8881	0.6970	1.2515	0.5910	185 MB	16,281,044
0	0.2	128	0.8632	0.7090	1.2756	0.5700	112 MB	9,855,316
0	0.2	256	0.9959	0.6630	1.2755	0.5780	185 MB	16,281,044
0.01	0.5	128	4.4488	0.0500	4.4374	0.0510	117.7 MB	9,855,316
0.01	0.5	256	5.6078	0.1880	5.3630	0.2795	194.8 MB	16,281,044
0.01	0.2	128	3.6754	0.3710	3.6977	0.3865	117.7 MB	9,855,316
0.01	0.2	256	5.4007	0.4580	5.4992	0.4165	194.8 MB	16,281,044

Model Performance Evaluation

- The result of training all layers 5 epochs
- The validation data is 2000 images of augmented validation data



Model	Loss	Top-1 Acc	Top-3 Acc	Val_loss	Top-1 Val_acc	Top-3 Val_acc
5 epochs	0.9600	0.6900	0.9020	1.2687	0.5835	0.8515

Model Performance Evaluation

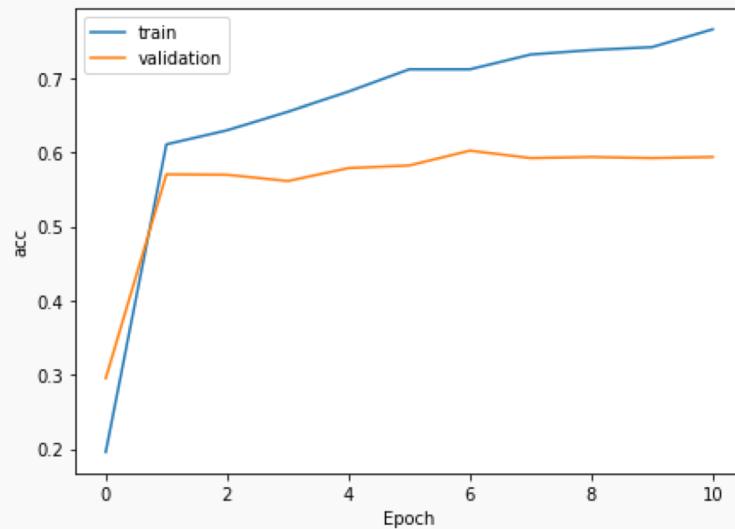
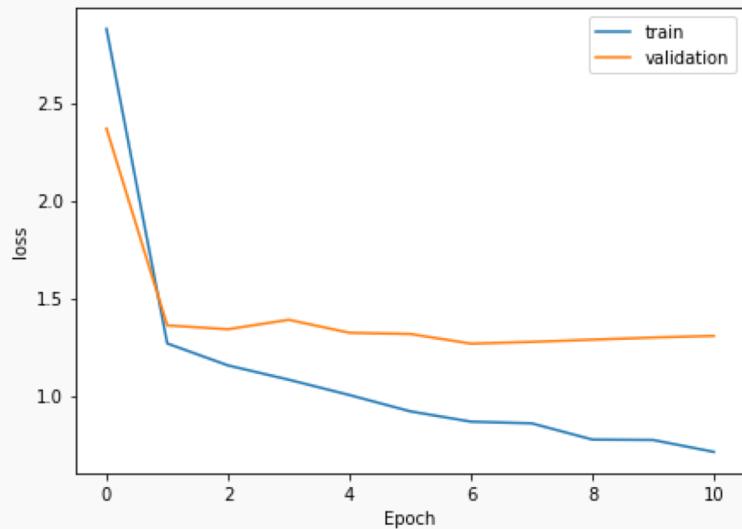
- The result of training all layers 5 epochs
- The test data is 2000 images of augmented validation data

Test Data

Model	Loss	Top-1 Acc	Top-3 Acc
5 epochs	1.2499	0.5970	0.8510

Model Performance Evaluation

- The result of training all layers 10 epochs
- First 5 epochs' LR are 0.0002, last 5 epochs' LR are 0.00002
- The validation data is 2000 images of augmented validation data



Model	Loss	Top-1 Acc	Top-3 Acc	Val_loss	Top-1 Val_acc	Top-3 Val_acc
5_5 epochs	0.8040	0.7380	0.9360	1.2467	0.6075	0.8535

Model Performance Evaluation

- The result of training all layers 10 epochs
- The test data is 2000 images of augmented validation data

Test Data

Model	Loss	Top-1 Acc	Top-3 Acc
5_5 epochs	1.2206	0.6140	0.8465

Recommendation System

Cosine similarity
after correction

$$\frac{\sum_{i=1}^{20} [(A_i - \mu_A) \times (B_i - \mu_B)]}{\sqrt{\sum_{i=1}^{20} (A_i - \mu_A)^2} \sqrt{\sum_{i=1}^{20} (B_i - \mu_B)^2}}$$

Cosine similarity

$$\frac{\sum_{i=1}^{20} (A_i \times B_i)}{\sqrt{\sum_{i=1}^{20} (A_i)^2} \sqrt{\sum_{i=1}^{20} (B_i)^2}}$$

Cosine similarity:
0.9967668105098986



Taipei 101



Xinbeitou Hot Spring Area



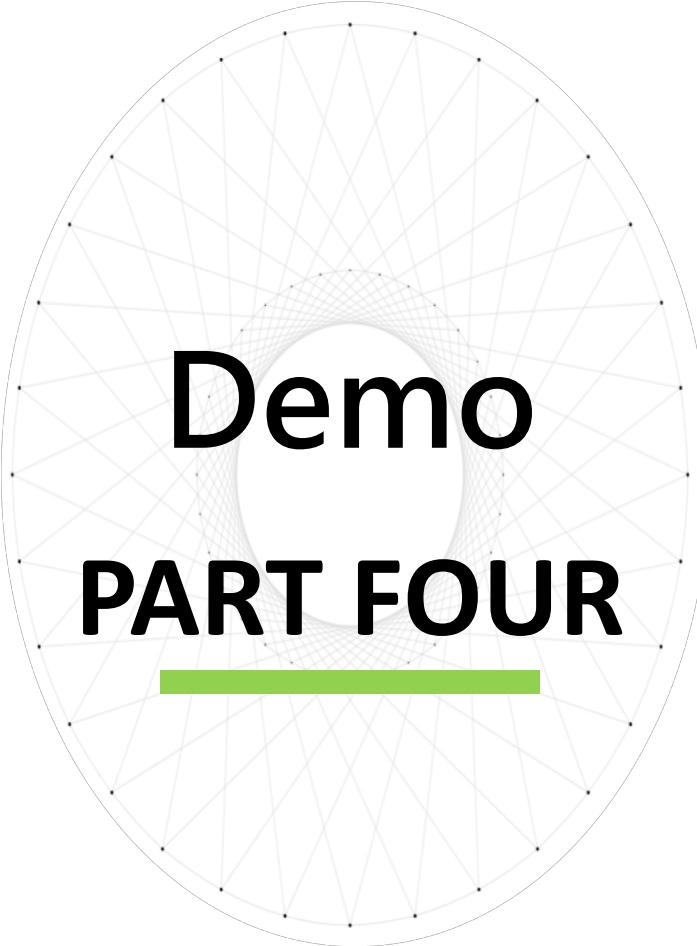
Cosine similarity:
-0.19923189416031017

Integrated into mobile devices

Core ML 2.0

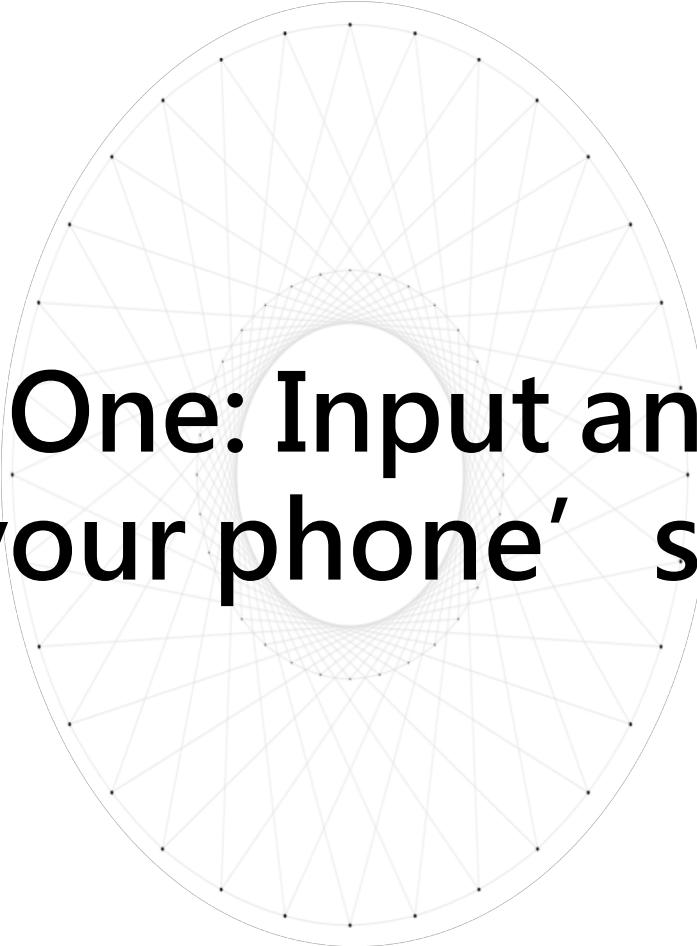
- Machine learning framework introduced by Apple at WWDC 2018
- Empowers the iOS app developers to integrate machine learning technology into their apps



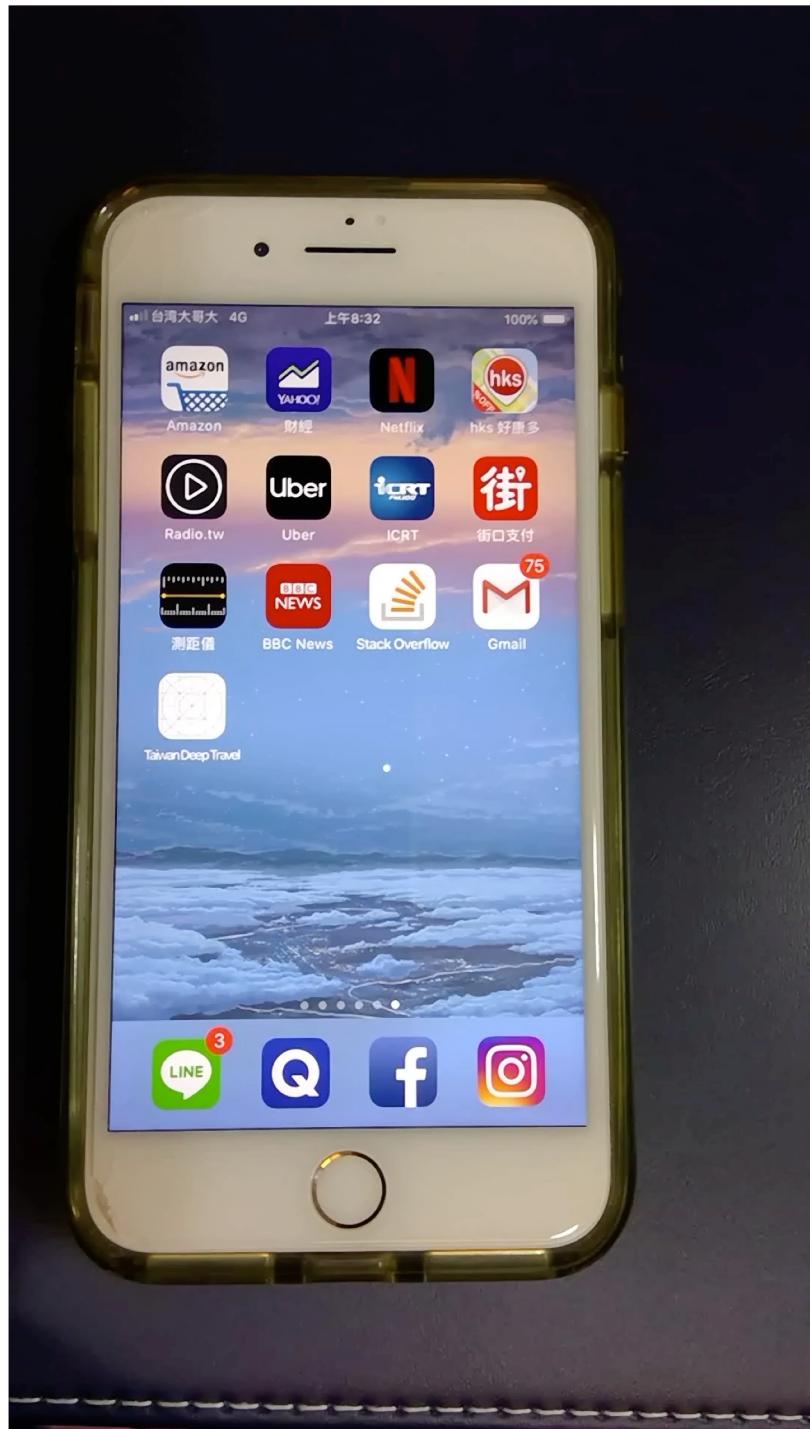


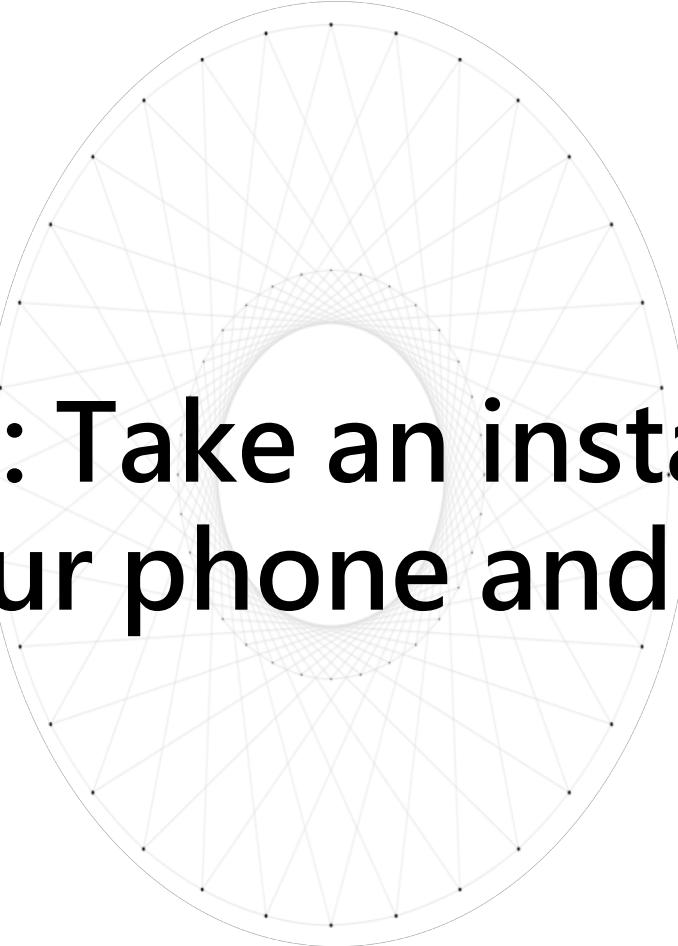
Demo

PART FOUR

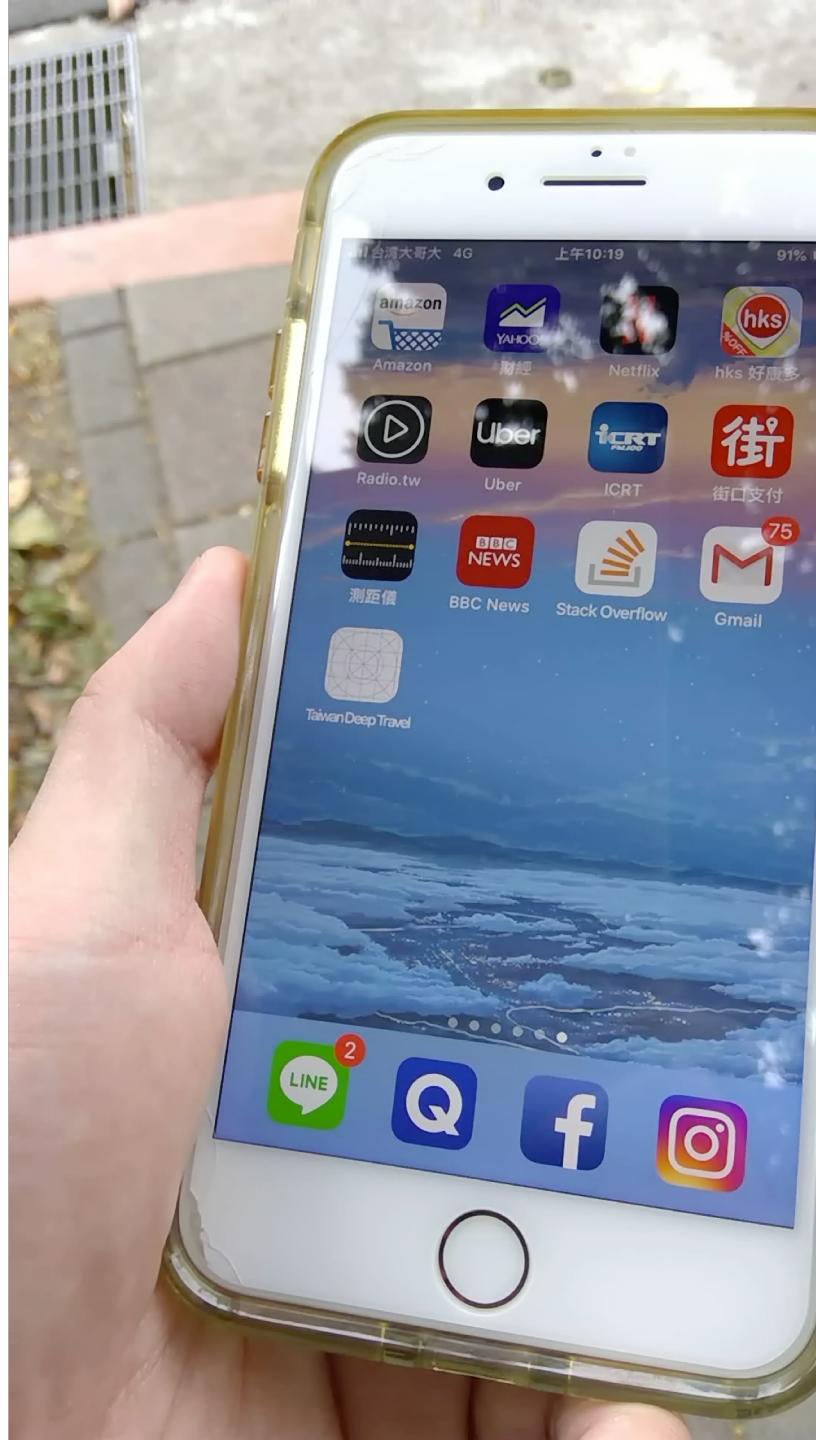


**Video One: Input an image
from your phone' s library**





**Video Two: Take an instant picture
with your phone and input it**



System Stability

Normal



TOP 1



TOP 2



TOP 3

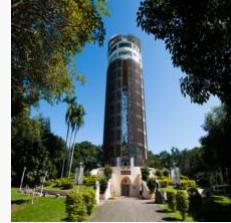


TOP 4



TOP 5

Rotate 45 Degrees



TOP 1



TOP 2



TOP 3



TOP 4



TOP 5

System Stability

Normal



TOP 1



TOP 2



TOP 3



TOP 4



TOP 5

Extremely Bright



TOP 1



TOP 2



TOP 3



TOP 4



TOP 5

System Stability

Normal



TOP 1



TOP 2



TOP 3



TOP 4



TOP 5

Twisted



TOP 1



TOP 2



TOP 3



TOP 4



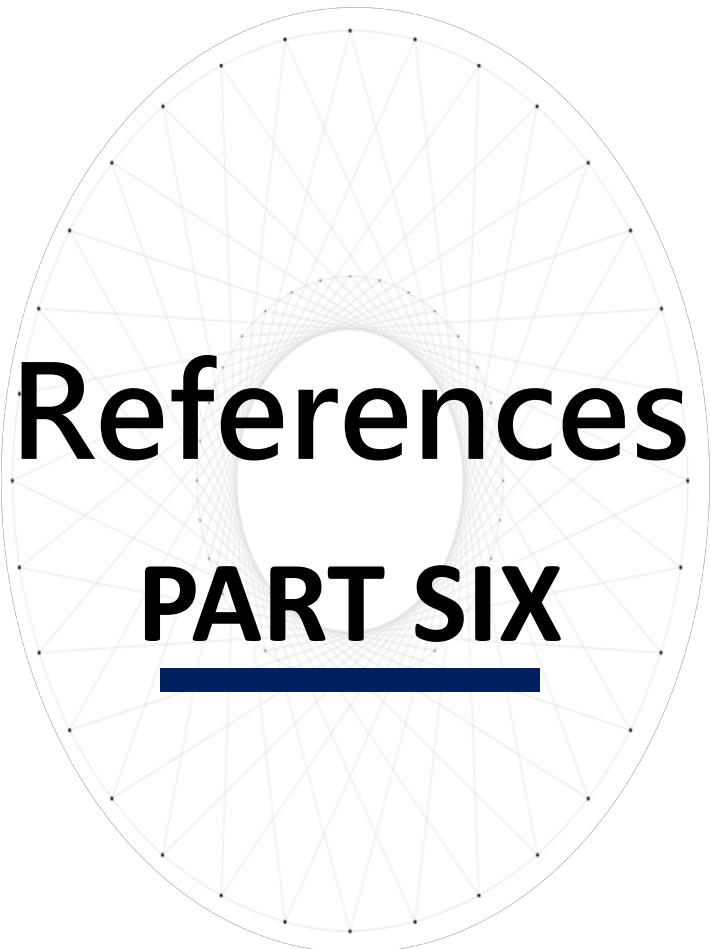
TOP 5



Research Prospect

PART FIVE

- 
- 
1. Use more scenes to train the model to improve its ability to identify scenes and the accuracy of recommended scenic spots
 2. Augment the training dataset for model training to improve the accuracy of validation and test data
 3. Add more hyperparameter combinations for comparison
 4. Use TensorBoard to visualize neural networks
 5. Use more Open Data to augment recommended attractions
 6. Extend the system to the Android mobile operating system
 7. Collect users' comments and suggestions on the APP



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PART SIX

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