

UNIVERSITY OF ZAGREB  
FACULTY OF ELECTRICAL ENGINEERING AND COMPUTING  
DIGITAL IMAGE PROCESSING AND ANALYSIS

# VPR

## VISUAL PLACE RECOGNITION

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# VISUAL PLACE RECOGNITION

- various domains of application
- using visual information
- role of deep learning and computer vision
- CNN models and specialized VPR models

# OVERVIEW OF METHODS

## ResNet

- „residual” connections
- output directly passed to the next layer

## NetVLAD

- efficient aggregation of local descriptors into a powerful global image representation

## DenseNet

- connecting each layer to all previous layers
- richer set of features

## TransVPR

- transformer architecture
- frequent changes in visual information

# RELATED WORK

- comparison of neural networks and transformers
- transformers proved to be much better
  - mechanism to process all parts of the input data simultaneously
- sequence matching
- sequential descriptors
- global descriptors

# USED MATERIALS AND METHODS

**1.**

**DATASET**

**2.**

**IMPLEMENTED  
SOLUTION**

**3.**

**RESULTS**

# DATASET – GSV-CITIES

- *Neurocomputing 2022: GSV-Cities: Toward Appropriate Supervised Visual Place Recognition*
- 530 000 images from over 62 000 different locations
- common image dataset problems:
  - geographical coverage
  - precision of reference data
  - perceptual diversity

# GSV-CITIES – BANGKOK



# IMPLEMENTED SOLUTION

1

DATA  
PREPARATION

2

SELECTION  
OF OPTIMAL  
HYPERPARAMETERS

3

MODEL  
DEFINITION

4

MODEL  
TRAINING

5

MODEL  
EVALUATION



# 1.

## DATA PREPARATION

- CustomImageDataset
- image transformations:
  - resizing to 224x224 pixels
  - random horizontal flipping
  - normalization
  - conversion to tensors

# 2.

## SELECTION OF OPTIMAL HYPERPARAMETERS

- different learning rates
- different optimization algorithms (SGD and Adam)
- different batch sizes
- 5% of the data

### **3. MODEL DEFINITION**

- pre-trained ResNet18 model

### **4. MODEL TRAINING**

- cross-validation and EarlyStopping
- 50 epochs

### **5. MODEL EVALUATION**

- overall loss and accuracy on training, validation and test sets

# RESULTS

## TRAINING

LOSS  
**0.0308**

ACCURACY  
**99.20%**

## VALIDATION

LOSS  
**0.2678**

ACCURACY  
**91.49%**

## TEST

LOSS  
**0.2693**

ACCURACY  
**91.53%**

# CONCLUSION

- significant enhancement of VPR system capabilities
- high efficiency demonstrated on the GSV-Cities dataset
- potential for better results with further research and refinement of transformer-based models
- wide range of possible applications of VPR

# VPR

## VISUAL PLACE RECOGNITION

# THANK YOU! FOR YOUR ATTENTION

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# LITERATURE

- [1] R. Mereu, G. Trivigno, G. Berton, C. Masone, B. Caputo, "**Learning Sequential Descriptors for Sequence-based Visual Place Recognition**", 2022.
- [2] S. Sundar Kannan, B. Min, "**PlaceFormer: Transformer-based Visual Place Recognition using Multi-Scale Patch Selection and Fusion**", 2024.
- [3] A. Ali-bey, B. Chaib-draa, P. Giguère "**GSV-CITIES: Toward Appropriate Supervised Visual Place Recognition**"