

2023 Internship Group Presentations

Berlin, 31.08.2023



Topic Overview / Agenda

Belntelli

- 1. TP2: AI Middleware Edge Cloud
- 2. TP3: Extrinsic Camera and Lidar Calibration
- 3. TP3: Localization by Visual Place Recognition
- 4. TP3: Free Parking Space Detection
- 5. TP4: Smart Parking Recommendation
- 6. TP4: AI Platform and Tools

COBRA-5G

1. Pod/Container-Monitoring in K8s Clusters

Go-KI

- 1. Smart Fridge Object Detection
- 2. Ground Assistance System Voice Interaction



Localization by Visual Place Recognition

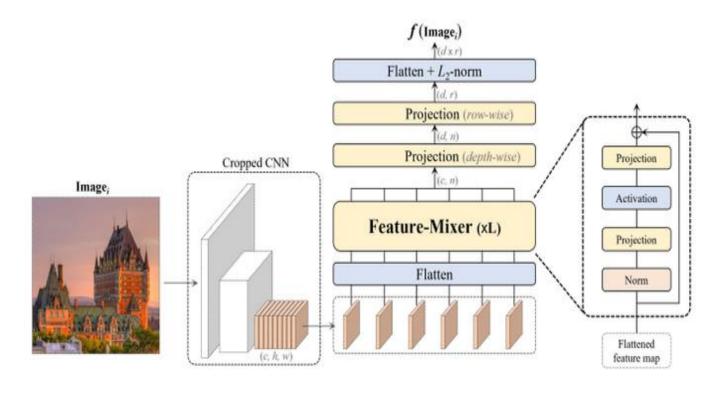
Group member:

Ahmet Alperen Güngör

- To improve the initialization of the localization algorithm^[1]
- Recognize environment by camera data using a NN-based regression model (MixVPR)^[2]
- Locate where our vehicle is and output GNSS coordinates.

Supervisor:

Ketan Motlag / Srinivas Ravuri





Literature Review - Manual Approaches:

Challenges:[3][4]

1)Viewpoint Change

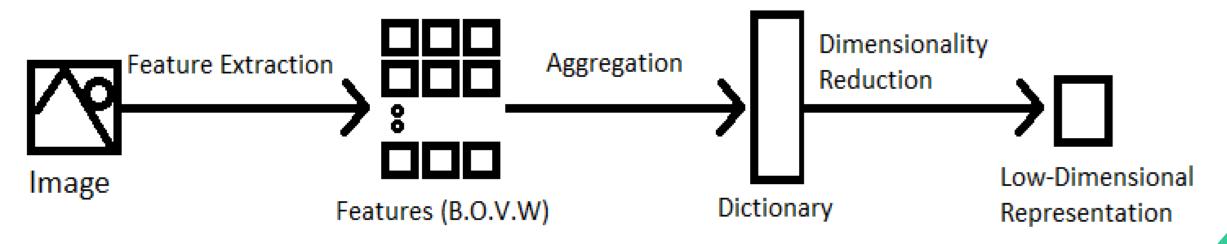
2)Perceptual Aliasing

3)Appearance

Change

Process:

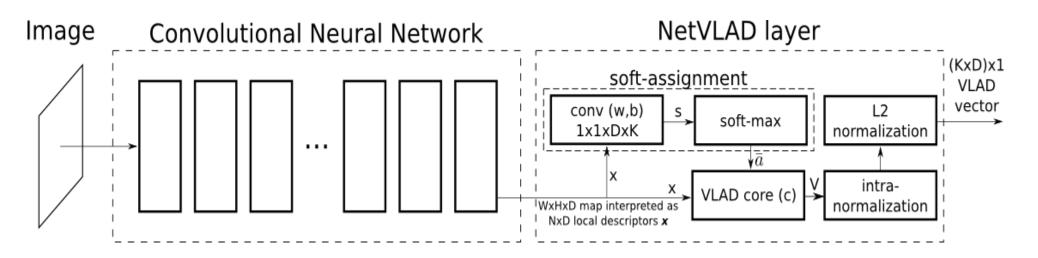
- Local (SIFT^[5]/SURF^[6]) or Global (GIST^[7]) Feature Extraction
- Aggregation (VLAD^{[8][9]})
- Dimensionality Reduction (PCA)



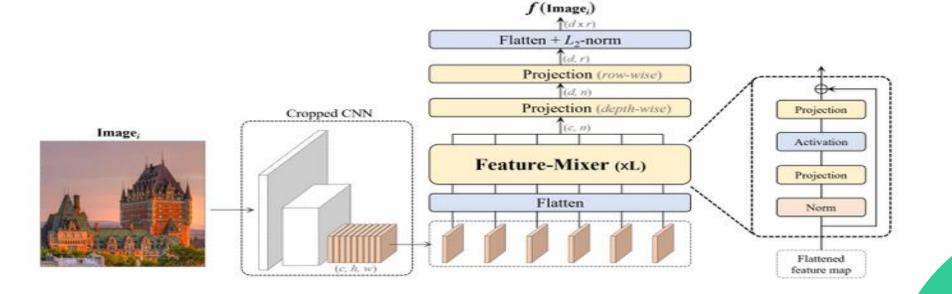


End-to-End Deep Learning Approaches:

NetVLAD[10][11][12]:



MixVPR:

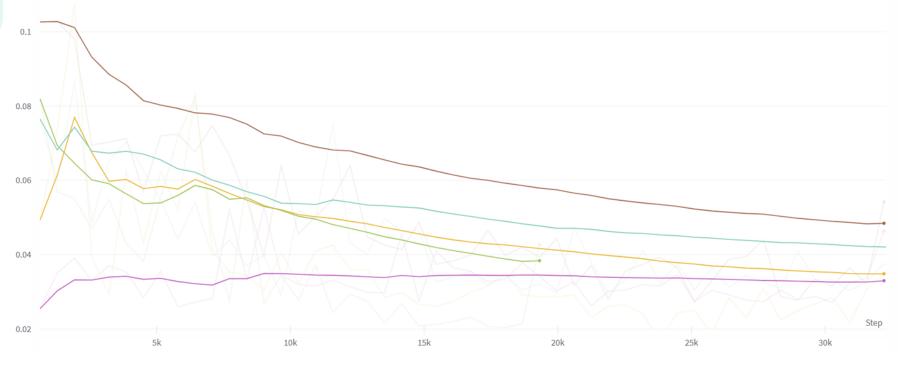


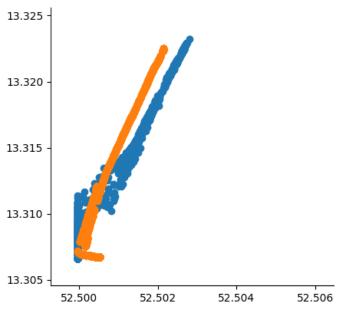
val_loss - mixvpr-mixer-2-beintelli - mixvpr_beintelli_weather_aug - MIXVPR-4-FM-RESNET50 - MIXVPR-1-FM-RESNET50 - NETVLAD-RESNET50

BeIntelli TP3

Results:







examples



pred: [52.50215322 13.32168739], target: [52.5019887 13.32144691]



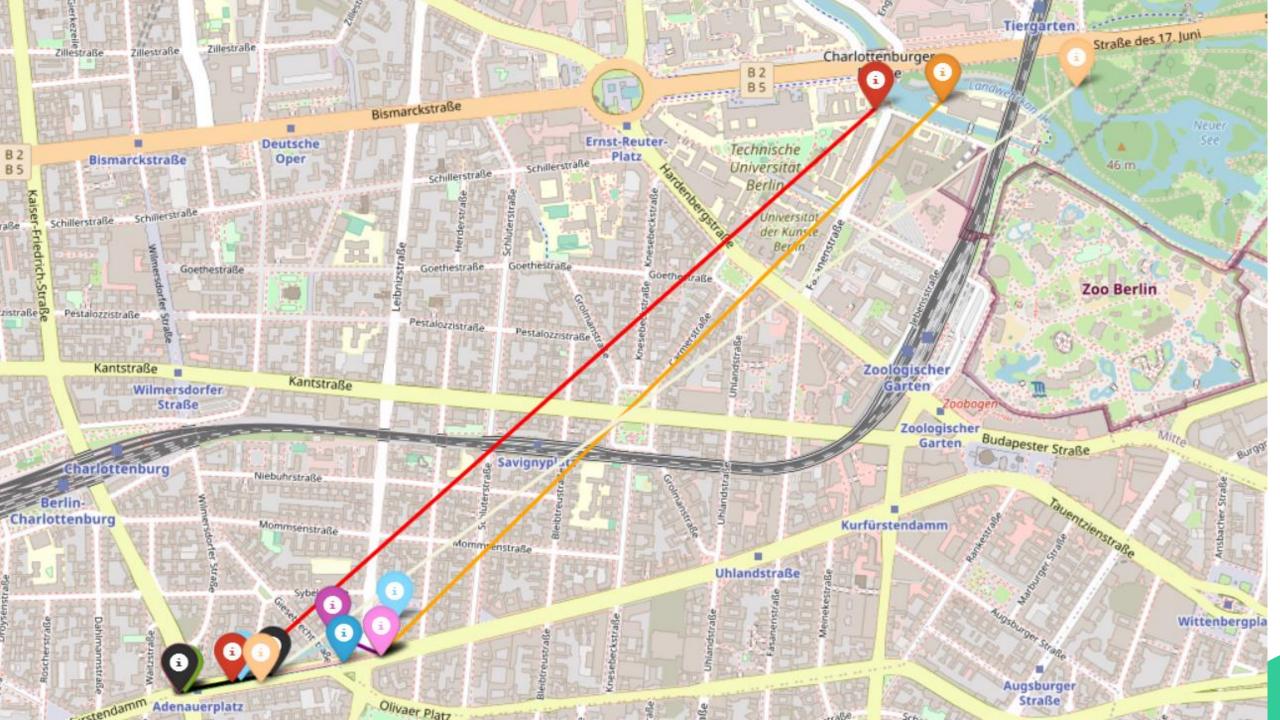
pred: [52.50109027 13.31465805], target: [52.50082308 13.31404815]



pred: [52.50203281 13.3214075], target: [52.50212322 13.32229852]



pred: [52.50136279 13.31668175], target: [52.50134043 13.31736934]





Future Works:

- Increasing accuracy (less loss):
 - More hyperparameter tuning and testing with current MixVPR model
 - O Instead of a basic CNN feature extraction, a more complex one such as attention-based transformers^[13]
 - Improvements in feature-mixer layer a more complex and accurate layer if possible
- Improved robustness to temporary and repeating objects like cars, humans and trees/buildings (ex. Zoo road and rural areas)^{[14][15][16]}
- Combining RGB image data with lidar data^[17]

BeIntelli TP3

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