

Automated Parking Space Detection System

A comprehensive computer vision system for detecting parking space occupancy using machine learning. The system supports training on the PKLot dataset and provides real-time detection capabilities through camera feeds.

Features

- **Dataset Support:** Compatible with PKLot dataset format
- **Machine Learning:** SVM classifier with feature engineering
- **Real-time Detection:** Live camera feed processing
- **Interactive ROI Selection:** Click-and-drag parking space selection
- **Performance Evaluation:** Comprehensive metrics and visualizations
- **Model Persistence:** Save and load trained models
- **Visualization:** Annotated output images with bounding boxes

Requirements

- Python 3.7+
- OpenCV 4.x
- Scikit-learn

- NumPy, Matplotlib, Seaborn
- PKLot dataset (optional, for training)

Installation

Option 1: Local Installation

1. Clone the repository:

```
bash
```

```
git clone <repository-url>  
cd parking-detection-system
```

2. Install dependencies:

```
bash
```

```
pip install -r requirements.txt
```

Option 2: Docker Installation

1. Build the Docker image:

```
bash
```

```
docker build -t parking-detector .
```

2. Run the container:

```
bash
```

```
# For training (mount your dataset)
```

```
docker run -v /path/to/dataset:/app/dataset -v /path/to/output:/app/output parking-detect
```

```
# For real-time detection (with camera access)
```

```
docker run --device=/dev/video0 -e DISPLAY=$DISPLAY -v /tmp/.X11-unix:/tmp/.X11-unix park:
```

Dataset Preparation

PKLot Dataset Structure

```
dataset/  
├─ PKLot/  
│   ├─ PUCPR/  
│   │   ├─ Cloudy/  
│   │   │   └─ 2012-09-12/  
│   │   │       ├─ image001.jpg  
│   │   │       ├─ image001.xml  
│   │   │       └─ ...  
│   │   └─ ...  
│   └─ ...  
└─ ...
```

The XML annotations should contain rotated rectangle coordinates for each parking space with occupancy labels.

Usage

Training Mode

Train a new model on the PKLot dataset:

```
bash
```

```
python parking_detection_system.py --mode train --dataset /path/to/pklot/dataset --model 1
```



Parameters:

- `--dataset`: Path to PKLot dataset root directory
- `--model`: Output path for trained model (default: parking_model.pkl)
- `--output`: Directory for evaluation results and visualizations

Testing Mode

Evaluate a trained model on test data:

```
bash
```

```
python parking_detection_system.py --mode test --dataset /path/to/test/dataset --model my_
```



Real-time Detection Mode

Run real-time detection using a camera:

```
bash
```

```
python parking_detection_system.py --mode realtime --model my_model.pkl --camera 0
```

Parameters:

- `--camera`: Camera device ID (default: 0)
- `--model`: Path to trained model file

Interactive Controls

- **Click and drag**: Select parking space regions (ROIs)
- **'r' key**: Reset all selected ROIs
- **'q' key**: Quit the application

API Usage

Basic Usage

python

```
from parking_detection_system import ParkingSpaceDetector

# Initialize detector
detector = ParkingSpaceDetector()

# Training
image_paths, rois, labels = detector.load_pk1ot_dataset('/path/to/dataset')
detector.train_classifier(image_paths, rois, labels)
detector.save_model('my_model.pkl')

# Prediction
detector.load_model('my_model.pkl')
predictions, probabilities = detector.predict_spaces(image, parking_rois)
```

Real-time Detection

python

```
from parking_detection_system import ParkingSpaceDetector, RealTimeParkingDetector

# Load trained detector
detector = ParkingSpaceDetector()
detector.load_model('parking_model.pkl')

# Initialize real-time detector
rt_detector = RealTimeParkingDetector(detector)

# Set predefined parking regions (optional)
parking_rois = [[x1, y1, x2, y2], ...] # List of bounding boxes
rt_detector.set_parking_regions(parking_rois)

# Start detection
rt_detector.detect_from_camera(camera_id=0)
```

Model Architecture

Feature Extraction

- **Statistical Features:** Mean intensity, variance, texture
- **Edge Features:** Sobel edge detection for boundary information
- **Preprocessing:** Histogram equalization, Gaussian blur

Classification

- **Algorithm:** Support Vector Machine (SVM) with RBF kernel
- **Features:** 4-dimensional feature vector per parking space
- **Scaling:** StandardScaler for feature normalization
- **Class Balance:** Weighted classes to handle imbalanced data

Output Files

Training Output

- `parking_model.pkl`: Trained classifier and scaler
- `confusion_matrix.png`: Classification performance visualization
- `empty_spaces_comparison.png`: Actual vs predicted empty spaces
- `*_annotated.jpg`: Annotated images with detection results

Performance Metrics

- **Precision:** Positive predictive value
- **Recall:** Sensitivity/True positive rate
- **F1-Score:** Harmonic mean of precision and recall
- **Accuracy:** Overall classification accuracy

Configuration

Model Parameters

You can modify the SVM parameters in the `train_classifier` method:

python

```
self.classifier = SVC(  
    kernel='rbf',          # Kernel type  
    C=1.0,                 # Regularization parameter  
    class_weight='balanced', # Handle class imbalance  
    probability=True,      # Enable probability estimates  
    random_state=42        # For reproducibility  
)
```

Feature Extraction Parameters

Adjust preprocessing parameters in the `preprocess_image` method:

python

```
# Gaussian blur kernel size  
blurred = cv2.GaussianBlur(equalized, (5, 5), 0)  
  
# Sobel kernel size for edge detection  
sobelx = cv2.Sobel(region, cv2.CV_64F, 1, 0, ksize=3)
```

Troubleshooting

Common Issues

1. Camera not detected:

```
bash
```

```
# List available cameras
```

```
ls /dev/video*
```

```
# Try different camera IDs
```

```
python parking_detection_system.py --mode realtime --camera 1
```

2. XML parsing errors:

- Ensure XML files match the expected PKLot format
- Check file permissions and paths

3. Memory issues with large datasets:

- Process dataset in batches
- Reduce image resolution if needed

4. Poor detection accuracy:

- Collect more training data
- Adjust feature extraction parameters
- Try different SVM parameters

Docker Issues

1. Camera access in Docker:

```
bash
```

```
# Add camera device access
```

```
docker run --device=/dev/video0 ...
```

2. Display issues:

```
bash
```

```
# Enable X11 forwarding
```

```
xhost +local:docker
```

```
docker run -e DISPLAY=$DISPLAY -v /tmp/.X11-unix:/tmp/.X11-unix ...
```

Performance Optimization

Speed Improvements

- Use smaller ROI regions
- Reduce image resolution
- Implement multi-threading for batch processing
- Consider using faster classifiers (e.g., Random Forest)

Accuracy Improvements

- Collect more diverse training data
- Implement data augmentation
- Use deep learning models (CNN)
- Add temporal consistency for video streams

Contributing

1. Fork the repository
2. Create a feature branch (`git checkout -b feature/new-feature`)
3. Commit changes (`git commit -am 'Add new feature'`)
4. Push to branch (`git push origin feature/new-feature`)
5. Create a Pull Request

License

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Citation

If you use this system in your research, please cite:

bibtex

```
@software{parking_detection_system,  
  title={Automated Parking Space Detection System},  
  author={Your Name},  
  year={2024},  
  url={https://github.com/your-repo/parking-detection-system}  
}
```

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- PKLot dataset creators for providing the benchmark dataset
- OpenCV community for computer vision tools
- Scikit-learn developers for machine learning algorithms

Contact

For questions and support, please open an issue on GitHub or contact [your-email@example.com].