



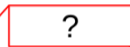
Sequence-based Recognition of License Plates with Severe Out-of-Distribution Degradations

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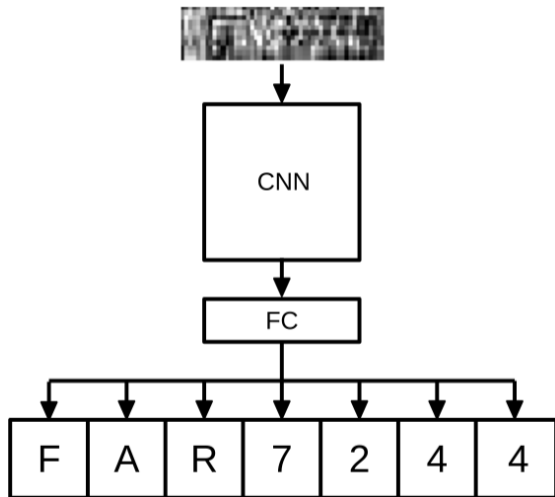
Challenges for Recognizing License Plates (LPs)

- **Goal:** Automatically deciphering LPs subjected to unknown degradations
- **Problem:** Digital image quality can heavily be impacted in practice
 - Images can be degraded beyond human recognition
 - Information may still be available for reconstruction



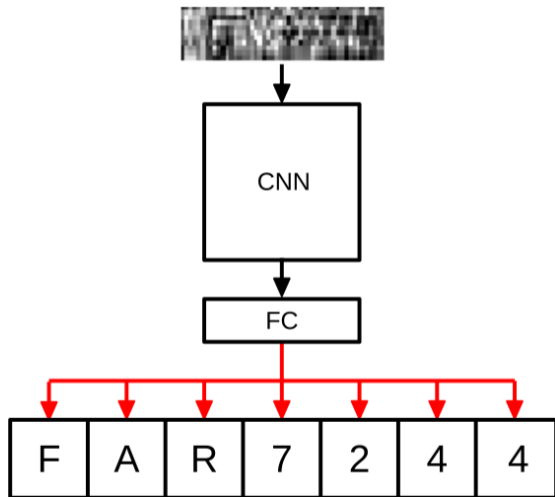
Prior Work and Research Questions

- Prior work: CNNs are applied to recognize severely degraded LPs



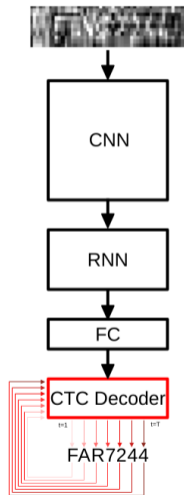
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 - **Sequential context information?**
 - LP caption = sequence
 - Prior work on recognizing high-quality LPs profit from CNN-RNN combinations
 - Top-n sequence predictions



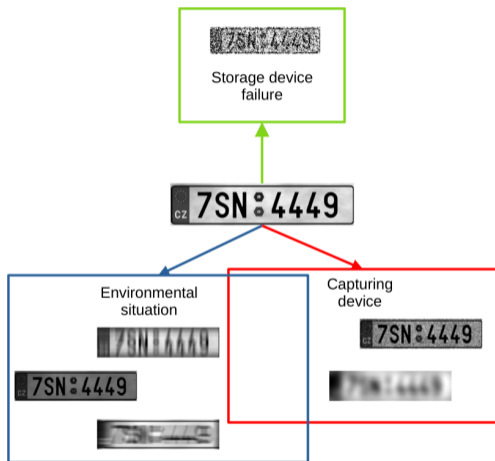
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 - **Robustness?**
 - Footage stems from uncontrolled sources
 - Degradation phenomena not covered in training data usually appear















Contributions

1. Adaptation of a sequence based approach for recognizing severely degraded license plates
2. Robustness evaluation on out-of-distribution data
3. Evaluation of top-n sequence predictions
4. Evaluation on in-distribution-data and real world data set (Reld [1])

Generation of Training Data

- Generation pipeline as described by Kaiser [3]
 - Synthetic LPs of Czech format
 - Degradation factors and strength controlled
 - Training set of 10 Mio samples
- Common degradation factors modeled by ...
 - Gaussian noise
 - Low resolution
 - JPEG-compression

		JPEG quality factor		
		95	30	1
width in px	180			
	120			
	70			
	30			

Evaluation

- Test data for robustness evaluation
 - Includes six degradation factors not occurring in the training set
 - strength of each factor ranges from low to high
 - Per test set: 10 000 samples of one degradation factor of one strength
- Metrics
 - Top-1 LP accuracies of CNN by Lorch *et al.* [4] and CRNN
 - CRNN top-n with $n \in \{3, 5, 10\}$ predictions from beam search decoding

Results - Underexposure

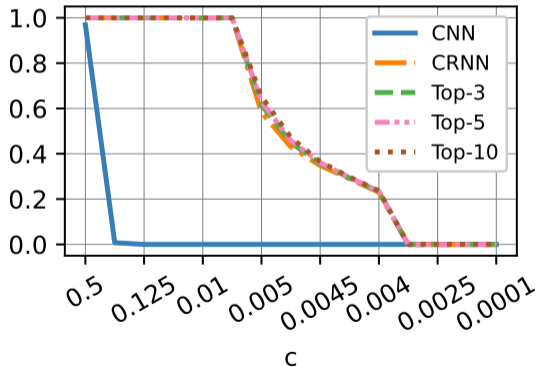
- Implementation

- Linear pixel shift by multiplicative constant c



- Remarks

- No. degradation levels: 15
- Beam search: hardly improves performance



Results - Shot Noise

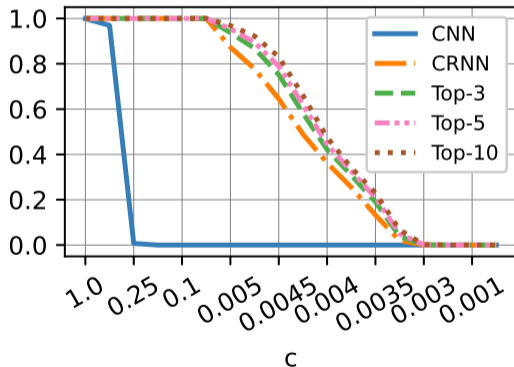
- Implementation

- Scaling by Poisson distribution $P_\lambda(n^p)$



- Remarks

- No. degradation levels: 18
- Beam search: more rewarding than for underexposure



Results - Salt and Pepper Noise

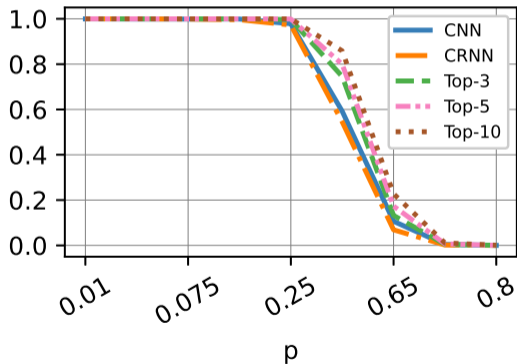
- Implementation

- Randomly set pixels to min. or max. value



- Remarks

- No. degradation levels: 9
- Beam search: $n = 3$ needed to surpass CNN



Results - Horizontal Motion Blur

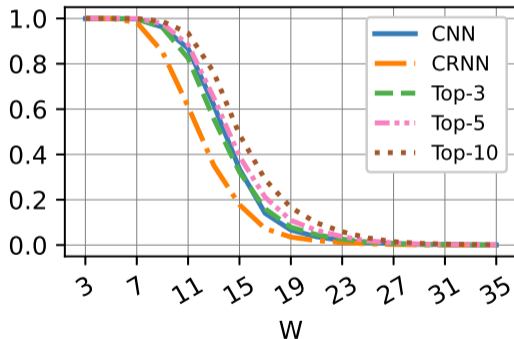
- Implementation

- Convolution with box kernel in horizontal direction



- Remarks

- No. degradation levels: 17
- Sequential processing disadvantageous for horizontally 'smeared' features
- Beam search: $n = 5$ needed to surpass CNN



Results - Vertical Motion Blur

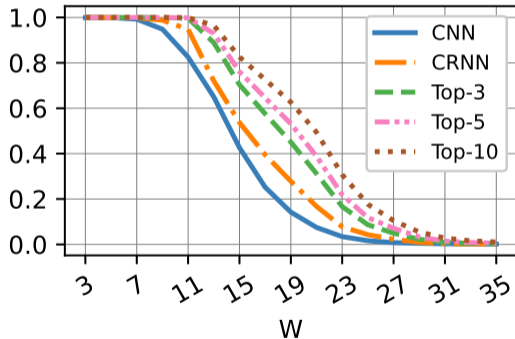
- Implementation

- Convolution with box kernel in vertical direction



- Remarks

- No. degradation levels: 17
- Beam search: vast performance increase



Results - Defocus Blur

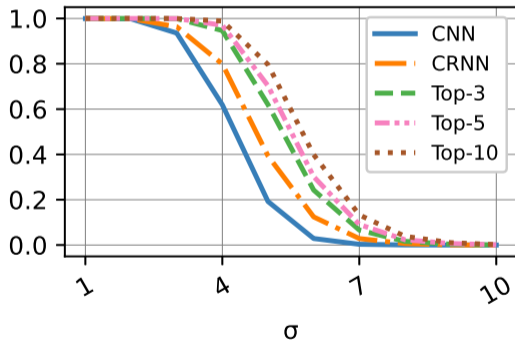
- Implementation

- Convolution with Gaussian kernel of width σ



- Remarks

- No. degradation levels: 10
- Beam search: exploits most information for $\sigma \in [4, 6]$



Conclusion and Future Work

- CRNN outperforms CNN on most out-of-distribution degradation factors
- Degradations in sequence direction are problematic (e.g. horizontal motion blur)
- Top-n sequence predictions pose an advantage for identifying LPs
- Future Work
 - Various additional degradation types remain to be investigated
 - Acquiring labeled real world data sets with controlled degradation types
 - Aim for more robustness towards degradation along the sequence dimension
 - Explore performance on LP formats with more syntactic sequence rules (e.g. German license plates)

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Thank you for listening.
Any questions?

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References



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- [1] Jakub Špaňhel *et al.* “Holistic recognition of low quality license plates by CNN using track annotated data”. In: *2017 14th IEEE International Conference on Advanced Video and Signal Based Surveillance (AVSS)*. IEEE. 2017, pp. 1–6.
- [2] Baoguang Shi, Xiang Bai, and Cong Yao. “An End-to-End Trainable Neural Network for Image-Based Sequence Recognition and Its Application to Scene Text Recognition”. In: *IEEE Transactions on Pattern Analysis and Machine Intelligence* 39.11 (Nov. 2017). Conference Name: IEEE Transactions on Pattern Analysis and Machine Intelligence, pp. 2298–2304. ISSN: 1939-3539. DOI: 10.1109/TPAMI.2016.2646371.
- [3] Paula Kaiser. “Learning to Decipher License Plates in Severely Degraded Images”. MA thesis. Friedrich-Alexander-Universität Erlangen-Nürnberg, Nov. 2019.
- [4] Benedikt Lorch, Shruti Agarwal, and Hany Farid. “Forensic reconstruction of severely degraded license plates”. In: *Electronic Imaging* 2019.5 (2019), pp. 529–1.