

Robust Lane Detection through Self Pre-training with Masked Sequential Autoencoders and Fine-tuning with Customized PolyLoss

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Robust Lane Detection through Self Pre-training with Masked Sequential Autoencoders and Fine-tuning with Customized PolyLoss



Background & Aim

- ❖ Lane detection is crucial for Automated Vehicles and ADAS
- ❖ Available vision based methods usually use one image to do lane detection
- ❖ Traditional methods usually adopted cumbersome hand-crafted features
- ❖ Deep learning based methods in literature still can not make full use of spatial-temporal information and correlation
- ❖ Available methods can not handle challenging driving scenes well

The main aim of this study is:

- To develop robust detection model handling challenging driving scenes
- To make full use of valuable features and aggregate contextual information
- To develop pre-training method for sequential vision based lane detection



Figure 1. Examples of challenging driving scenes.

The framework of the proposed pipeline

- End-to-end Encoder-decoder Structure
- Self Pre-training to Reconstruct Images
 - Masked sequential autoencoders
- Fine-tuning Segmentation
 - Transfer pre-trained model weights to the segmentation model
- Customized PolyLoss
- Post-processing with clustering & curve fitting
- Tested and verified on two data sets
 - tvtLANE normal (TuSimple lane)
 - tvtLANE challenging (12 cases)

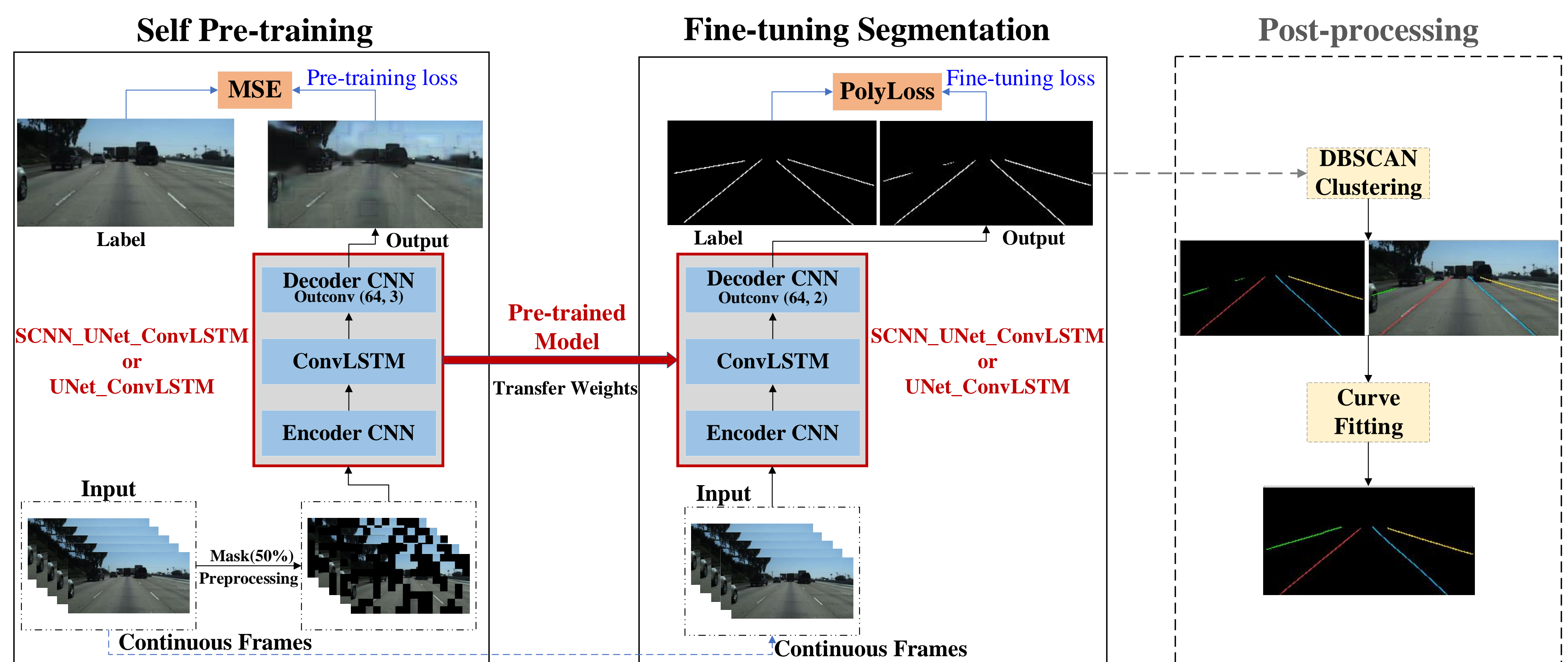


Figure 2. The framework of the proposed pipeline

Evaluation Metrics

- Accuracy
- Precision
- Parameter Size
- F1-Measure
- Recall
- MACs (Multiply-accumulate operations)

Results

Models	Test_Acc (%)	Precision	Recall	F1-Measure	MACs (G)	Params (M)
Using single image						
Baseline Models						
SegNet	96.93	0.796	0.962	0.871	50.2	29.4
UNet	96.54	0.790	0.985	0.877	15.5	13.4
SCNN*	96.79	0.654	0.808	0.722	77.7	19.2
LaneNet*	97.94	0.875	0.927	0.901	44.5	19.7
SegNet_ConvLSTM	97.92	0.874	0.931	0.901	217.0	67.2
UNet_ConvLSTM	98.00	0.857	0.958	0.904	69.0	51.1
Pre-trained Models						
UNet_ConvLSTM_CE**	98.19	0.882	0.940	0.910	69.0	51.1
UNet_ConvLSTM_PL**	98.34	0.921	0.909	0.915	69.0	51.1
Using multi-continuous images						
Baseline Models						
SCNN_SegNet	98.07	0.893	0.928	0.910	223.0	67.3
SCNN_UNet_ConvLSTM	98.19	0.889	0.950	0.918	93.0	51.3
Pre-trained Models						
SCNN_UNet_ConvLSTM_CE**	98.20	0.891	0.952	0.921	93.0	51.3
SCNN_UNet_ConvLSTM_PL**	98.38	0.929	0.915	0.922	93.0	51.3

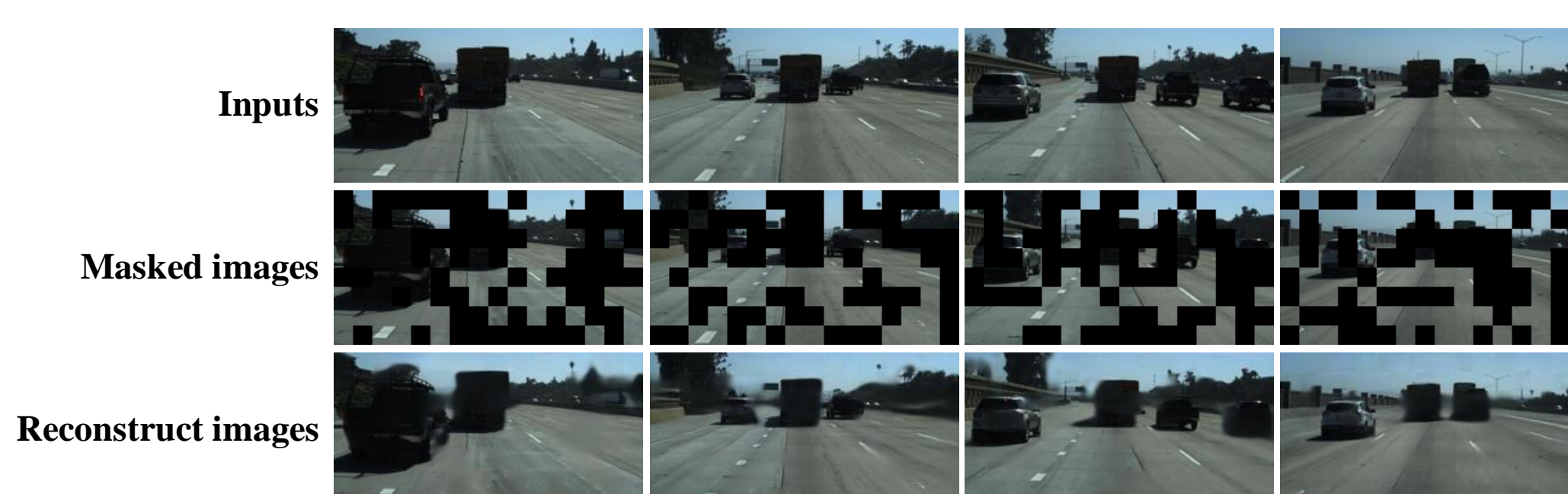


Figure 3. Visualization of the reconstructing results in the pre-training phase.

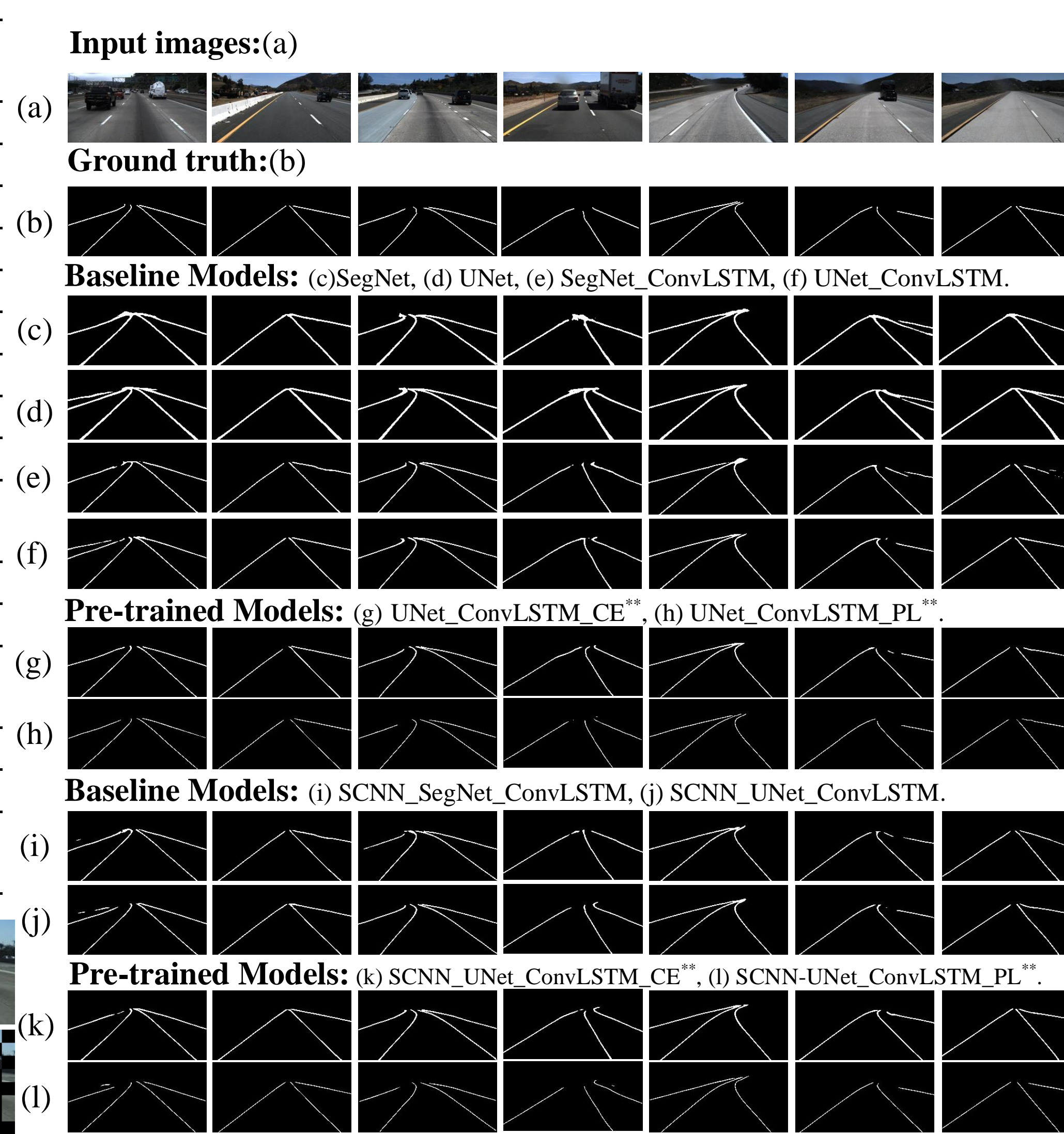


Figure 4. Visualization of lane-detection results on normal cases.

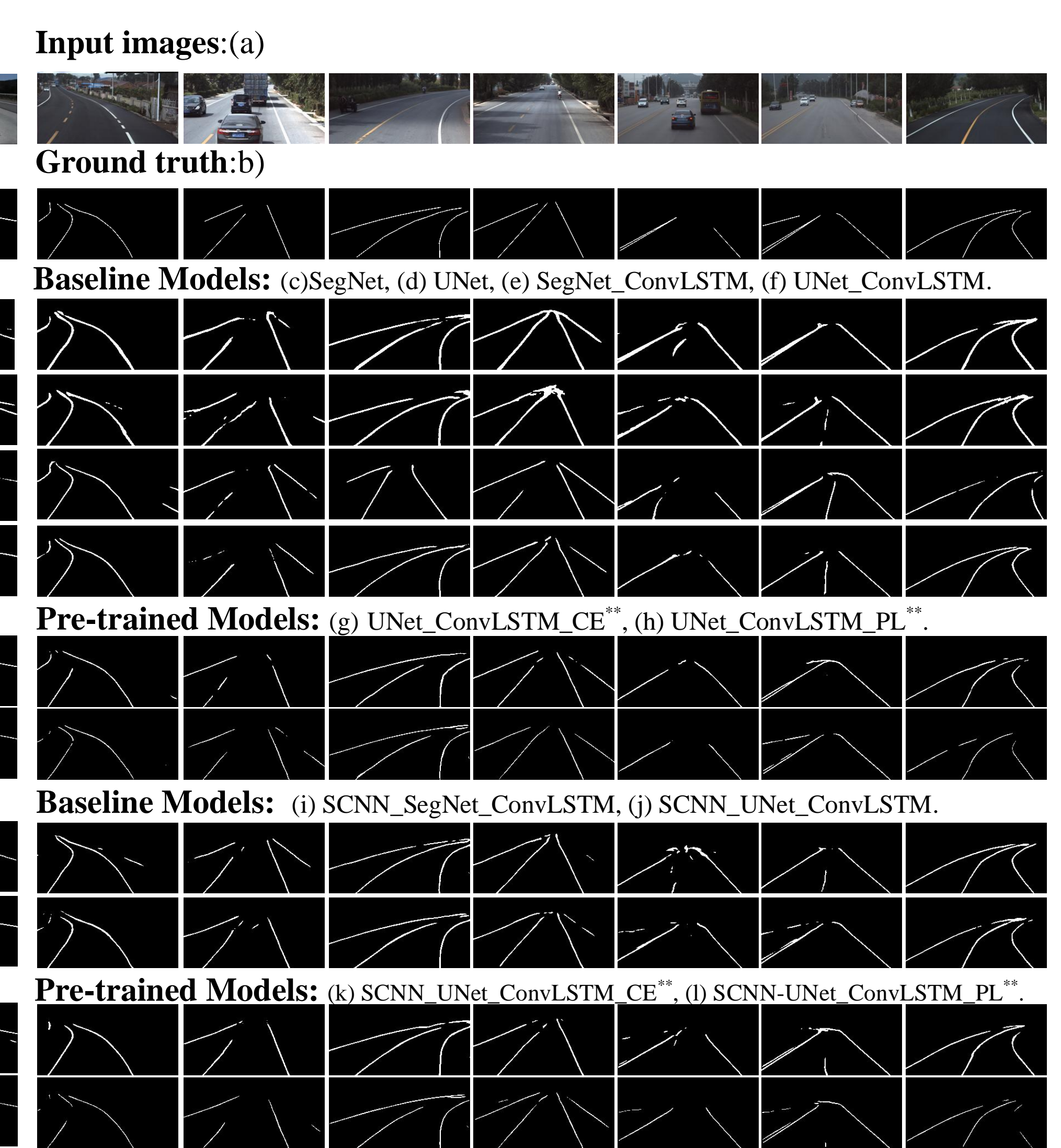


Figure 5. Visualization of lane-detection results on 7 challenging driving scenes.

Ablation Study

Testing Datasets	Testing different loss functions and models	Testset #1 (Normal Situations)				Testset #2 (Challenging Situations)			
		Loss Function	Test_Acc (%)	Precision	F1-Measure	Loss Function	Test_Acc (%)	Precision	F1-Measure
UNet_ConvLSTM	CE	98.19	0.882	0.910	CE	98.13	0.7932	0.6537	
	PL	98.34	0.921	0.915	PL	98.38	0.8331	0.6284	
SCNN_UNet_ConvLSTM	CE	98.20	0.891	0.921	CE	98.03	0.8001	0.7327	
	PL	98.38	0.929	0.922	PL	98.36	0.8444	0.6711	

Conclusions

- The proposed masked sequential autoencoder based pre-training and customized PolyLoss are useful
- The proposed pipeline is effective and robust which can improve the performances of SOTA models in both normal and challenging cases