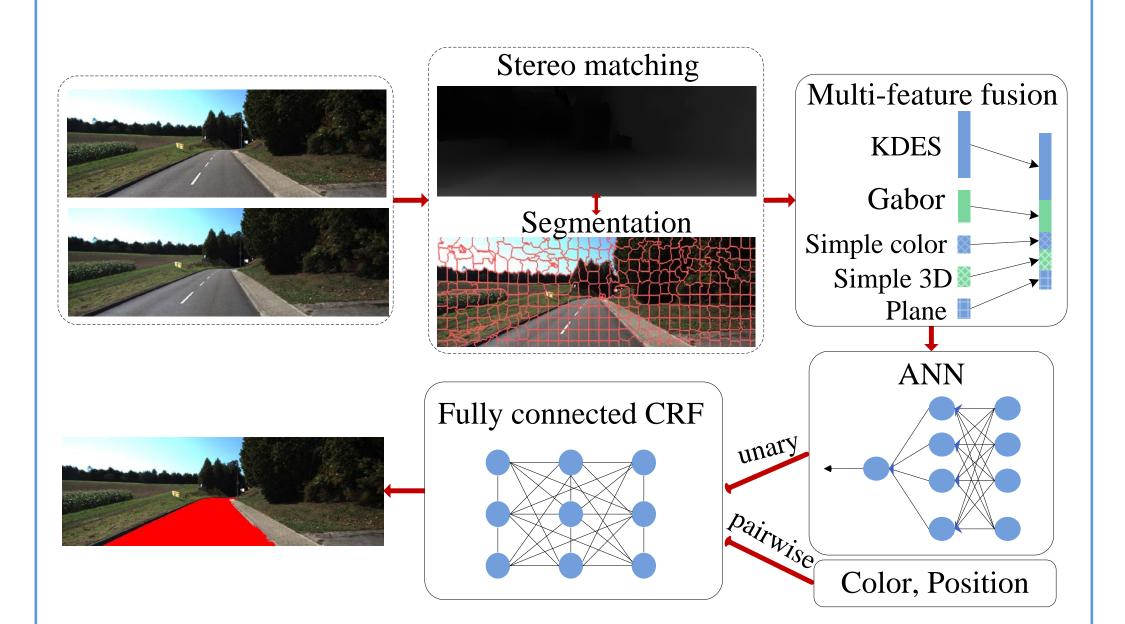


# FEATURE++: CROSS DIMENSION FEATURE FUSION FOR ROAD DETECTION

### **ABSTRACT**

Road detection is a key component of Advanced Driving Assistance Systems, which provides valid space and candidate regions of objects for vehicles. Mainstream road detection methods have focused on how to extract discriminative features. In this paper, we propose a robust feature fusion framework, called "Feature++", which is combined with superpixel feature and 3D feature extracted from stereo images. Then a neural network classifier is trained to decide whether a superpixel is road region or not. Finally, the classified results are further refined by conditional random field. Experiments conducted on the KITTI ROAD benchmark show that the proposed "Feature++" method outperforms most manually designed features, and are comparable with state-of-the-art methods that based on deep learning architecture.

### **METHOD**



### **Features**:

- The 2D features include RGB, Gabor, Gradient kernel descriptors (GKDES) and RGB kernel descriptors (RGBKDES).
- The 3D features consist of Plane(angles and inliers percentage), Pos3D, Depth gradient kernel descriptors (DGKDES) and SPIN kernel descriptors(SPINKDES).

### **Fully connected CRF**:

- Unary potential takes the negative log-likelihood of the road confidence.
- Pairwise potential =  $\omega^{(1)} exp\left(-\frac{|p_i p_j|^2}{2\Omega^2} \frac{|I_i I_j|^2}{2\Omega^2}\right)$

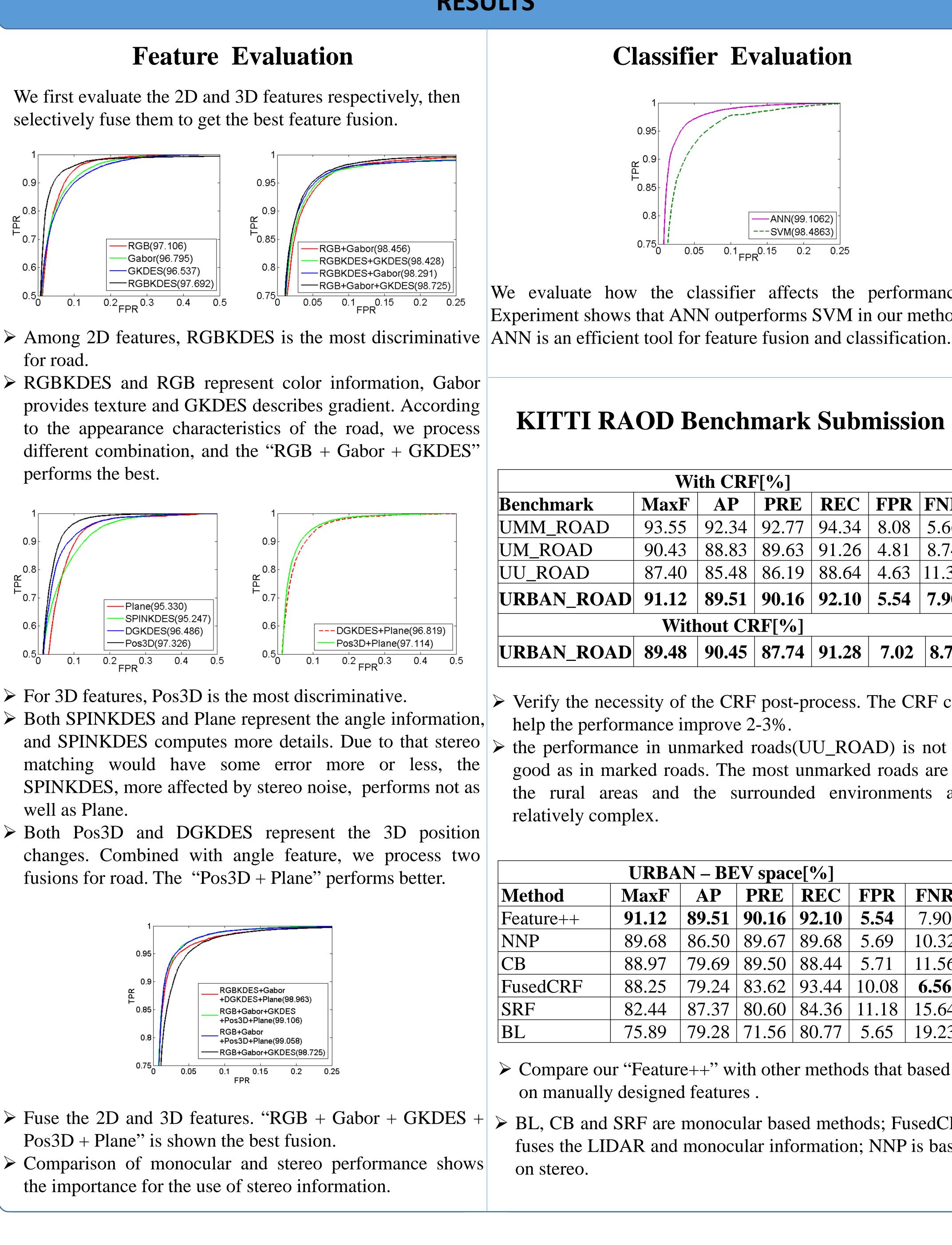
$$+\omega^{(2)}exp\left(-\frac{|p_i-p_j|^2}{2\theta_{\gamma}^2}\right)$$

*I* is the color vector and *p* represents position.

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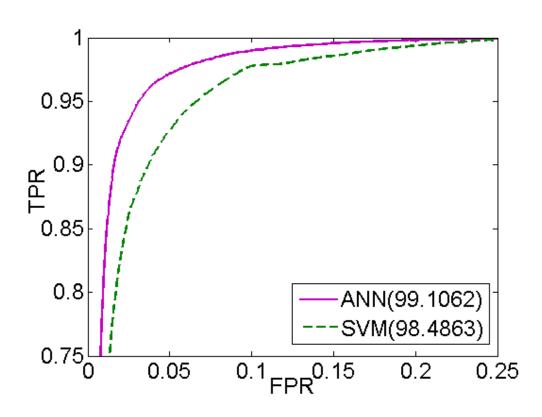
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## **Classifier Evaluation**



We evaluate how the classifier affects the performance. Experiment shows that ANN outperforms SVM in our method.

# **KITTI RAOD Benchmark Submission**

With CRF[%]										
nchmark	MaxF	AP	PRE	REC	FPR	FNR				
/IM_ROAD	93.55	92.34	92.77	94.34	8.08	5.66				
/I_ROAD	90.43	88.83	89.63	91.26	4.81	8.74				
J_ROAD	87.40	85.48	86.19	88.64	4.63	11.36				
RBAN_ROAD	91.12	89.51	90.16	92.10	5.54	7.90				
Without CRF[%]										
RBAN_ROAD	89.48	90.45	87.74	91.28	7.02	8.72				

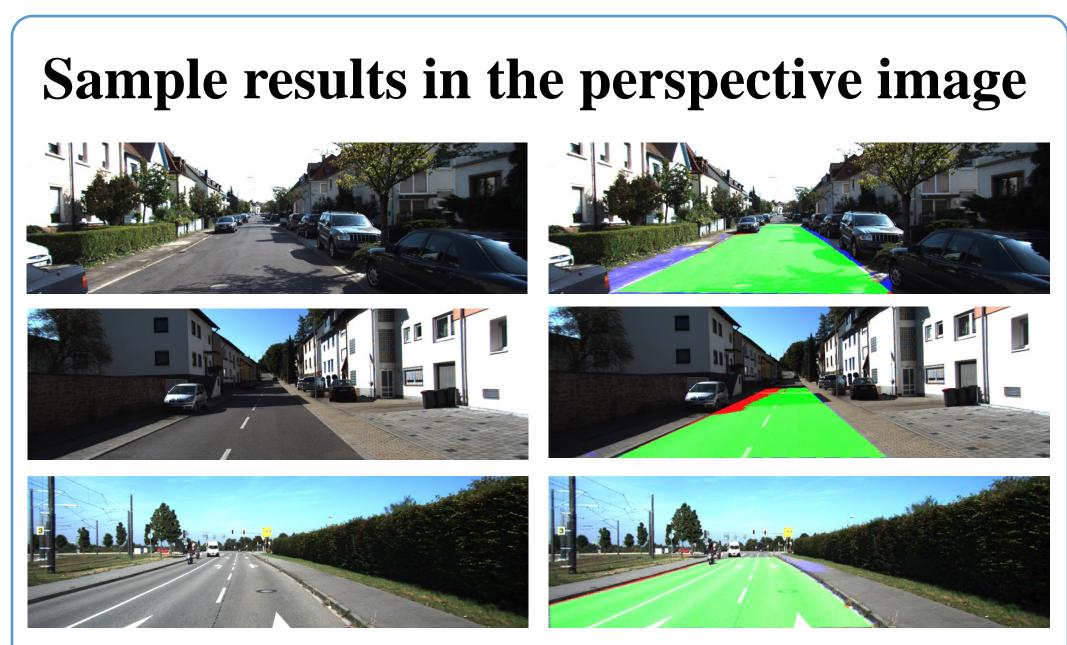
➤ Verify the necessity of the CRF post-process. The CRF can help the performance improve 2-3%.

➤ the performance in unmarked roads(UU\_ROAD) is not as good as in marked roads. The most unmarked roads are in the rural areas and the surrounded environments are relatively complex.

URBAN – BEV space[%]										
ethod	MaxF	AP	<b>_</b>	REC	FPR	FNR				
ature++	91.12	89.51	90.16	92.10	5.54	7.90				
1P	89.68	86.50	89.67	89.68	5.69	10.32				
	88.97	79.69	89.50	88.44	5.71	11.56				
sedCRF	88.25	79.24	83.62	93.44	10.08	6.56				
F	82.44	87.37	80.60	84.36	11.18	15.64				
1	75.89	79.28	71.56	80.77	5.65	19.23				

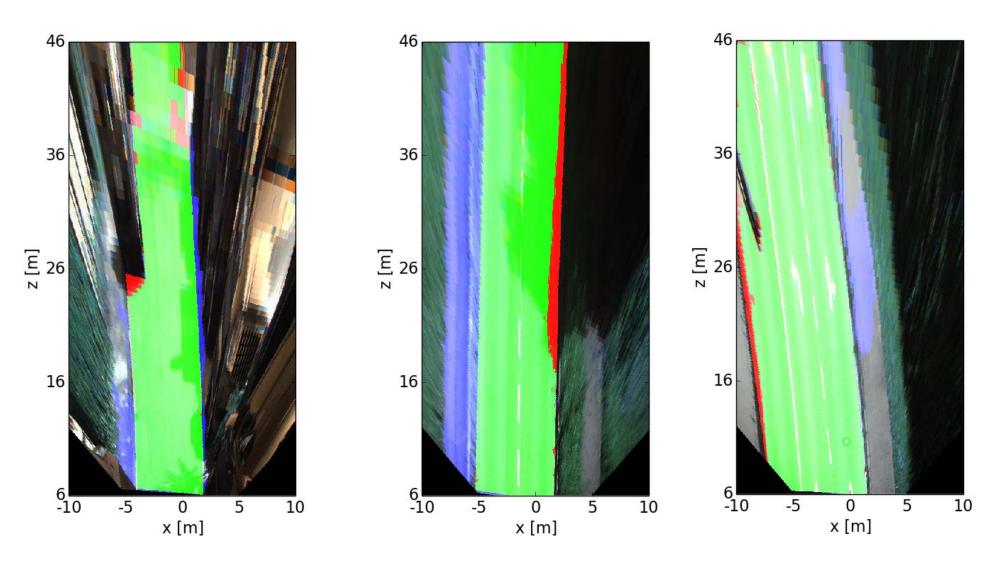
 $\succ$  Compare our "Feature++" with other methods that based on manually designed features.

> BL, CB and SRF are monocular based methods; FusedCRF fuses the LIDAR and monocular information; NNP is based on stereo.



Red denotes false negatives, blue areas correspond to false positives and green represents true positives(top to bottom: UU, UM, UMM).





In this paper, we have proposed a robust feature for road detection, which selectively fuses the Gabor, kernel descriptors, simple color and 3D spatial information. Meanwhile, we have evaluated each component of our system. Experiments show that our method outperforms the most of methods which use the manually designed features on the KITTI ROAD benchmark. However, Illumination conditions usually affect the performance, especially in unmarked road scenes. In our future works, illumination invariant features would be included in the road feature fusion and used in the pairwise potentials of CRF.

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### CONCLUSION

# ACKNOWLEDGMENTS