

Run-time Monitoring of 3D Object Detection in Automated Driving Systems Using Early Layer Neural Activation Patterns

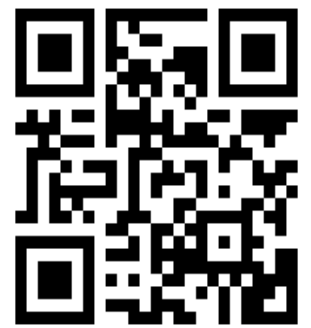
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Presenter: Dannier Xiao

Contact



Paper



- Introduction
- Methodology
- Results
- Conclusion



<https://www.abc15.com/news/region-southeast-valley/chandler/waymo-car-involved-in-chandler-arizona-crash>

<https://abc7.com/tesla-gardena-crash-driver/11873142/#:~:text=GARDENA%2C%20Calif.%20%28KABC%29%20--%20The%20driver%20of%20a,vehicular%20manslaughter%2C%20a%20Los%20Angeles%20County%20judge%20said.>

- Object detection is essential for understanding the environment in automated driving systems (ADS).
- Failure in object detection may cause significant harm to road users.
- Deep neural network-based mechanisms are prone to errors.

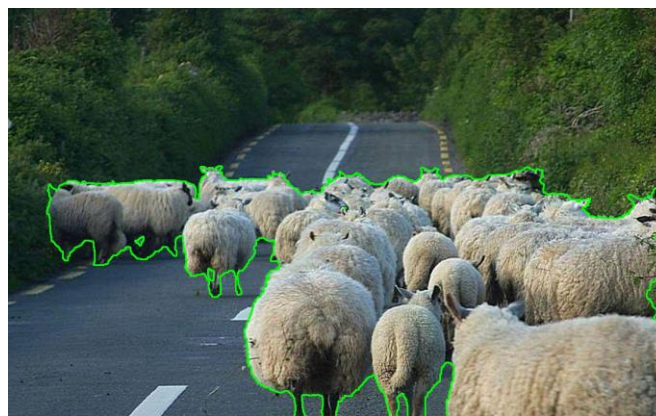
In Distribution

Missing or misclassifying known objects.



Out-of-Distribution

Not being able to detect unknown objects.



Adversarial

Missing or misclassifying known objects due to external perturbation.

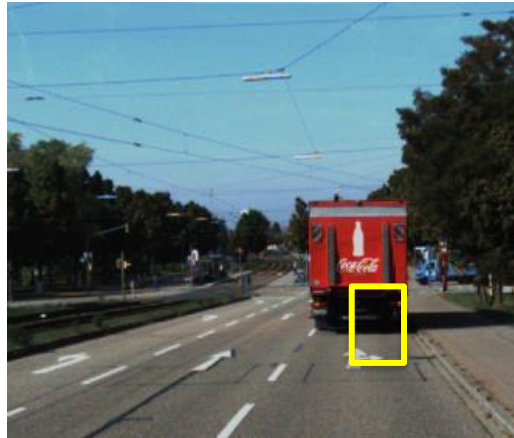


classified as
Stop Sign



classified as
Max Speed 100

Missing an object

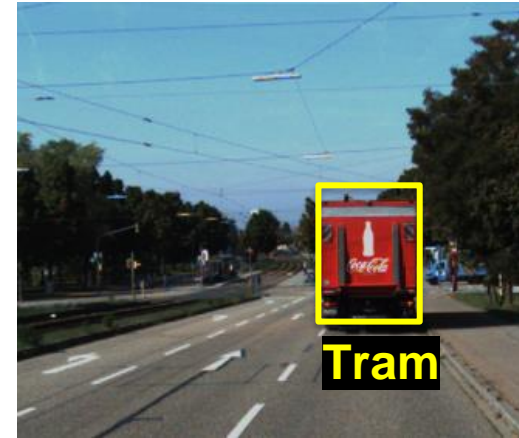


Poor localisation



Not being able to detect at all

Misclassifying an object



Incorrect label

 Prediction

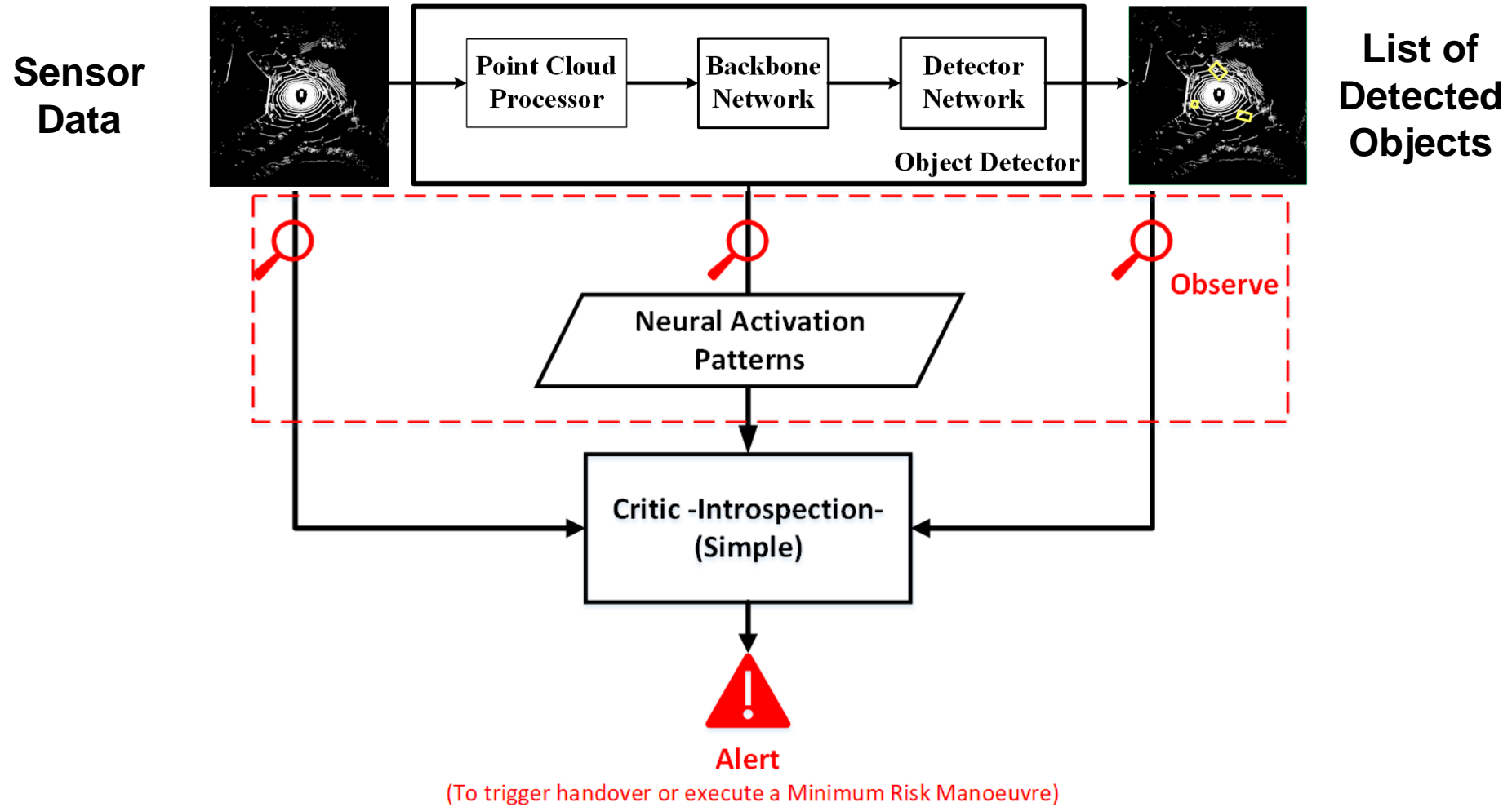
- Primary work on out-of-distribution errors, and uncertainty awareness.
- Gap for in-distribution

- Primary work on 2D object detection.
- Gap for 3D data and 3D object detection.

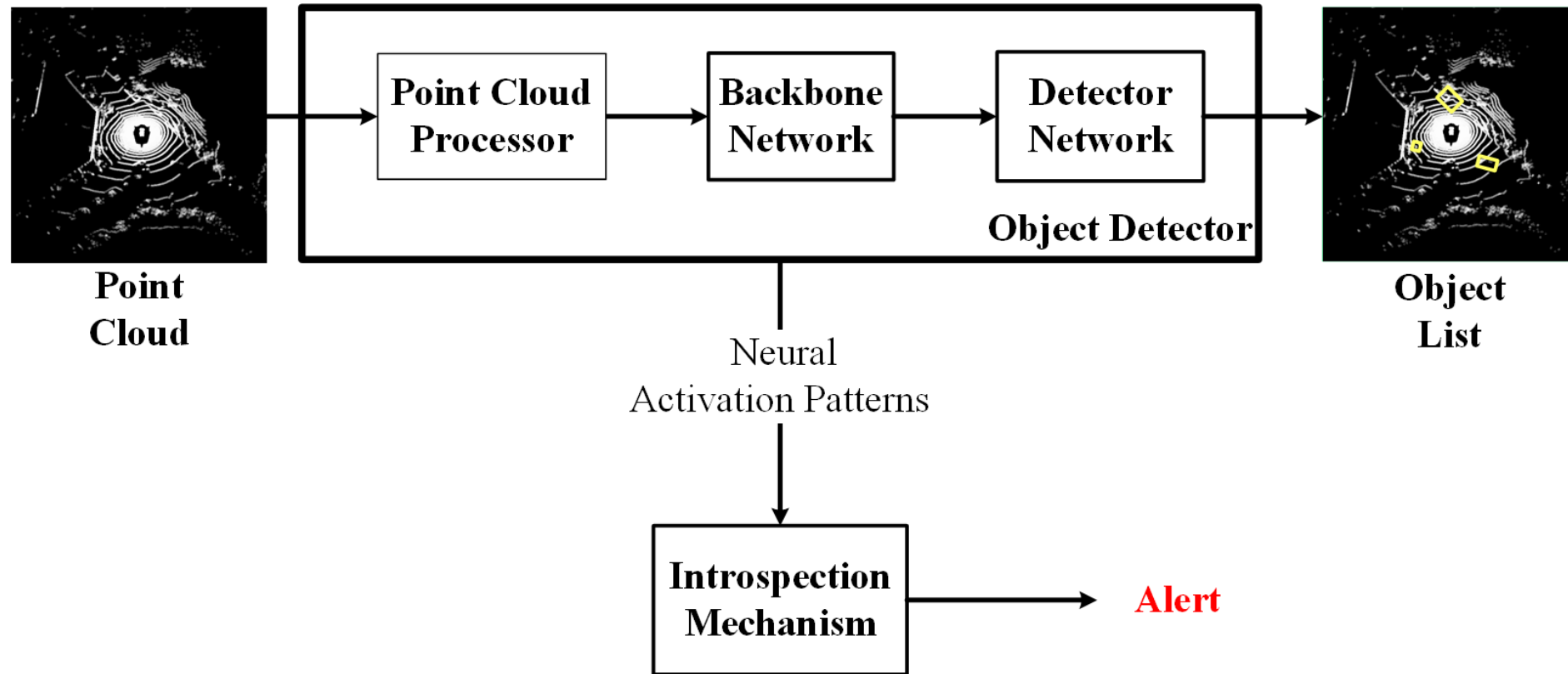
- Primary work on last layer activation patterns from object detectors for monitoring.
- Gap exploring different layers for monitoring.

Actor-Critic Architecture

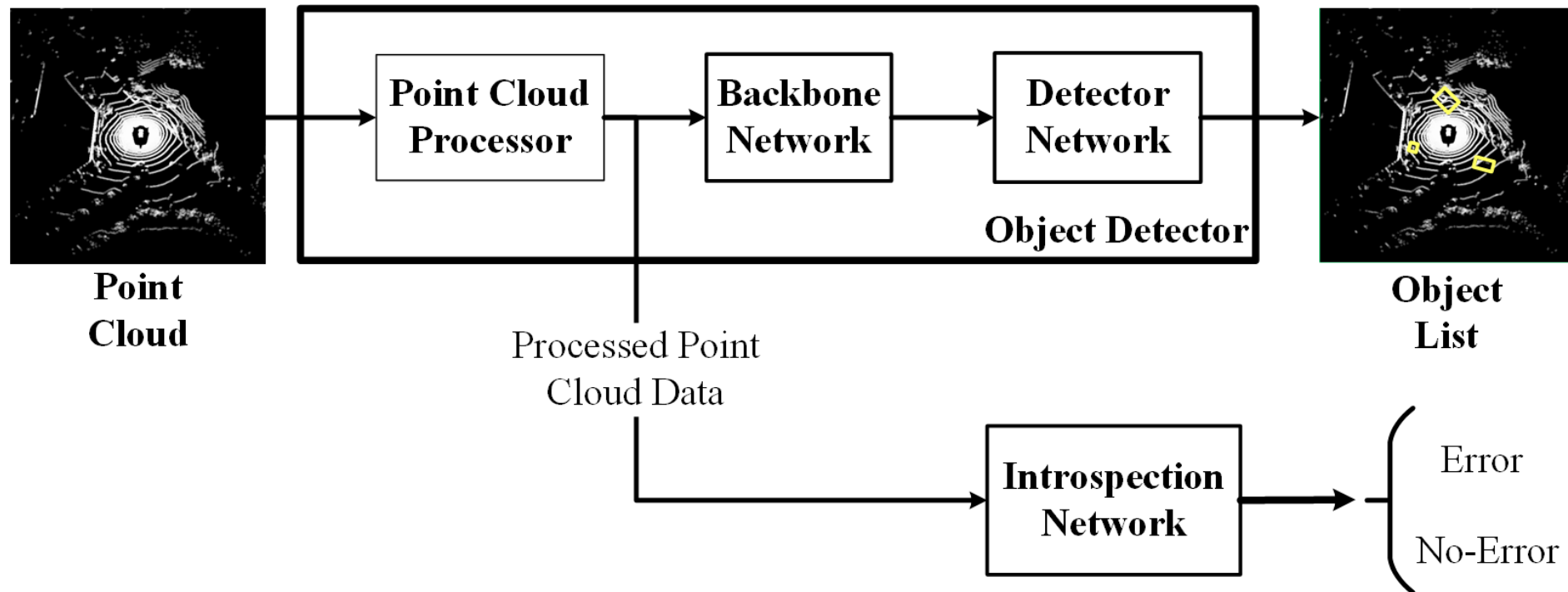
7



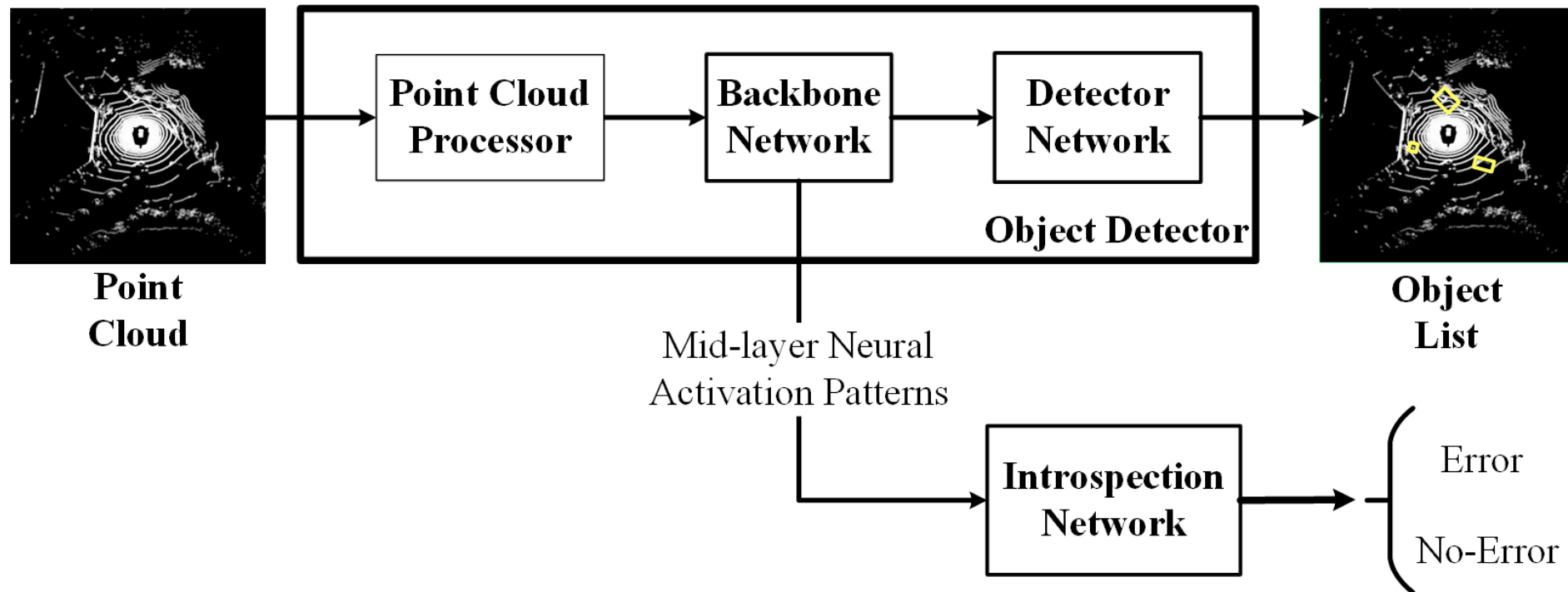
Method: Actor-Critic Architecture



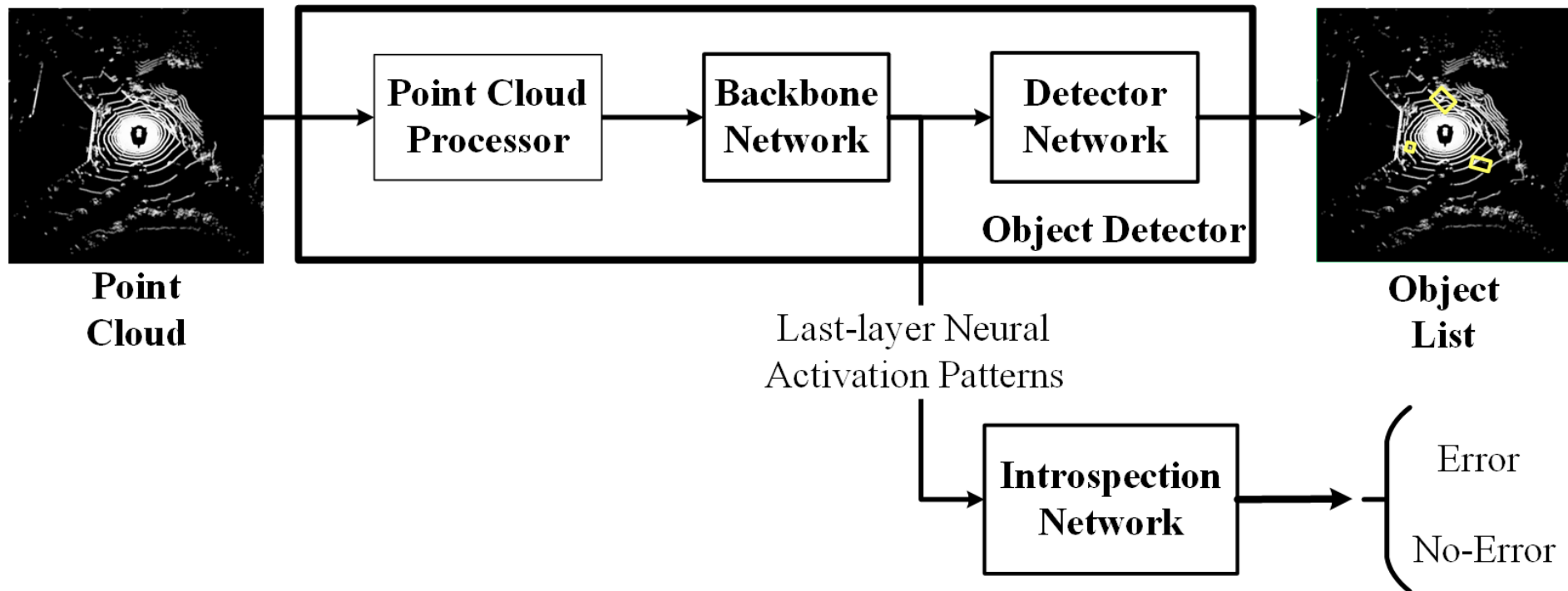
Method: Early Layer Activations



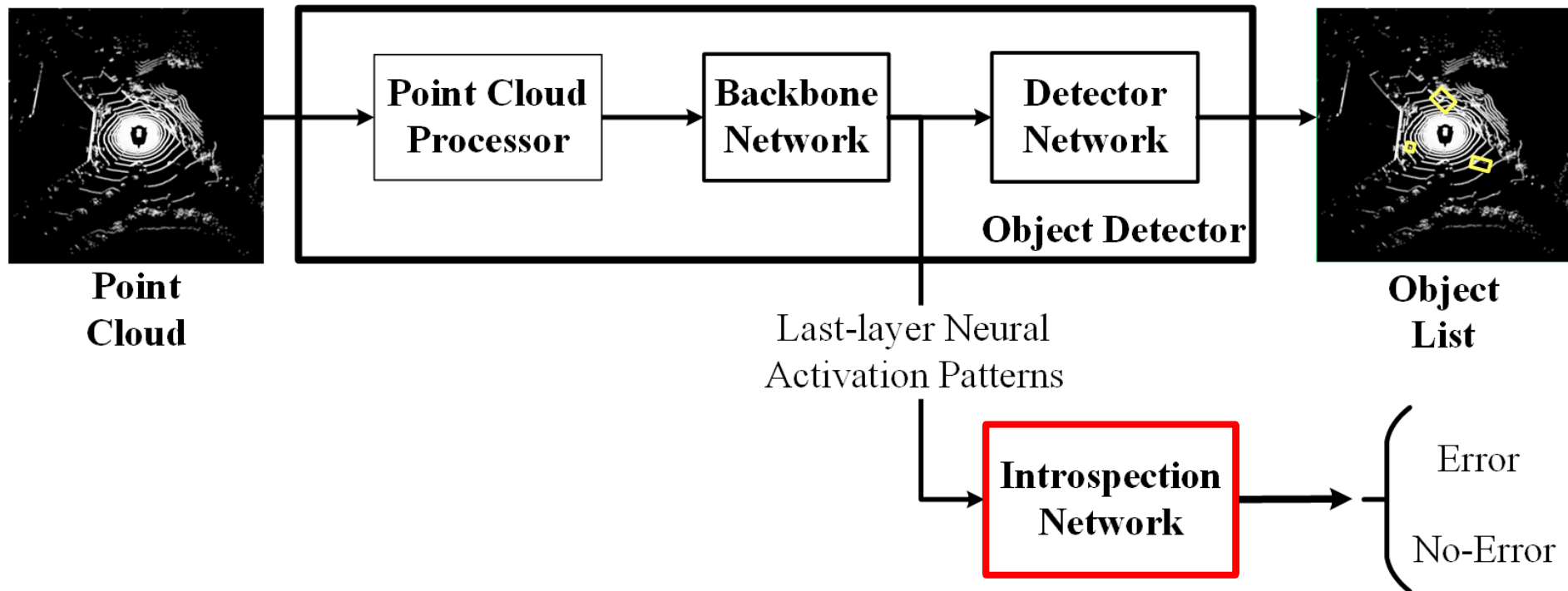
Method: Mid-Layer Activations

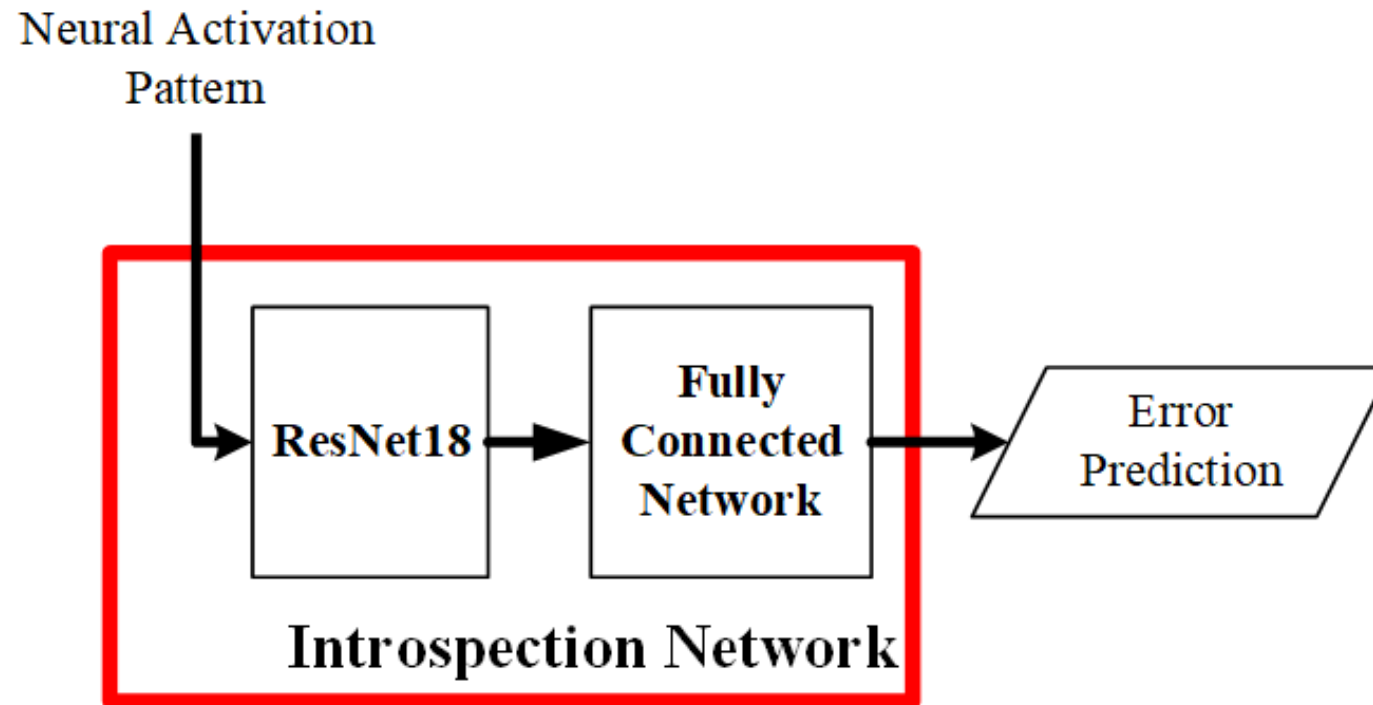


Method: Late Layer Activations

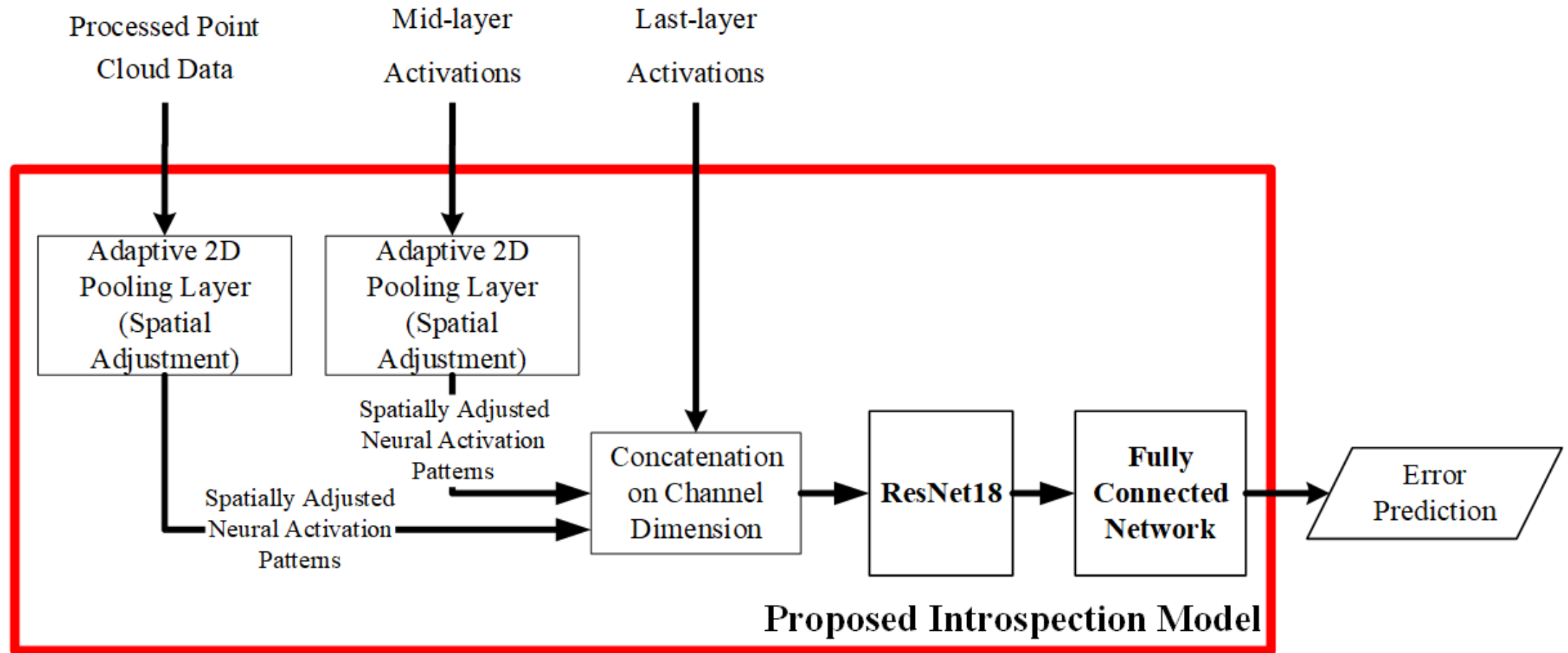


Method: Late Layer Activations





Method: Proposed Mechanism



Dataset / Model	Input	Rec.(-)	Rec.(+)	AUROC
Kitti / PointPillars	EARLY	0.7764	0.7524	0.8420
	MID	<u>0.7500</u>	<u>0.7460</u>	<u>0.8368</u>
	LATE	0.6268	0.8104	0.8036
	CONCAT	0.7077	0.7858	0.8309
NuScenes / CenterPoint	EARLY	0.7945	0.8995	0.9198
	MID	0.7945	0.9060	0.9330
	LATE	0.7123	0.8581	0.8919
	CONCAT	<u>0.8650</u>	<u>0.8630</u>	<u>0.9288</u>

Best result is in bold.

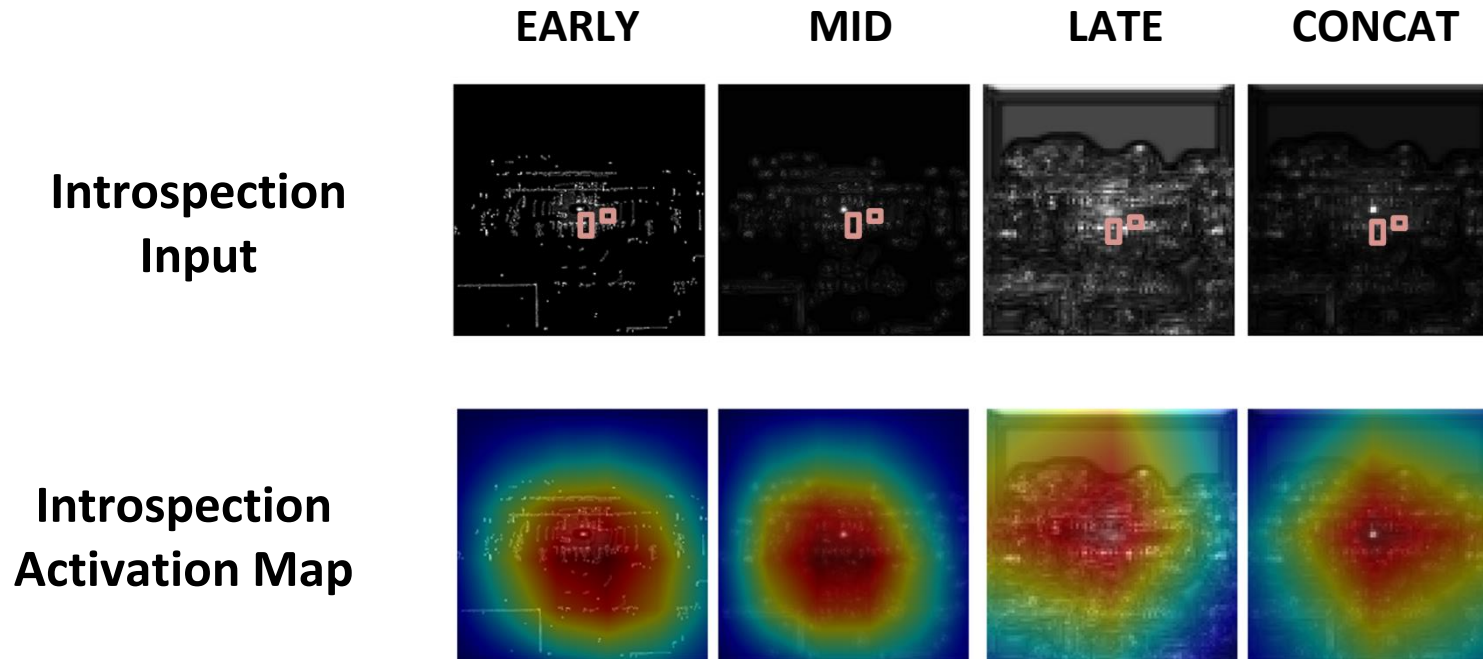
Second best is underlined.

Computational Requirement Comparison

Method	CPU Time (ms)	GPU Time (ms)	FLOPs (G)
EARLY	54.32 (9.54)	11.47 (1.21)	36.32
MID	9.43 (3.26)	2.01 (0.10)	3.68
LAST	5.01 (0.47)	1.80 (0.06)	1.60
CONCAT	<u>4.94 (0.32)</u>	<u>1.95 (0.07)</u>	<u>2.60</u>

Best result is in bold.

Second best is underlined.



Driving direction is from right to left in the example.

- **Early Layer Activations:** Improve error detection in 3D object detection, particularly for complex datasets.
- **Combining Layers:** Offers a balanced approach, enhancing both performance and computational efficiency.
- **Future Research:** Explore dataset-shift and methods for multi-layer activation utilization.

Thank You!



Contact



Paper



Q&A

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