

Vehicle Classification from Profile Measures

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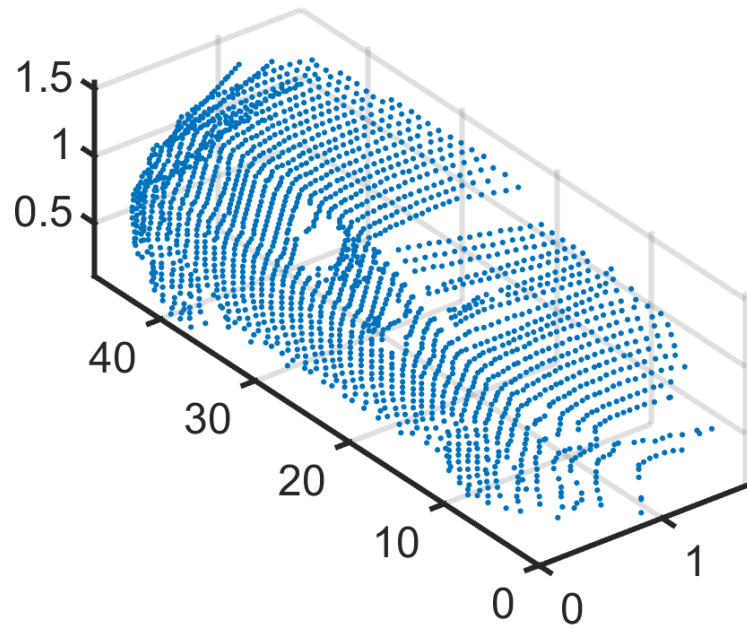
Università degli studi di Udine

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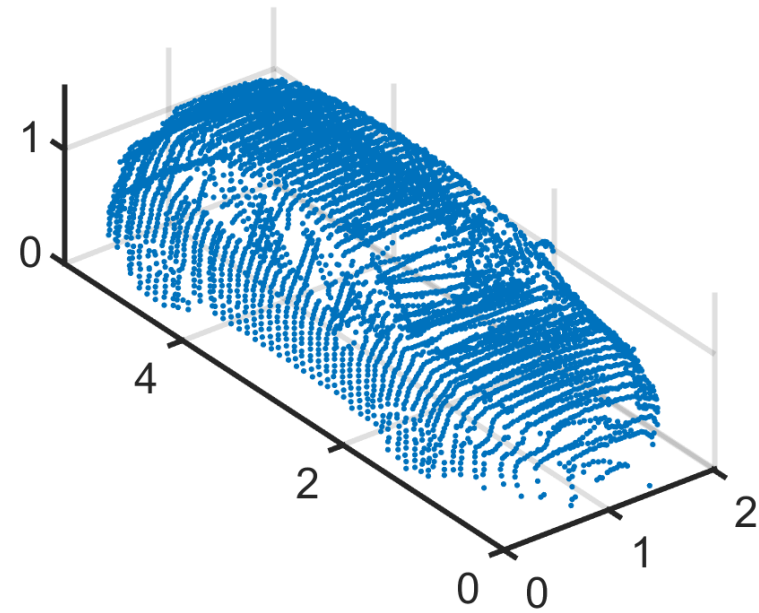


Vehicle Classification Problem

Goal: Classify vehicles from **point clouds** in two configurations



Non-metric configuration



Metric configuration

State of the art: VoxNet

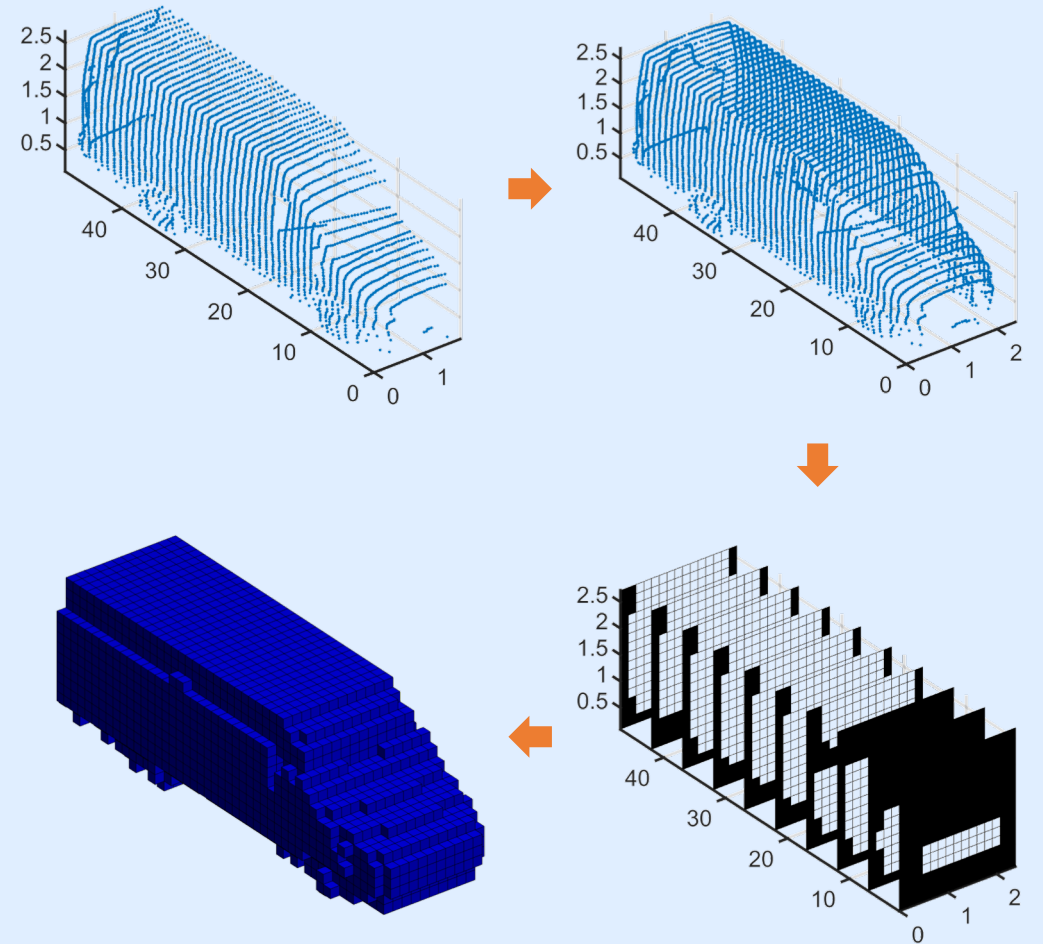
CNN with 3-dimensional convolutional layers

Advantages

- Extraction of rich (shape) features
- Works in both configurations

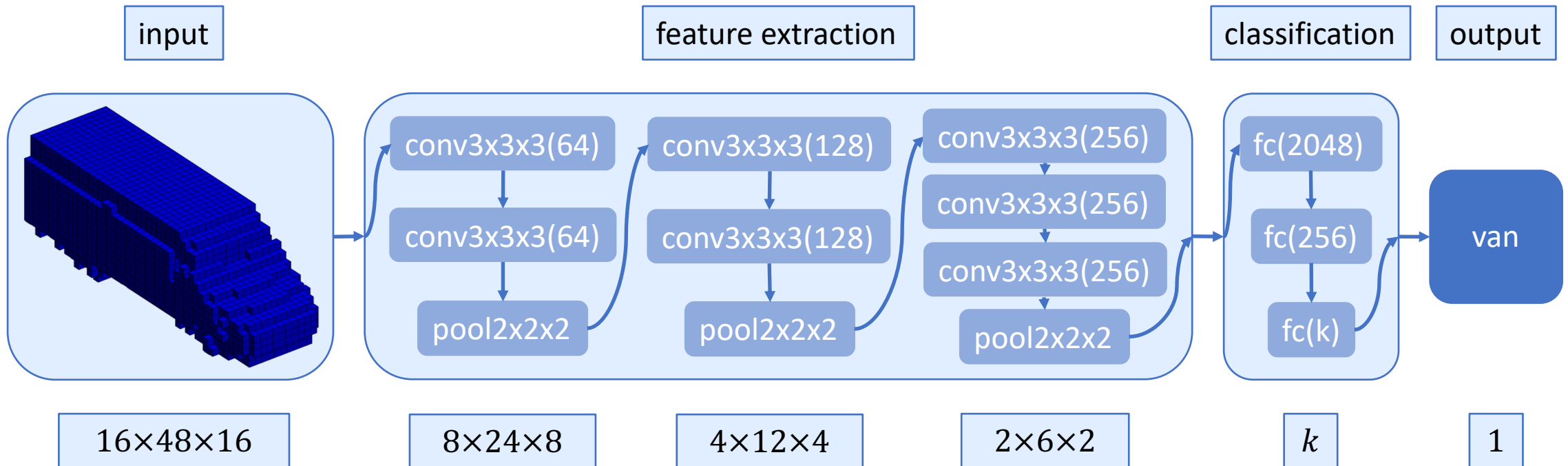
Disadvantages

- Lack of metric information
- Low voxel resolution (16×48×16)
- High network complexity



Voxelization

State of the art: VoxNet architecture



State of the art: PointNet

CNN with *pointwise* convolutions

Advantages

- Extraction of metric features
- No resolution problem
- Low network complexity

Disadvantages

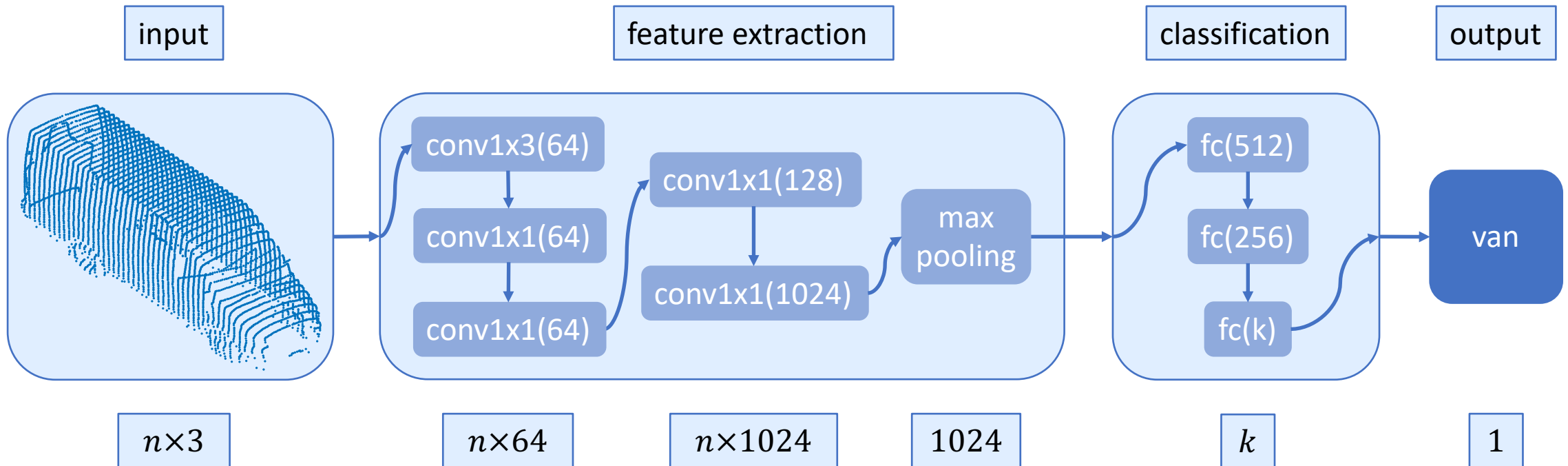
- Works in metric configuration only
- Can't extract neighbourhood information well

Input:

$n \times 3$

which is **unordered** and
with **variable length** n

State of the art: PointNet architecture



Our proposal

A CNN which is able to extract metric information from x and z coordinates, and non-metric one over y axis



The input is modified from point cloud to **structured point cloud**, reordering it in a tensor of shape (number of slices) \times (number of points per slice) $\times 2$

$n \times 3$



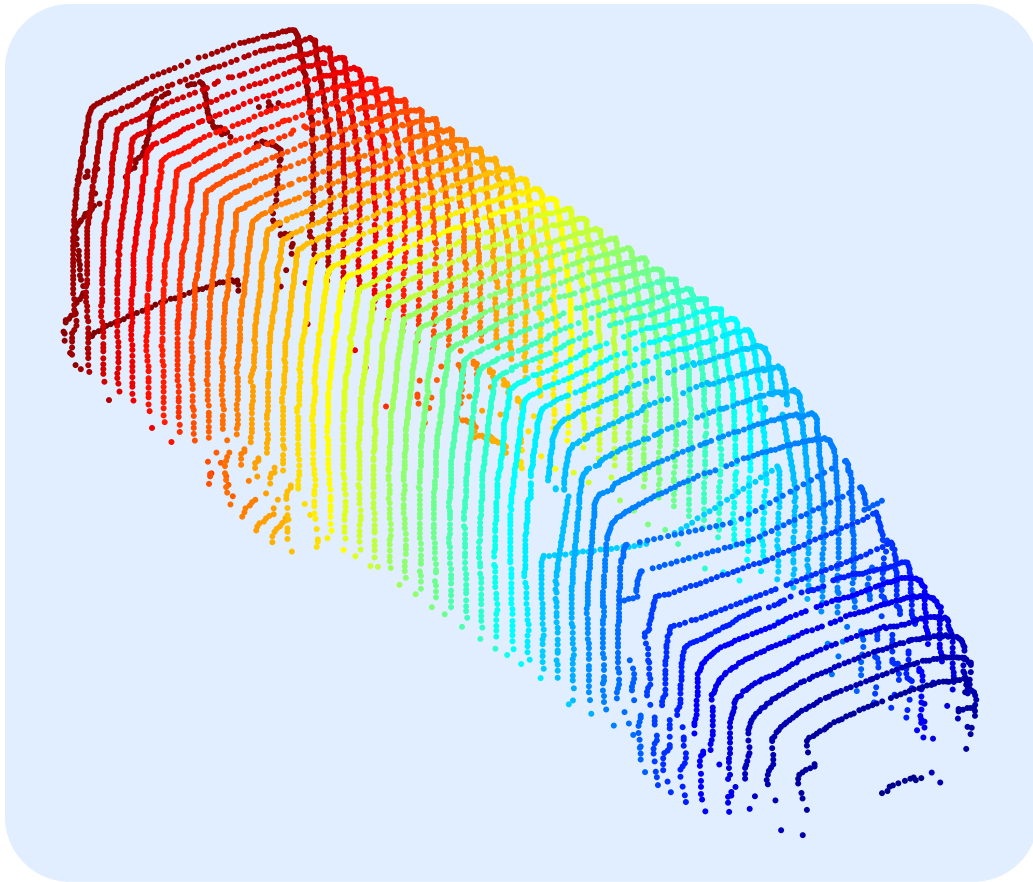
$s \times m \times 2$

where n is the number of points,
 s is the number of slices,

$$m = \max_{i=1, \dots, s} n_i, \text{ and}$$

n_i is the number of points in the
 i -th slice.

Our proposal: SliceNet



Combination of PointNet and VoxNet

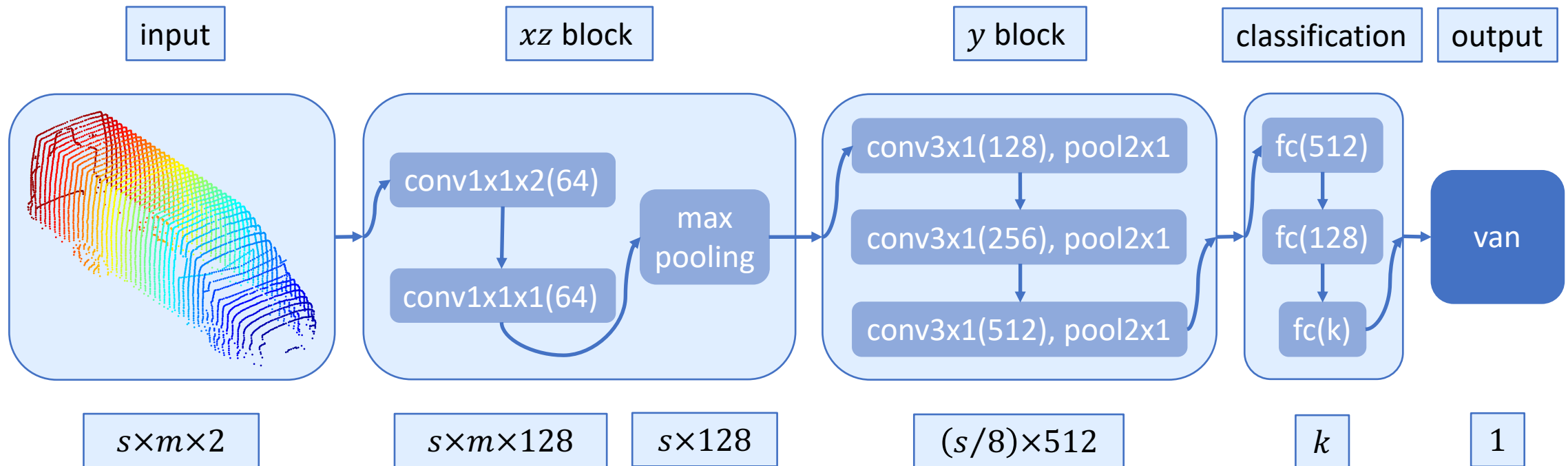
Advantages

- Extraction of metric features over x and z axis
- Extraction of neighbour information over y axis
- Low network complexity
- Works in both configurations

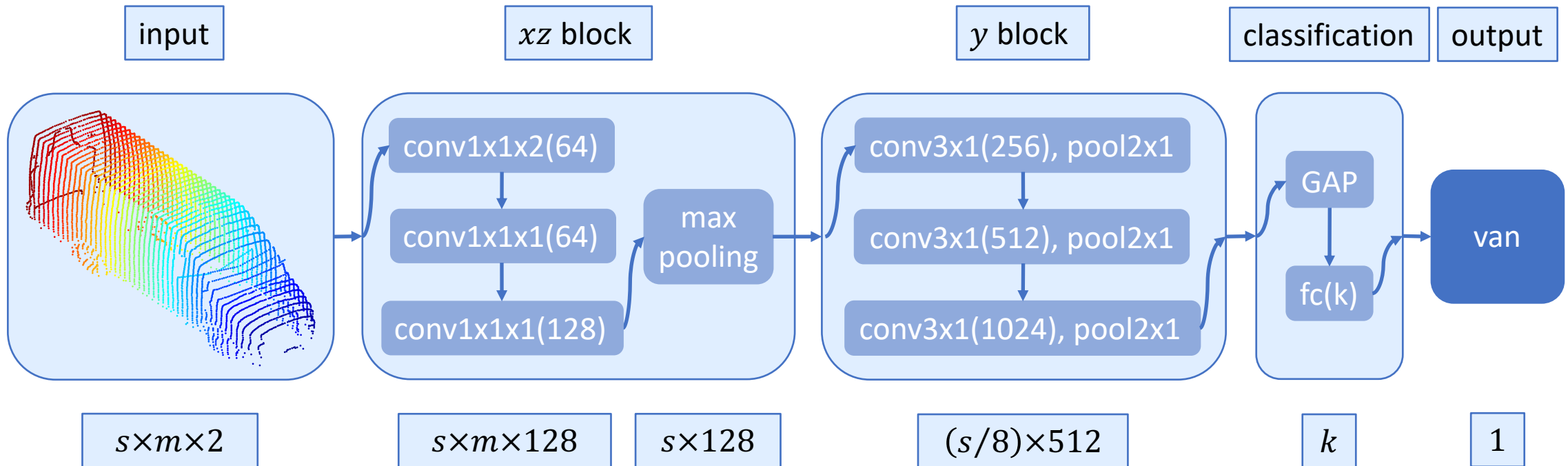
Disadvantages

- In metric configuration y coordinate is not exploited

Our proposal: SliceNet architecture

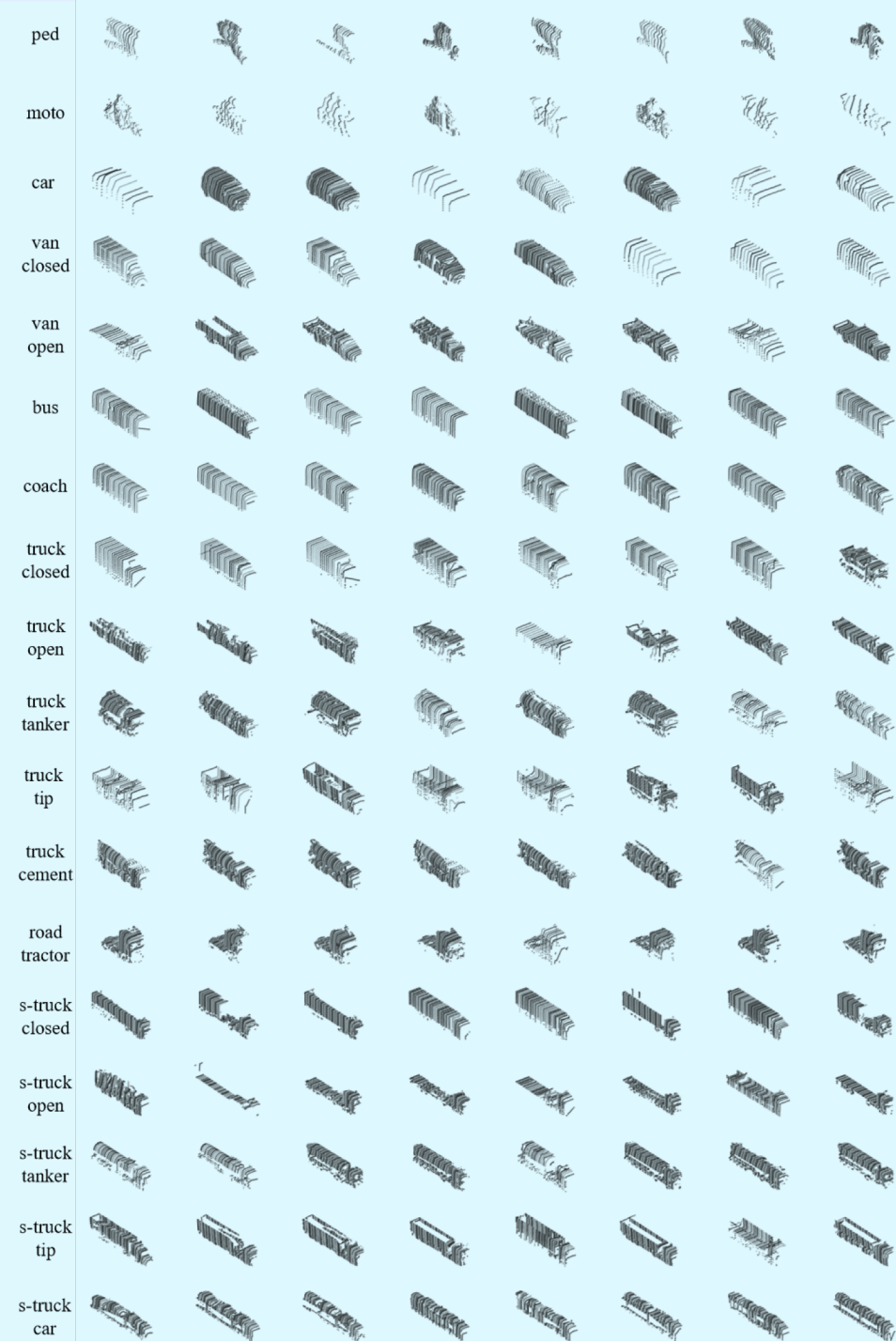
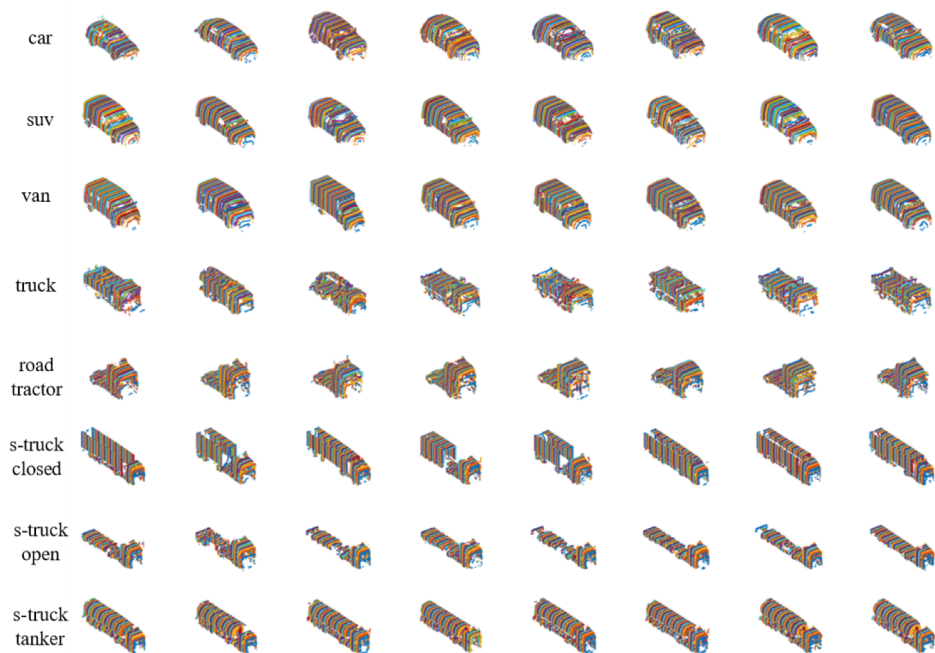


Our proposal: SliceNet GAP architecture

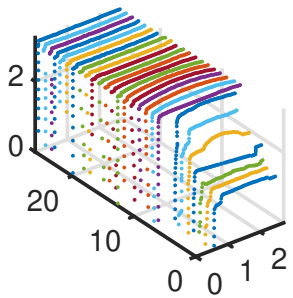


Experiments: Datasets

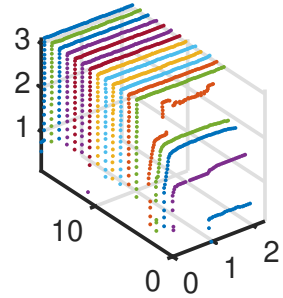
| | # samples | # classes (k) |
|------------|-----------|-------------------|
| Non-metric | ~6900 | 18 |
| Metric | ~2500 | 8 |



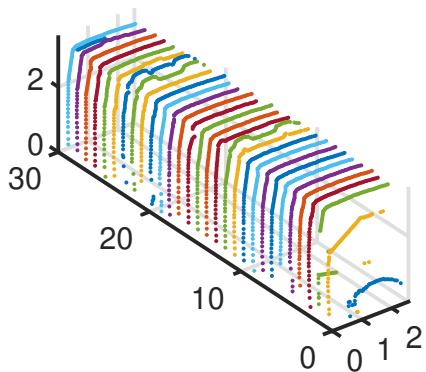
Experiments: Results



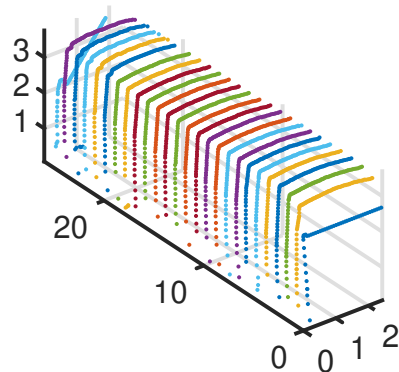
(a) Van, misclassified as truck.



(b) Truck, misclassified as van.



(c) Bus, misclassified as coach.



(d) Coach, misclassified as bus.

| | Accuracy | |
|-----------------|------------|--------|
| | Non-metric | Metric |
| VoxNet | 96.6 | 99.3 |
| PointNet | 96.7 | 99.4 |
| SliceNet | 97.2 | 99.4 |
| SliceNet GAP | 97.1 | 99.2 |

Thank you for your attention!