Table 6: The ablation of mIoU improvement of each component on the nuScenes and WOD val split when trained only for the segmentation task. XSF and STD stand for cross-space transformer and segmentation transformer decoder.

| Baseline (3.1) | STD | Multi-frame | XSF | nuScenes | WOD |
|----------------|-----|-------------|--------------|----------------------|--------------------|
| √ | | | | 76.6 | 70.3 |
| ✓ | ✓ | | | 78.3 (+1.7) | 70.6 (+0.3) |
| ✓ | ✓ | ✓ | | 80.8 (+4.2) | 71.2 (+0.9) |
| ✓ | ✓ | ✓ | \checkmark | 81.7 (+5.1) | 71.3 (+1.0) |

Table 8: Panoptic segmentation result on nuScenes val split.

| | stage | PQ | SQ | RQ | mIoU |
|--------------------|---------|------|------|------|------|
| LidarMultiNet [66] | 2-stage | 81.8 | 90.8 | 89.7 | 83.6 |
| LiDARFormer | 1-stage | 81.8 | 90.7 | 89.9 | 84.1 |

Table 9: Design choice of the segmentation decoder on the nuScenes val split.

| LiDARFormer seg only result without XSF (mIoU) | 80.8 |
|--|--|
| w/o voxel to class attention | 80.4 (-0.4) |
| w/o class to voxel attention | 80.1 (-0.7) |
| w/o dynamic kernel | 80.3 (-0.5) |
| w/o class embedding initialization | 80.4 (-0.4) 80.1 (-0.7) 80.3 (-0.5) 80.5 (-0.3) |

formance, particularly for the detection task. We also evaluate the panoptic segmentation performance of our multitask network in Table 8. Even without a second stage dedicated to panoptic segmentation, our model achieves competitive results compared to the previous best method, LidarMultiNet. This demonstrates the ability of our multi-task transformer decoder to generate more compatible results for both tasks.

Table 7: The ablation of the improvement of shared transformer decoder on the nuScenes val split when jointly trained with detection task.

| Baseline [66] | XTF Seg Det | | XSF | mIoU | mAP | NDS | |
|---------------|----------------|--------------|--------------|-------------|--------------------------|----------------------|--|
| √ | | | | 81.8 | 65.2 | 70.0 | |
| ✓ | ✓ | | | 82.1 (+0.3) | 65.4 (+0.2) | 70.2 (+0.2) | |
| ✓ | | \checkmark | | 82.4 (+0.6) | 65.9 (+0.7) | 70.3 (+0.3) | |
| ✓ | ✓ | \checkmark | | 82.6 (+0.8) | 66.0 (+0.8) | 70.2 (+0.2) | |
| ✓ | ✓ | \checkmark | \checkmark | 82.7 (+0.9) | 66.6 (+1.4) | 70.8 (+0.8) | |

Table 10: The ablation of XSF on the nuScenes val split. $S\rightarrow D$ and $D\rightarrow S$ denote sparse-to-dense (3b) and dense-to-sparse (3a) XSFs.

| $S \rightarrow D$ | $D \rightarrow S$ | Add Convs | mIoU | mAP | NDS | | | |
|---------------------------|-------------------|--------------|-------------|--------------------------|-------------|--|--|--|
| Segmentation Only | | | | | | | | |
| $\overline{\hspace{1em}}$ | √ | | 81.7 | - | - | | | |
| | | \checkmark | 80.9 (-0.8) | - | - | | | |
| Multi-task | | | | | | | | |
| $\overline{\hspace{1cm}}$ | √ | | 82.7 | 66.6 | 70.8 | | | |
| | \checkmark | ✓ | 82.8 (+0.1) | 66.0 (-0.6) | 70.5 (-0.3) | | | |

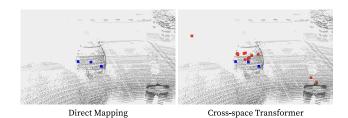


Figure 5: **Visualization of the learned offsets.** We showcase the features of a car's 3D voxels (blue) and their corresponding deformable offsets (red) that were learned in our XSF module. For a better visual representation, we only highlight the offsets with high attention scores.