

Dense Captioning for 3D Scenes

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Motivation

it is a black office chair next to the round table. it is against the wall.

this is a round table. it is between two black chairs, it is in the center of this room.

this is a black office chair. it is in the corner next to a black chair.

this is a black office chair. it is against the wall and facing the round table.

it is a black office chair. it is facing the round table.

this is a small trashcan. it is in front of a glass window.

it is a glass window. it is set next to the wooden door.

it is a brown door next to a window. it is wooden and closed.

Input: 3D Point Clouds

| | Object Detection | | |
|---------------------|------------------|--|--|
| Baseline (Scan2Cap) | VoteNet | | |
| Ours | SoftGroup | | |

Bounding Boxes and Descriptions

Better Object Detection

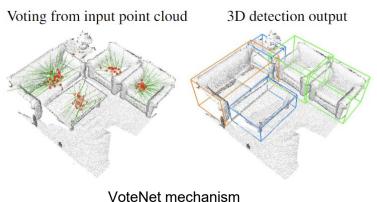
↓

Better Object features

Better Object Captions



Motivation



color map Otherfurniture Cabinet Classification → Cabinet Soft Grouping Classification → Background

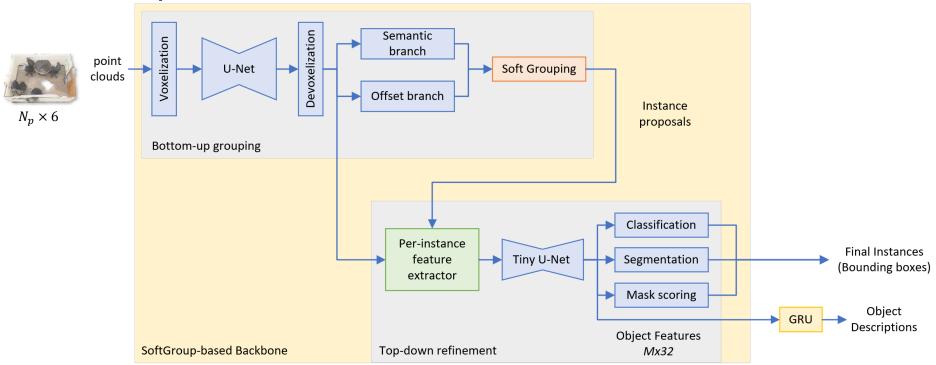
SoftGroup mechanism

| | mAP@0.25loU | mAP@0.5loU | |
|-----------|-------------|------------|--|
| VoteNet | 58.6 | 33.5 | |
| SoftGroup | 71.6 | 59.4 | |

Semantic

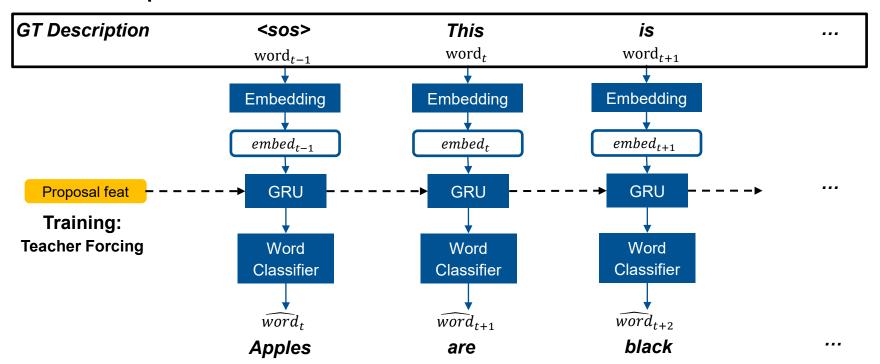


SoftGroup-based network architecture



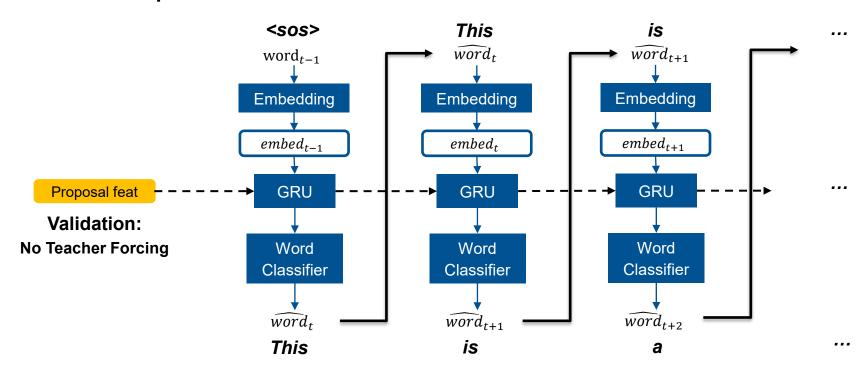


SoftGroup-based network architecture





SoftGroup-based network architecture

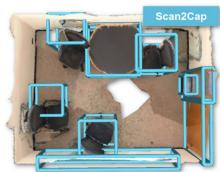




Training and Inference results

| | Network Architecture | CIDEr @0.5loU | BLEU-4 @0.5loU | METEOR @0.5loU | ROUGE @0.5loU | Box mAP @0.5loU |
|----------|-------------------------|------------------|-------------------|-------------------|------------------|--------------------|
| Scan2Cap | VoteNet+GRU | 34.31 | 21.42 | 20.13 | 41.33 | 32.21 |
| Ours | SoftGroup+GRU | 34.70 | 24.46 | 22.23 | 46.95 | 45.56 |







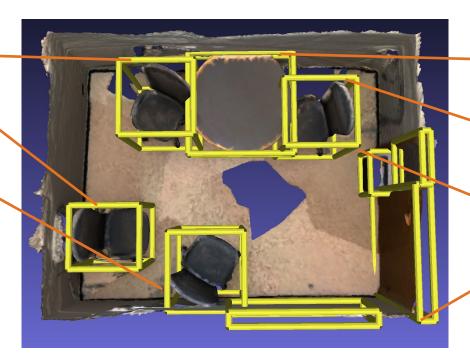


Training and Inference results

This is a black office chair. It is in the corner of the room.

This is a black office chair. It is in the corner of the room.

This is a black office chair. It is in the corner of the room.



This is a brown table. It is in the center of the room.

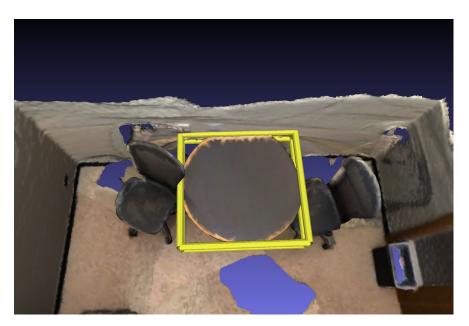
This is a black office chair. It is in the corner of the room.

The trash can is blue. It is located to the right of the trash can.

The door is white. It is located to the right of the trash can



Training and Inference results



GT:

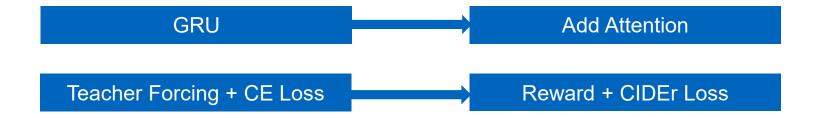
the **round table** with the **black top** is **in the corner** of the room. **two black office chairs around** the table.

Inference:

This is a **brown table**. It is **in the center of** the **room**



Challenges and Future works





Thank you for your attention!

Any Question?



Sparse Convolution Loss Formula

- - -

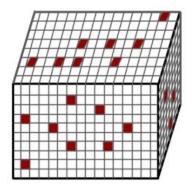


Sparse Convolution

- Regular: calculate output when kernel covers an active input
- Submanifold: calculate output only when the kernel center covers an active input

Reason:

- In 3D Space, many voxels are empty, so the point clouds data are always sparse.
- Using Sparse Convolution can help us calculate the feature more efficiently





Loss Formula

$$L_{\text{semantic}} = \frac{1}{N} \sum_{i=1}^{N} \text{CE}(\boldsymbol{s}_i, s_i^*),$$

$$L_{ ext{offset}} = rac{1}{\sum_{i=1}^{N} \mathbb{1}_{\{m{p}_i\}}} \sum_{i=1}^{N} \mathbb{1}_{\{m{p}_i\}} \|m{o}_i - m{o}_i^*\|_1,$$

$$L_{\mathrm{class}} = \frac{1}{K} \sum_{k=1}^{K} \mathrm{CE}(\boldsymbol{c}_k, c_k^*),$$

$$L_{\text{mask}} = \frac{1}{\sum_{k=1}^{K} \mathbb{1}_{\{\boldsymbol{m}_k\}}} \sum_{k=1}^{K} \mathbb{1}_{\{\boldsymbol{m}_k\}} \text{BCE}(\boldsymbol{m}_k, \boldsymbol{m}_k^*),$$

$$L_{ ext{mask_score}} = rac{1}{\sum_{k=1}^{K} \mathbb{1}_{\{m{e}_k\}}} \sum_{k=1}^{K} \mathbb{1}_{\{m{e}_k\}} \|m{e}_k - m{e}_k^*\|_2.$$

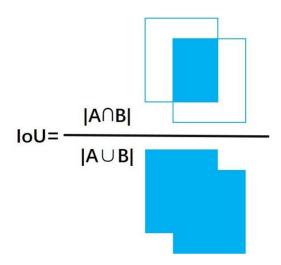
GT:

- Semantic: GT semantic class of the point
- Offset: GT offset vector that shifts the point to corresponding instance center
- Classification: class of the GT instance with highest IoU
- Mask: the mask of the assigned GT instance
- Mask score: IoU between the predicted mask and the GT

We treat all instance proposals having IoU with a ground-truth instance higher than 50% as the positive samples and the rest as negatives.



IoU: Intersection of Union





Program Results (NLP metrics & Box mAP@0.5IoU)

```
loading corpus...
generating descriptions...
computing scores...
BLEU-1,2,3,4 are: [0.5248459993766885, 0.41716240214357864, 0.31902357245013113, 0.24464104223299268]
CIDEr is: 0.3470318672484949
BLEU-4 is: 0.24464104223299268
METEOR is: 0.22290531565740027
ROUGE is: 0.4694894696899399
```

```
Evaluating...
mAP: 0.4555707314393957
```



Description Results (>0.5 bounding box iou)

```
"scene0427_00|5|chair": [
   "sos this is a black chair . it is in the corner of the room . eos"
"scene0427 00|8|door": [
   "sos the door is white . it is located to the right of the trash can . eos"
"scene0427 00|0|trash can": [
   "sos the trash can is blue . it is located to the right of the trash can . eos"
"scene0427 00|7|chair": [
   "sos this is a black chair . it is in the corner of the room . eos"
"scene0427 00|9|table": [
   "sos this is a brown table . it is in the center of the room . eos"
"scene0427 00|4|chair": [
   "sos this is a black chair . it is in the corner of the room . eos"
"scene0427_00|6|chair": [
    "sos this is a black chair . it is in the corner of the room . eos"
"scene0427_00|2|window": [
    "sos eos"
```