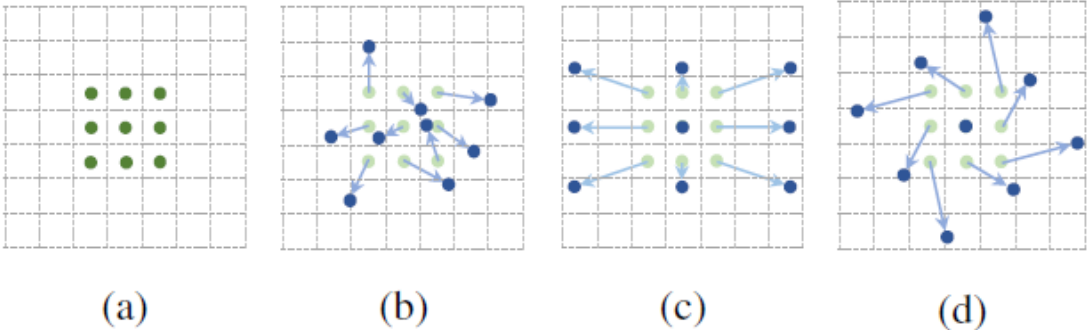


# Deformable convolution and adaptive key points

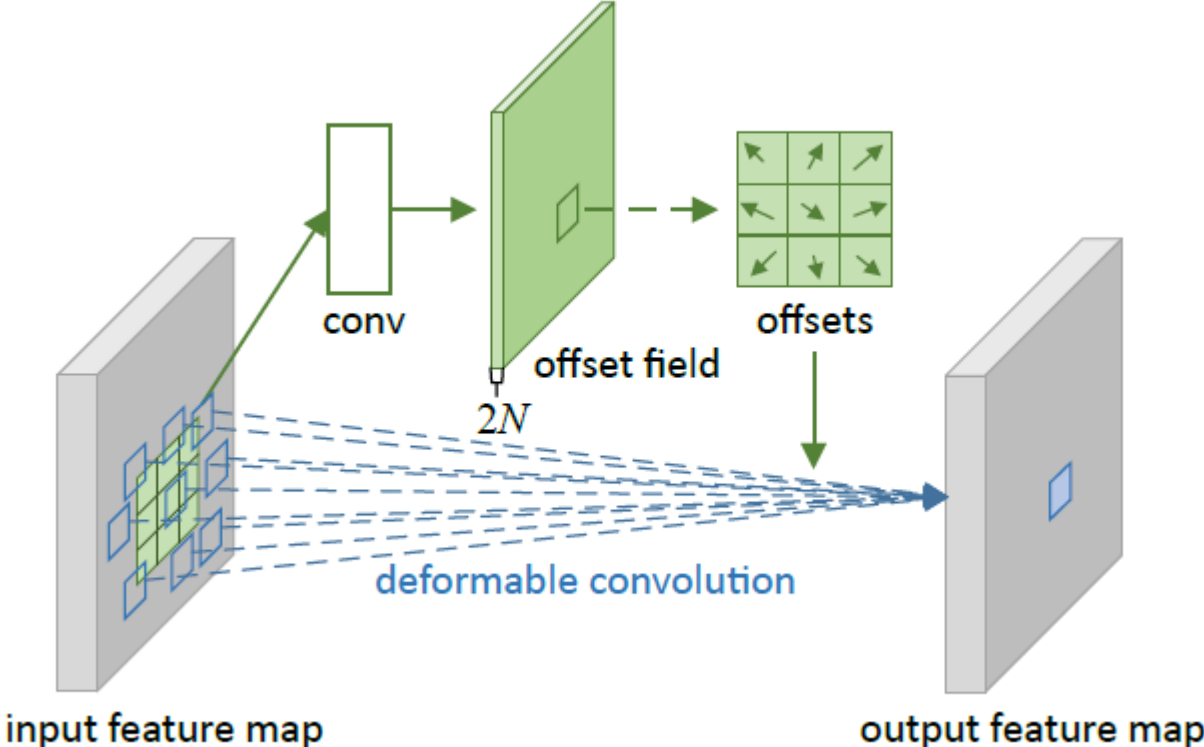
Outline:

- Deformable convolution network(DCN) and DCNv2
- RepPoint for object detection on image
- FGFA and STSN for video application

# Deformable convolution



Regular convolution vs deformable convolution



Realization of deformable convolution

Reference:

- 1. Deformable Convolutional Networks
- 2. Deformable ConvNets v2: More Deformable, Better Results

# Result of deformable convolution

DCN v1 result



background

foreground

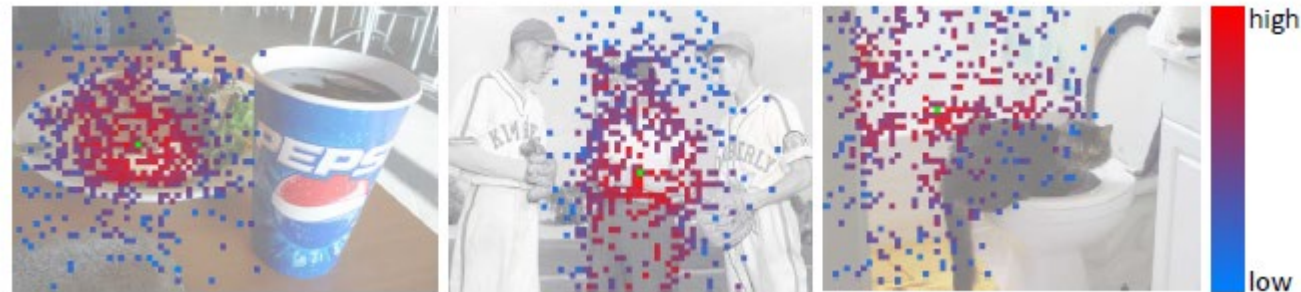
foreground

background

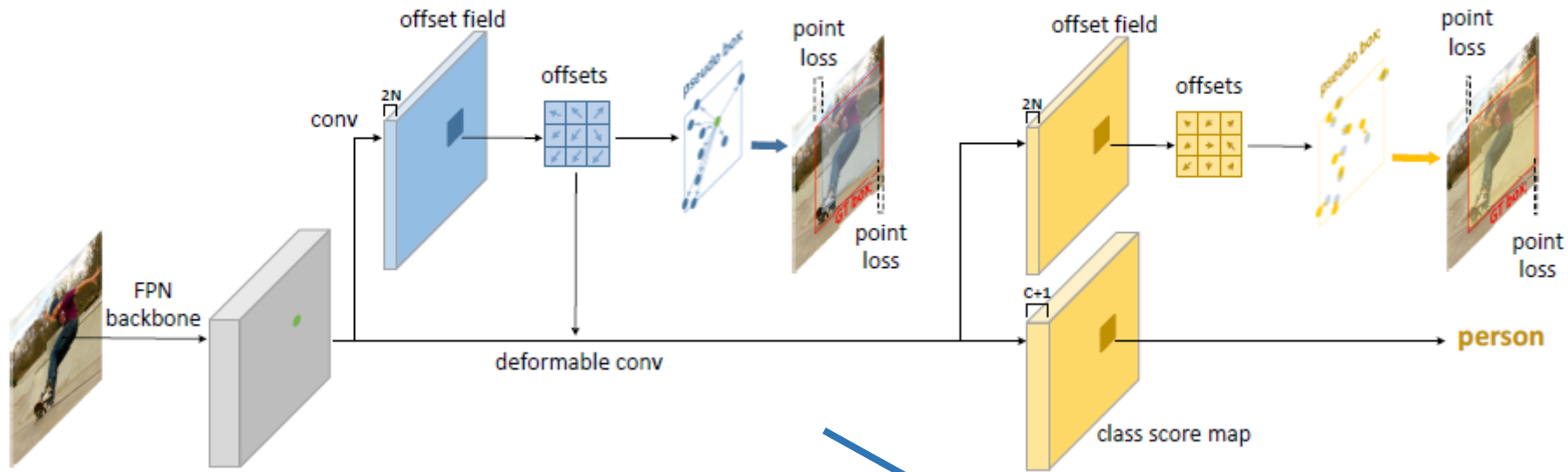
foreground

foreground

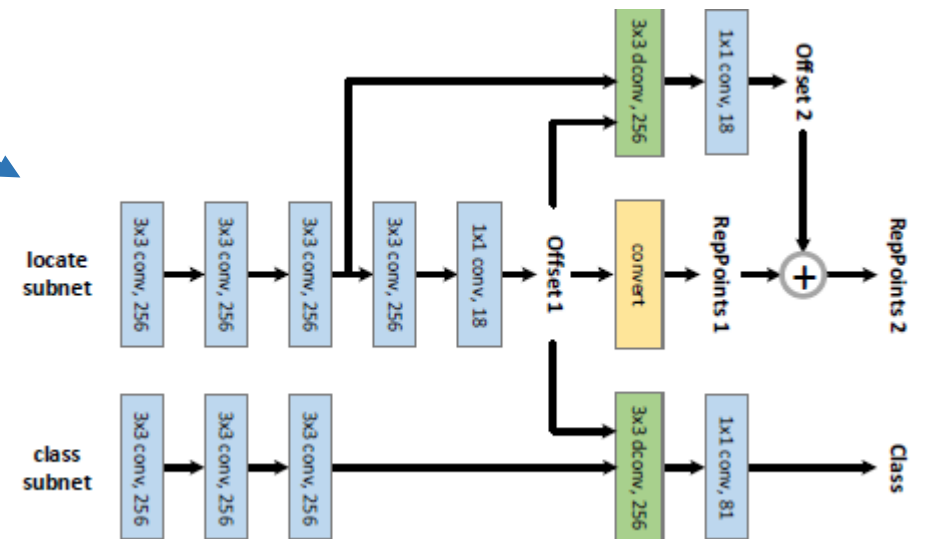
DCN v2 result



# RepPoints



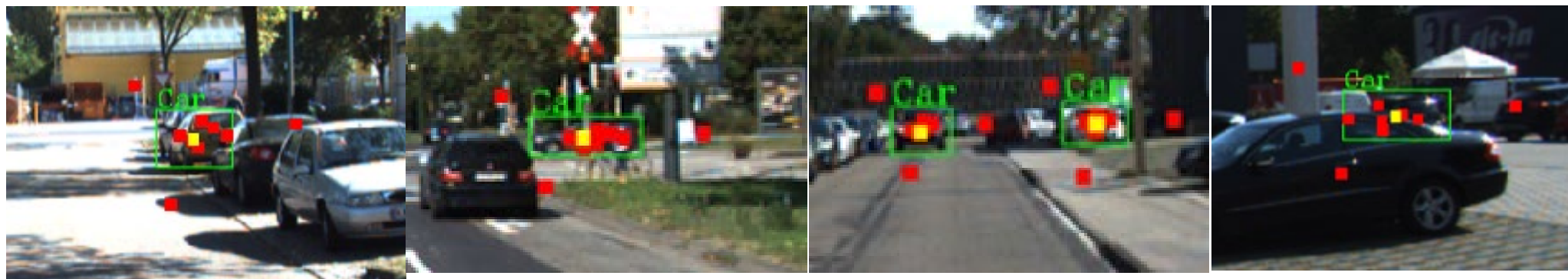
## Realization of RepPoints



## RepPoints Head



# RepPoints Result on KITTI



Car

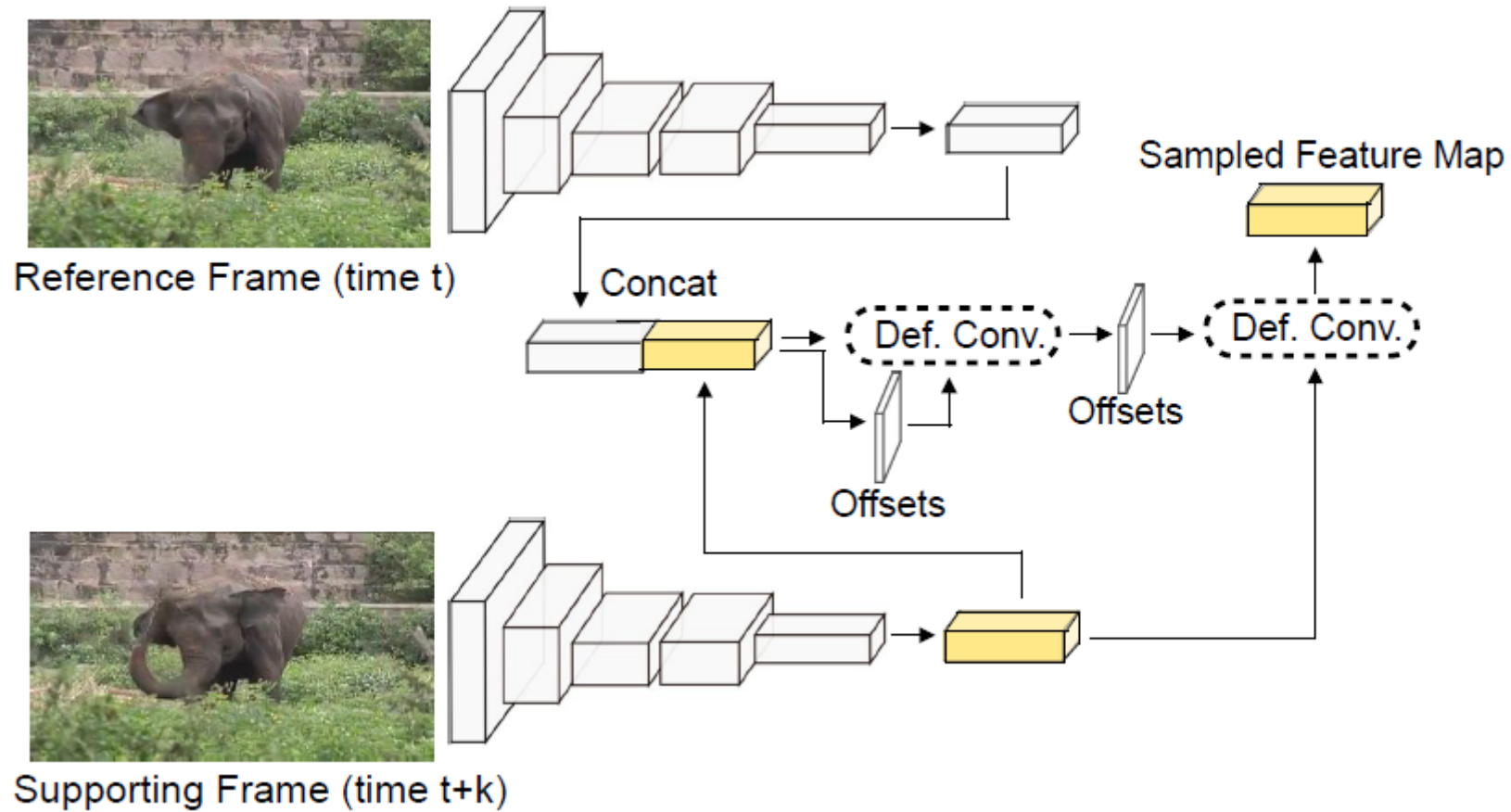


Car



Person

# Spatiotemporal Sampling Networks(STSN)



# STSN result on VID



Reference Frame (t)

Supporting Frame (t+9)

Reference Frame (t)

Supporting Frame (t+9)



Supporting Frame (t-9)

Supporting Frame (t-4)

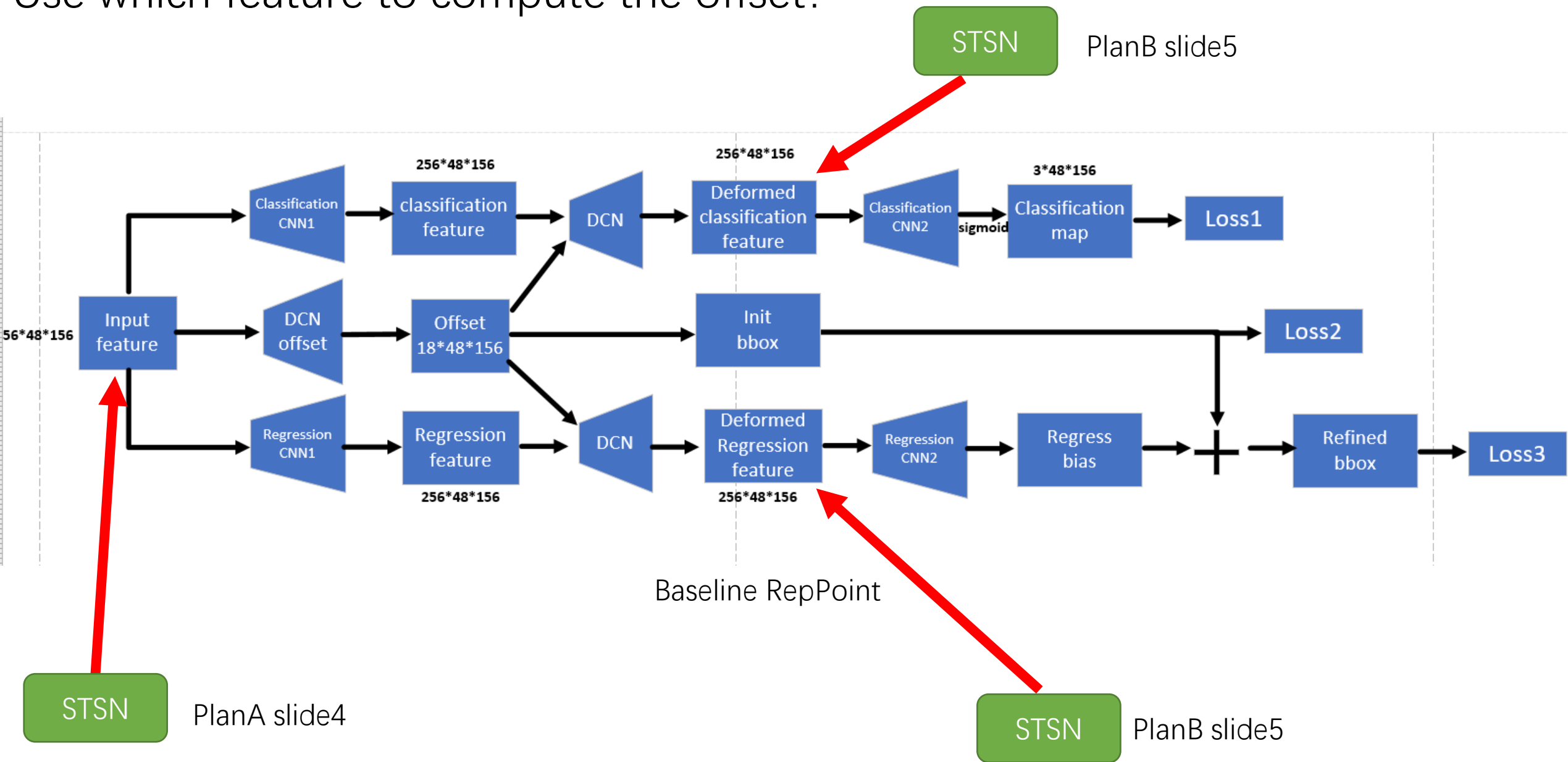
Reference Frame (t)

Supporting Frame (t+4)

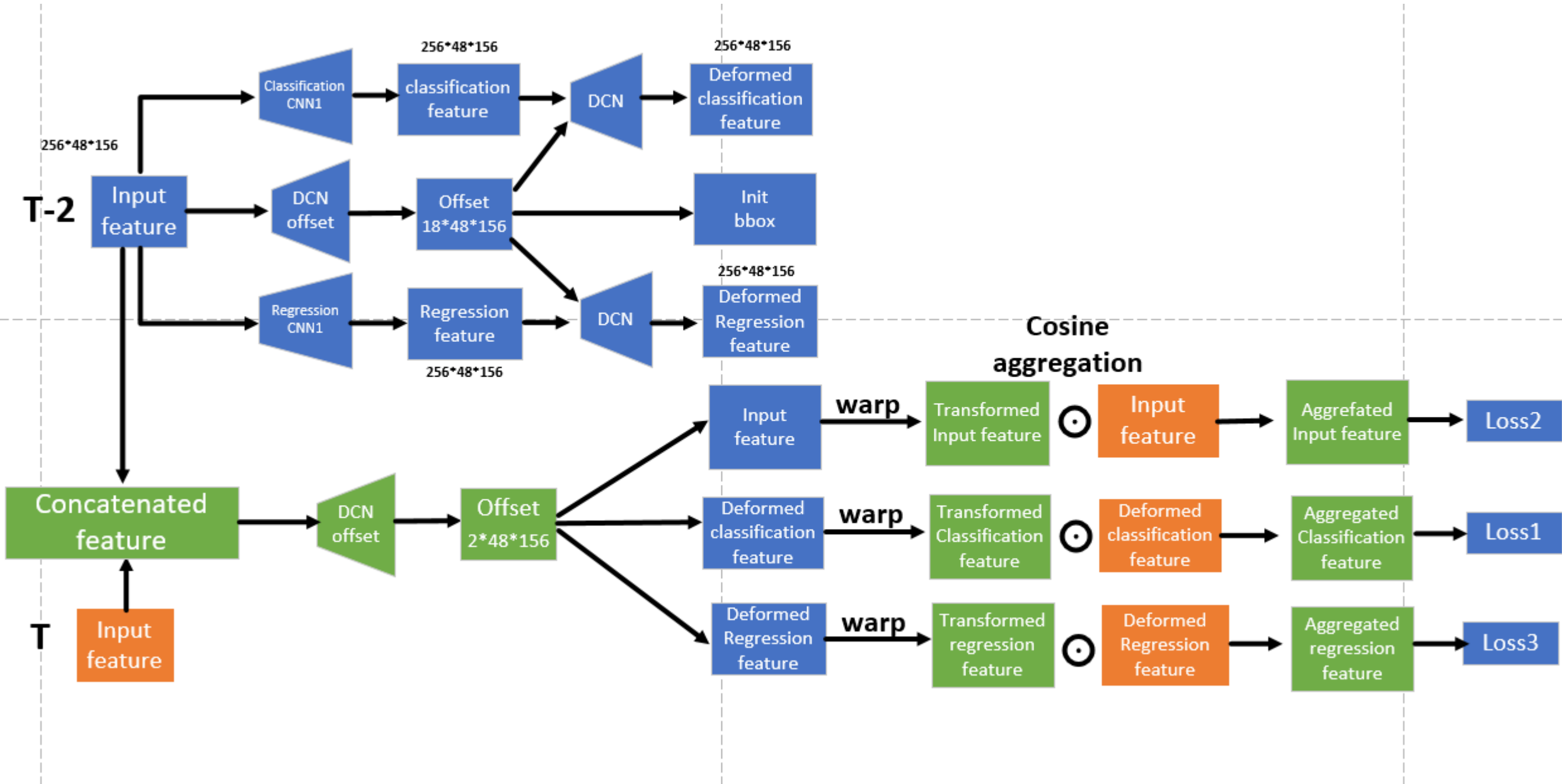
Supporting Frame (t+9)



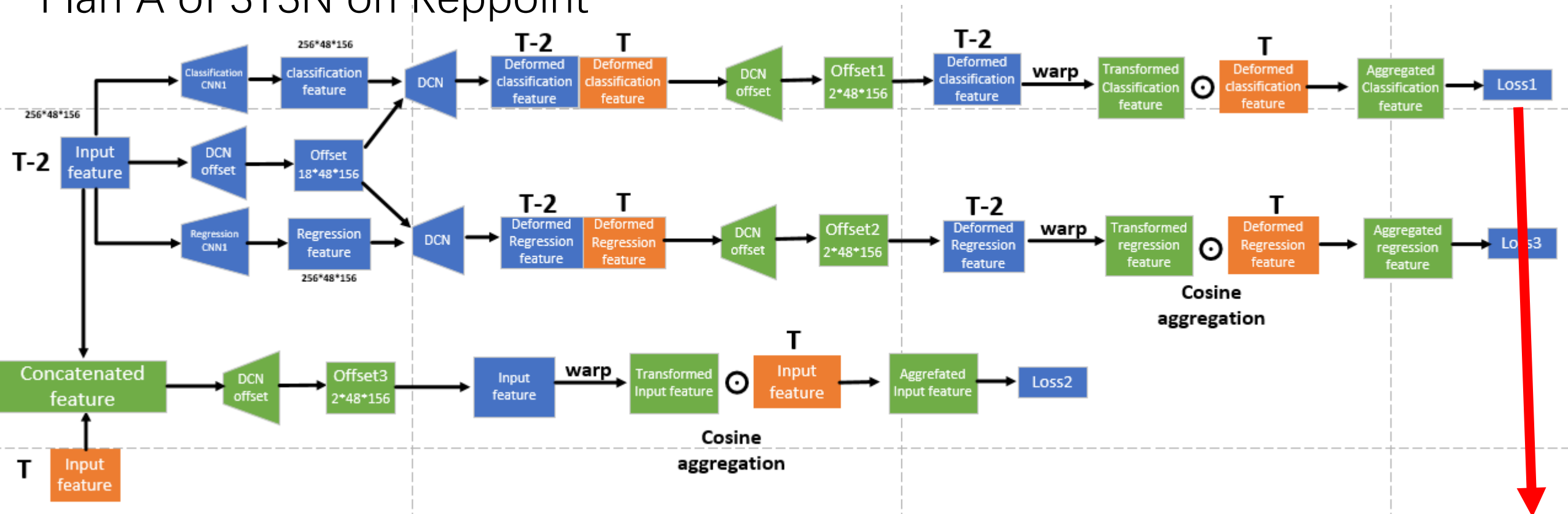
Use which feature to compute the offset?



# Plan A of STSN on Reppoint, result on Slides 2



# Plan A of STSN on Reppoint



Old Result  
Need  
update

evaluating result of refer

class	gts	dets	recall	precision	ap
Vehicle	10747	14870	0.794	0.574	0.743
Pedestrian	9429	11400	0.678	0.560	0.620
Cyclist	509	1441	0.497	0.176	0.187
mAP					0.517

baseline

evaluating result of agg epoch29 from the pretrained epoch9 optimize whole

class	gts	dets	recall	precision	ap
Vehicle	10747	26993	0.876	0.349	0.774
Pedestrian	9429	15771	0.675	0.404	0.598
Cyclist	509	4518	0.701	0.079	0.297
mAP					0.556

Only warp the classification feature

# New KITTI split

Before

	Vehicle	Person	Cyclist
Train	20680	2740	#
test	9921	9406	#

Now

	Vehicle	Person	Cyclist
Train	22543	8702	1354
test	8058	3444	584

train frames: 5976

test frames: 2032

all frames: 8008

```
evaluating result of refer
+-----+-----+-----+-----+-----+
| class   | gts   | dets  | recall | precision | ap   |
+-----+-----+-----+-----+-----+
| Vehicle | 10747 | 14870 | 0.794  | 0.574     | 0.743 |
| Pedestrian | 9429 | 11400 | 0.678  | 0.560     | 0.620 |
| Cyclist  | 509   | 1441  | 0.497  | 0.176     | 0.187 |
+-----+-----+-----+-----+-----+
| mAP     |       |       |        |           | 0.517 |
+-----+-----+-----+-----+-----+
```

Baseline

```
evaluating result of baseline
+-----+-----+-----+-----+-----+
| class   | gts   | dets  | recall | precision | ap   |
+-----+-----+-----+-----+-----+
| Vehicle | 8058  | 10070 | 0.954  | 0.764     | 0.947 |
| Pedestrian | 3444 | 5463  | 0.825  | 0.520     | 0.789 |
| Cyclist  | 584   | 1089  | 0.777  | 0.417     | 0.739 |
+-----+-----+-----+-----+-----+
| mAP     |       |       |        |           | 0.825 |
+-----+-----+-----+-----+-----+
```

Baseline

This performance might be too high to improve. Lixin just talked to me today that he also need the BDD data, so I also re-split the BDD tracking data.

# Training and test on the new split KITTI dataset

evaluating result of baseline

class	gts	dets	recall	precision	ap
Vehicle	8058	10070	0.954	0.764	0.947
Pedestrian	3444	5463	0.825	0.520	0.789
Cyclist	584	1089	0.777	0.417	0.739
mAP					0.825

Baseline  
Figure is in  
Slides 3

evaluating result of agg epoch19

class	gts	dets	recall	precision	ap
Vehicle	8058	11140	0.958	0.693	0.950
Pedestrian	3444	6706	0.826	0.424	0.781
Cyclist	584	1276	0.769	0.352	0.725
mAP					0.819

evaluating result of agg epoch 39

class	gts	dets	recall	precision	ap
Vehicle	8058	10799	0.962	0.717	0.953
Pedestrian	3444	6033	0.817	0.467	0.776
Cyclist	584	1243	0.765	0.360	0.731
mAP					0.820

evaluating result of agg epoch29

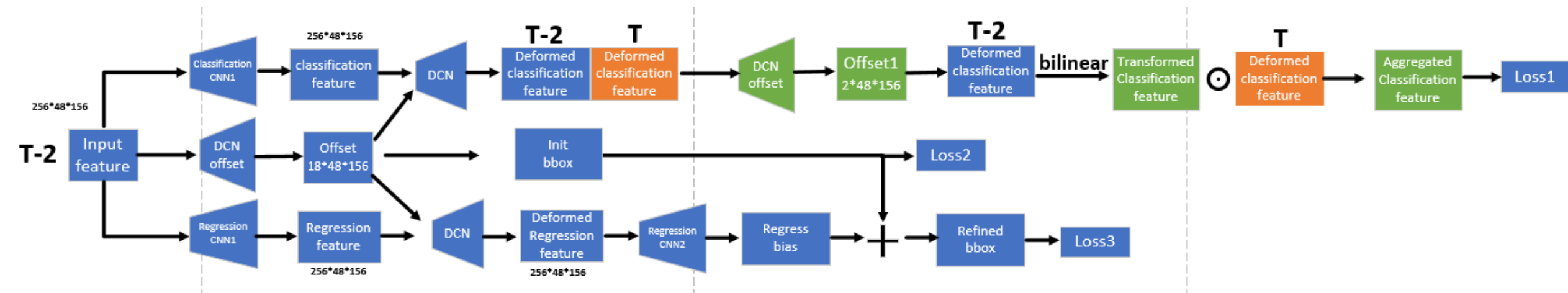
class	gts	dets	recall	precision	ap
Vehicle	8058	10546	0.949	0.725	0.942
Pedestrian	3444	3998	0.713	0.614	0.654
Cyclist	584	927	0.740	0.466	0.707
mAP					0.767

evaluating result of agg epoch38

class	gts	dets	recall	precision	ap
Vehicle	8058	10188	0.957	0.757	0.949
Pedestrian	3444	5574	0.827	0.511	0.785
Cyclist	584	1056	0.779	0.431	0.742
mAP					0.825

STSN  
Figure is in  
Slides 4

# One offset warp the classification feature



# Comparison between different feature to compute the offset

```
evaluating result of baseline
```

class	gts	dets	recall	precision	ap
Vehicle	8058	10070	0.954	0.764	0.947
Pedestrian	3444	5463	0.825	0.520	0.789
Cyclist	584	1089	0.777	0.417	0.739
mAP					0.825

baseline

```
evaluation of agg classification baseline epoch19
```

class	gts	dets	recall	precision	ap
Vehicle	8058	11994	0.965	0.648	0.956
Pedestrian	3444	6000	0.823	0.472	0.786
Cyclist	584	1507	0.791	0.307	0.737
mAP					0.827

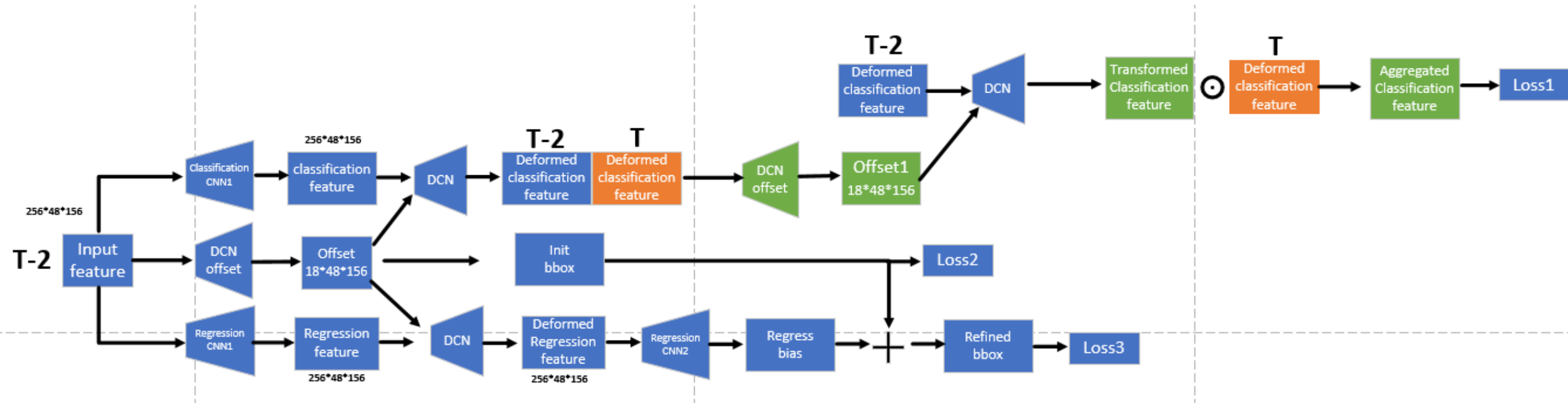
One offset warp the classification feature

```
evaluating result of agg epoch19
```

class	gts	dets	recall	precision	ap
Vehicle	8058	11140	0.958	0.693	0.950
Pedestrian	3444	6706	0.826	0.424	0.781
Cyclist	584	1276	0.769	0.352	0.725
mAP					0.819

Use backbone feature to compute the offset and warp the classification and regression feature.

# Analyze of the fail case of stsn and compare with reppoint



The stsn version I analyze in this report

With learnable dcn to warp the support feature  
Support frames are  $N-2$  and  $N+2$   
Aggregation between  $N$ ,  $N-2$  and  $N+2$   
Fix the reppoint weight



# Car -- not-detected bbox and offset

Green bbox is the ground bbox not-detected.

Yellow point is the center of the bbox

Red points are the offset of the yellow point.

The result is checked frame by frame, if the bbox only shows on frame 3 and frame 5, that means frame 4 detects all objects.

Vehicle class consists of car, van and trunk classes of kitti

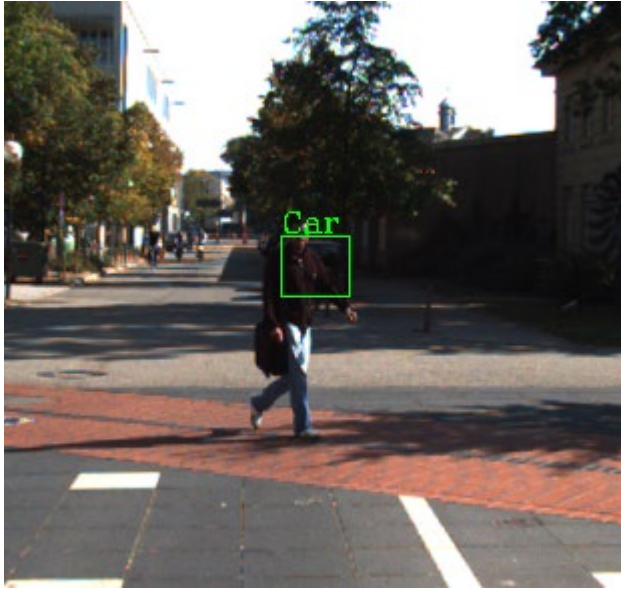
Support1 is N-2 frame, support2 is N+2 frame

Reppoint+stsn is two dcn layers, so we get 81 offsets to show

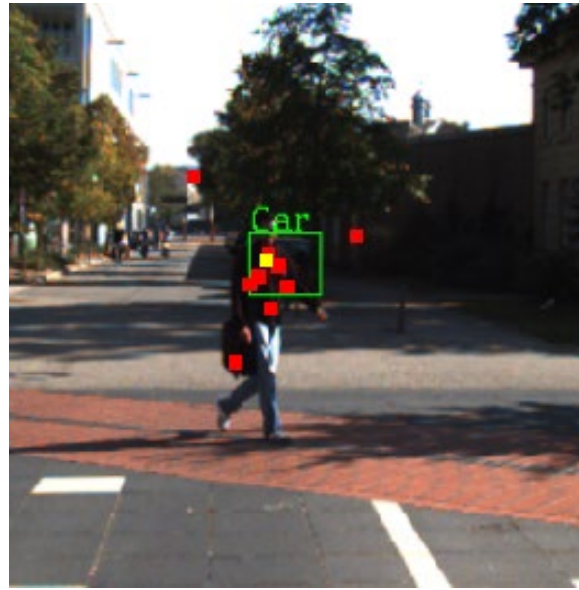
	Vehicle	Person	Cyclist
Train	22543	8702	1354
test	8058	3444	584

- First show failure case of reppoint but can be detected by stsn
- Then the failure case of stsn

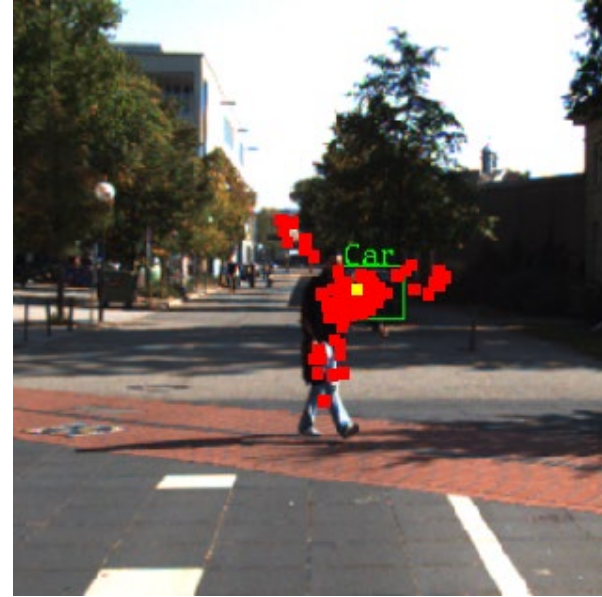
# Stsn gets not-detected bbox of reppoint



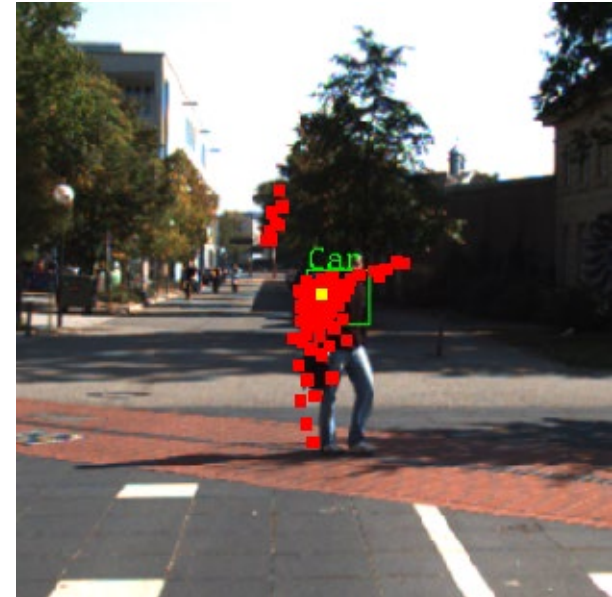
Reppoint not-detected



Reppoint offset



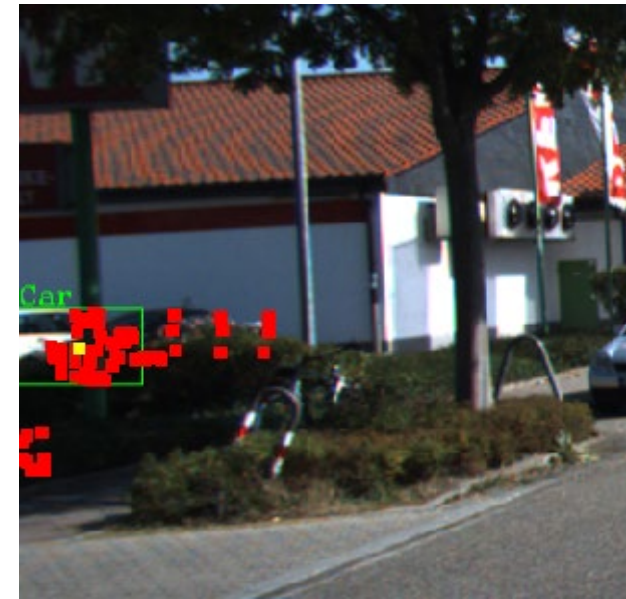
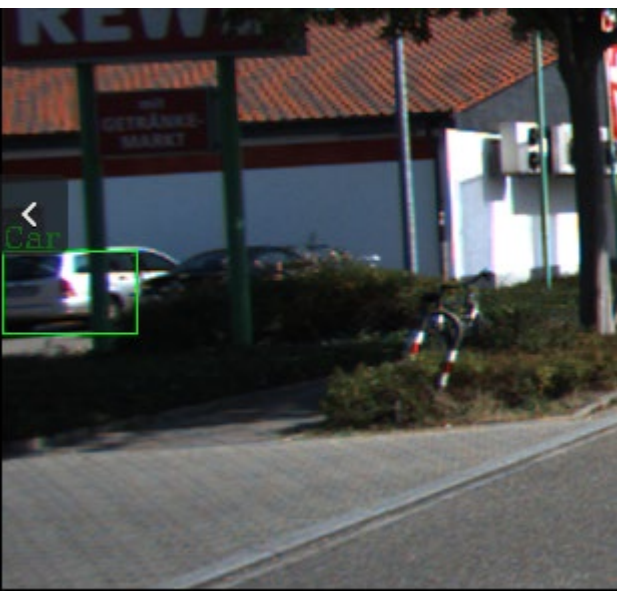
support1 offset



support2 offset

The car is highly occluded by the person, most reppoints are on the person  
Stsn adds more points on the car.

# Stsn gets not-detected bbox of reppoint



Reppoint not-detected

Reppoint offset

support1 offset

support2 offset

Stsn adds a denser sampling on the car from N-2 frame

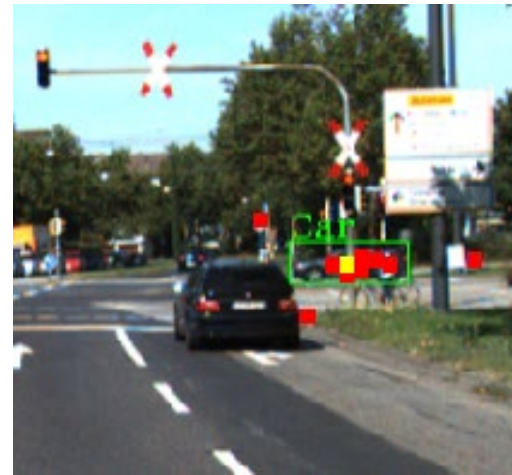
It seems the reppoint allow offsets lie on background or other object, these offset may have very low weight, which are shown as grid on the image.

# Stsn gets not-detected bbox of reppoint

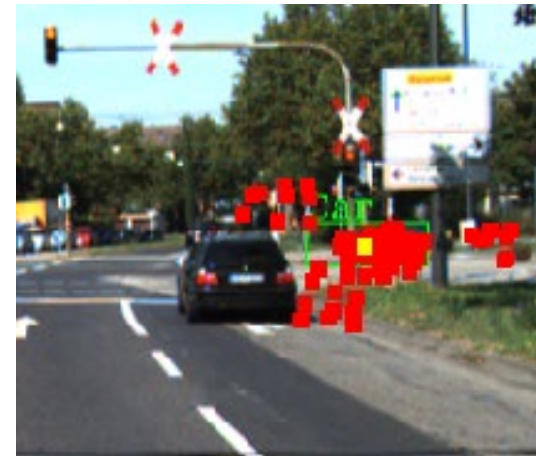
Frame  
74



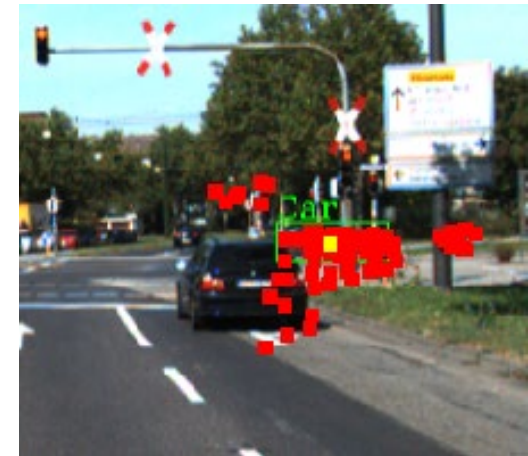
Reppoint not-detected



Reppoint offset

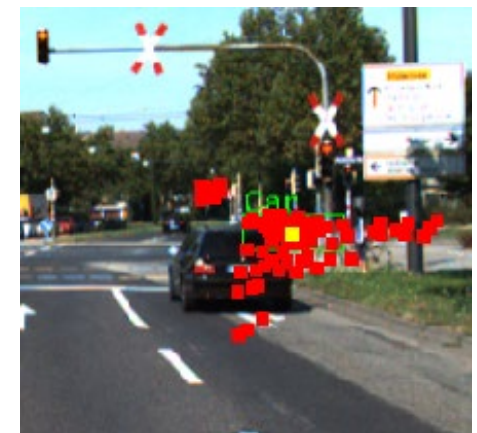
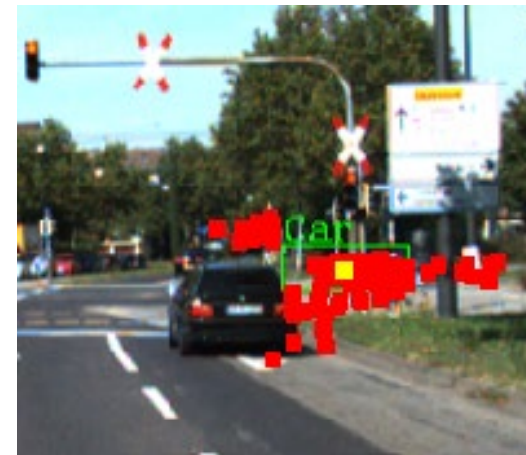


support1 offset



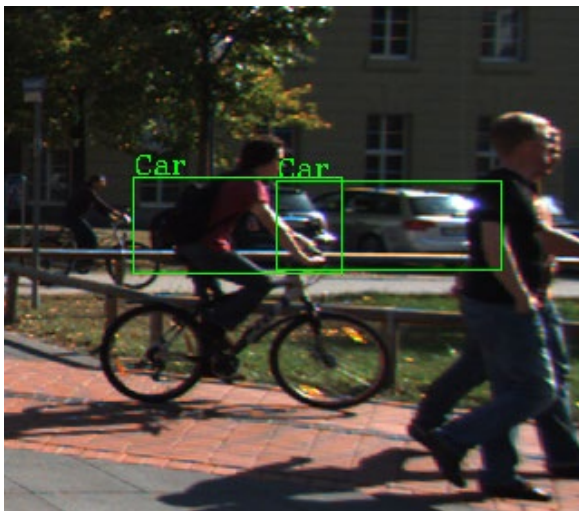
support2 offset

Frame  
75

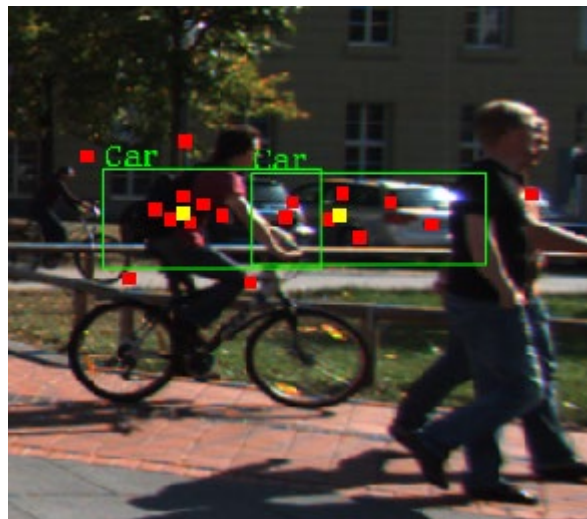


On small objects, the offset visualization is hard to see  
Im working on the one offset warp model for the visualization.

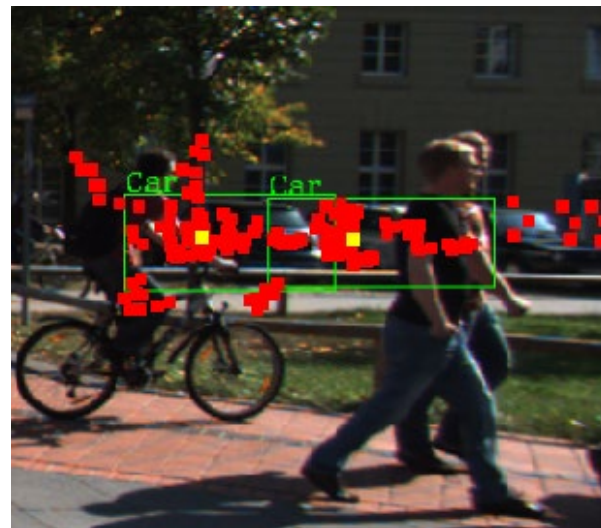
# Stsn get not-detected bbox of reppoint



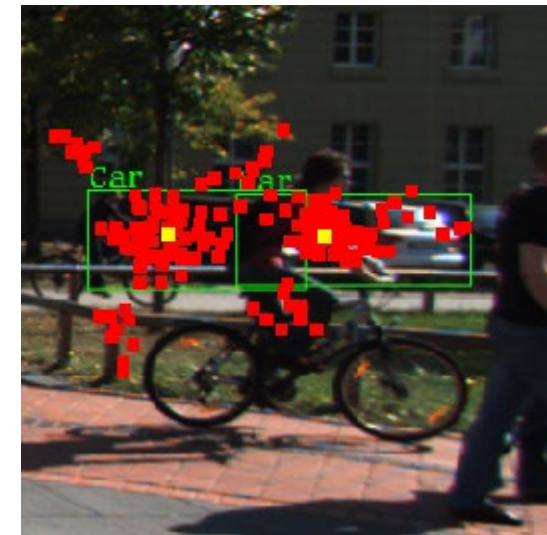
Reppoint not-detected



Reppoint offset



support1 offset



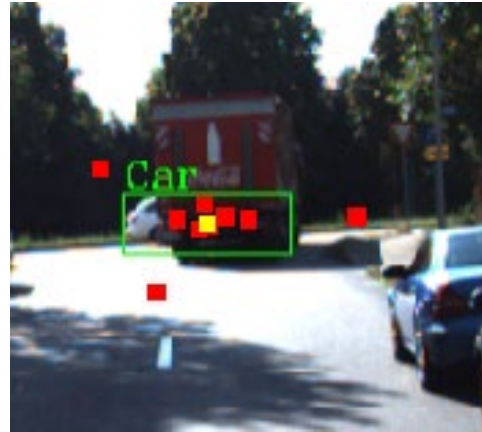
support2 offset

For the left car, on N+2, the stsn brings into the right signal

# Stsn get not-detected bbox of reppoint



Reppoint not-detected



Reppoint offset



support1 offset

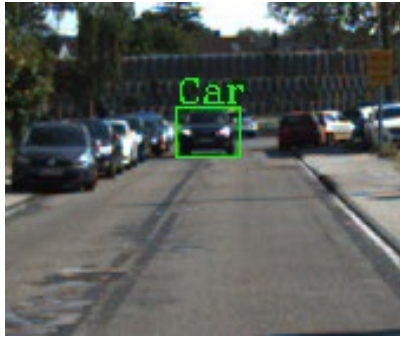


support2 offset

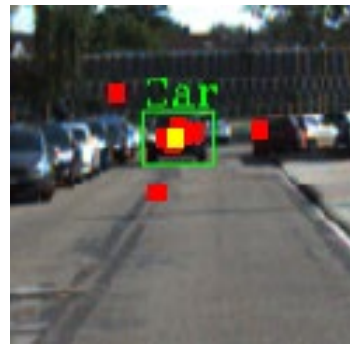
- It is hard to say why it works. Some points from  $N+2$  frame is right, but I don't think they get a high weight.
- Maybe the offset I visualized has a bias, the yellow point should be on the left rather than on the center.
- On the  $N+2$  frame, we can see the offset is trying to find the signal from white color.

# Stsn get not-detected bbox of reppoint

Frame  
26



Reppoint not-detected



Reppoint offset

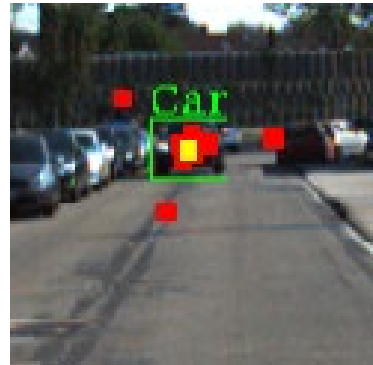
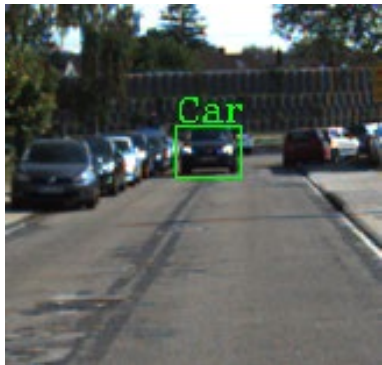


support1 offset



support2 offset

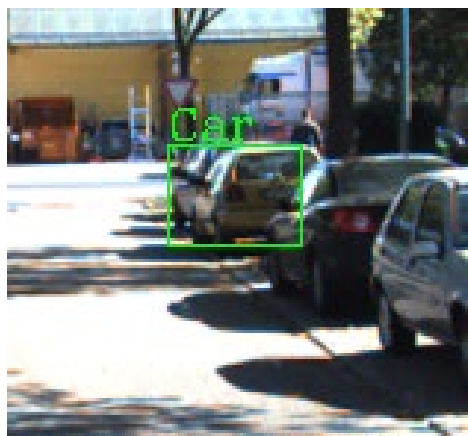
Frame  
28



Signal is obviously enhanced by the stsn, the ratio of point on the car is higher than only using one frame.

# fail case of stsn

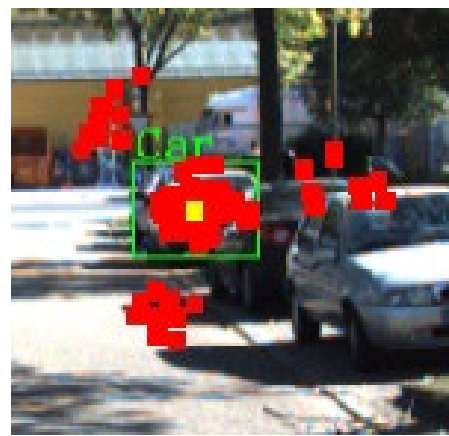
Frame  
119



Stsn not-detected



Reppoint offset

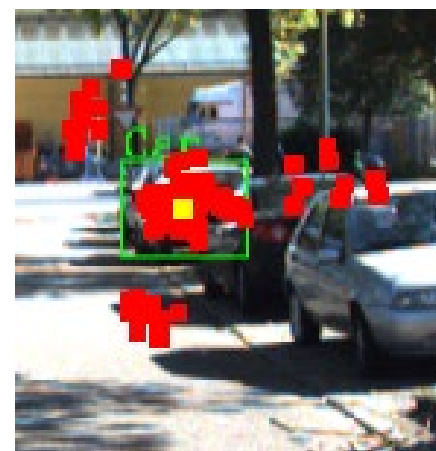
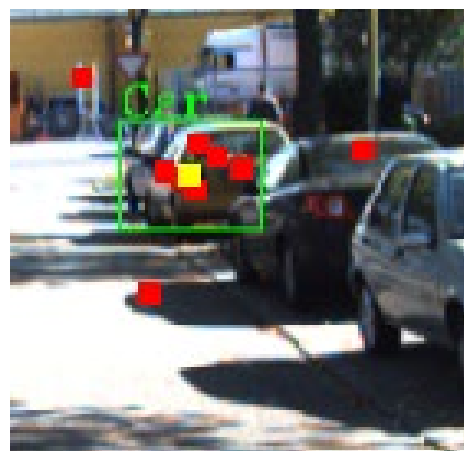


support1 offset



support2 offset

Frame  
120



The offset seems right.

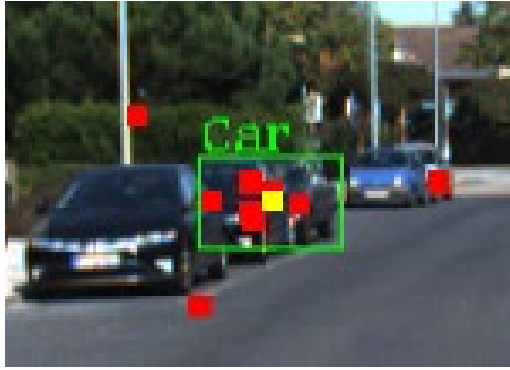
This high occlusion may not be the detection problem but the nms kill the occluded bbox.



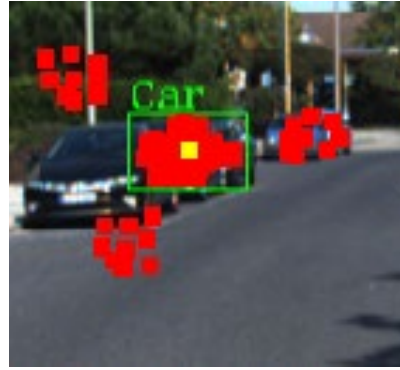
# fail case of stsn



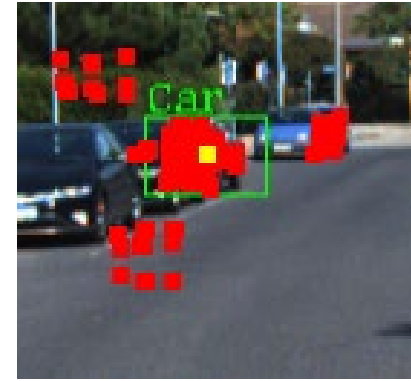
stsn not-detected



Reppoint offset

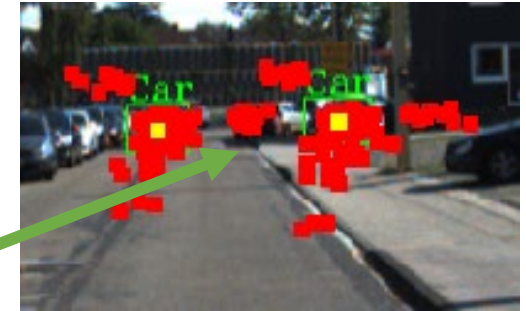
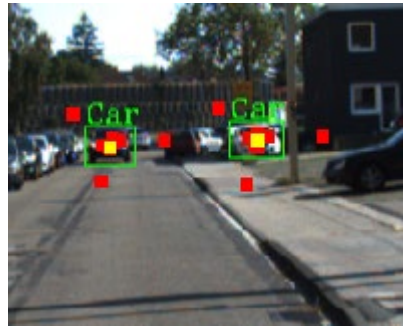
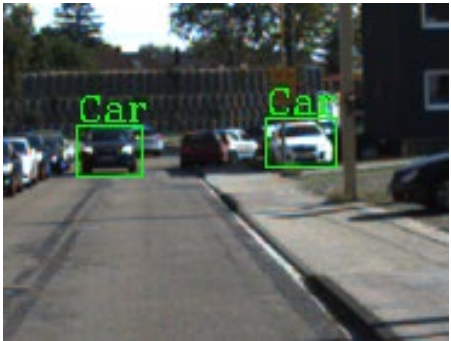


support1 offset



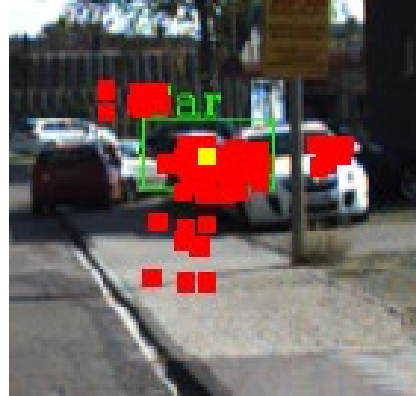
support2 offset

Maybe two cars fused as a object not a car?



Again, for small object, too much signal is aggregated.  
But we can see the offset trying to find the signal on cars.

# fail case of stsn

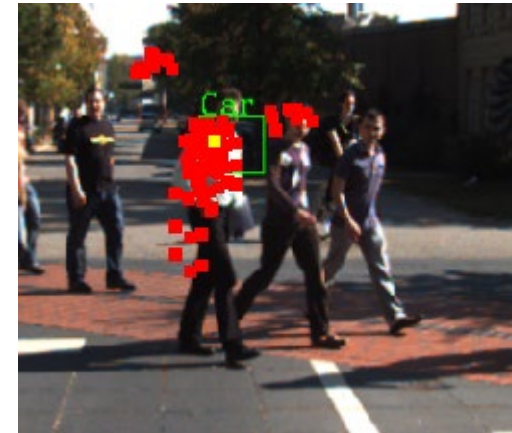
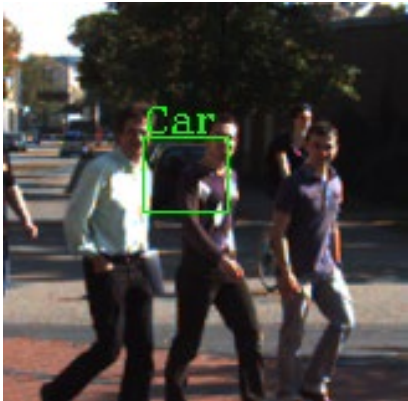


stsn not-detected

Reppoint offset

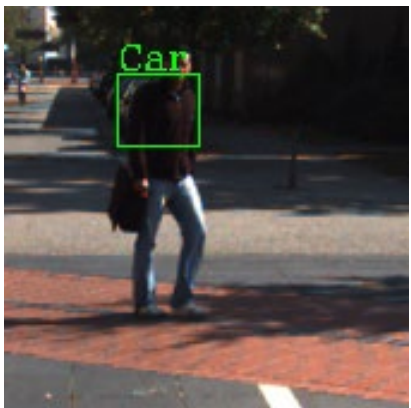
support1 offset

support2 offset

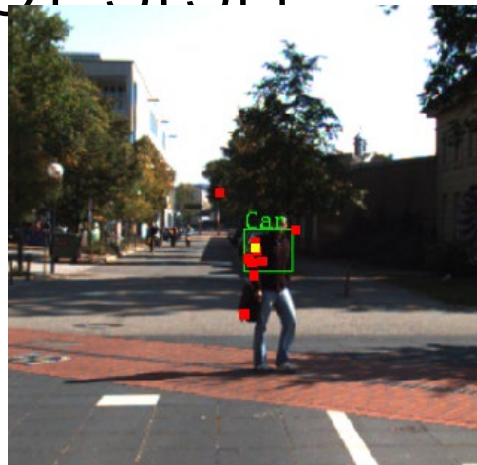


- For car occluded by car, the two cars should have similar color, this is also we can see from slides7.
- For car occluded by person, it will fail if the car is similar to the background and the person is good visible, offsets will lie on strong signal on the person.

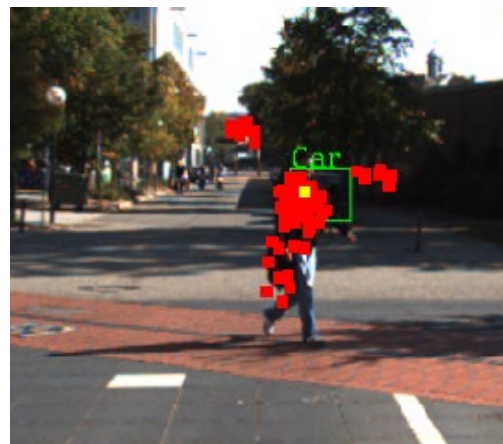
# fail case of stsn



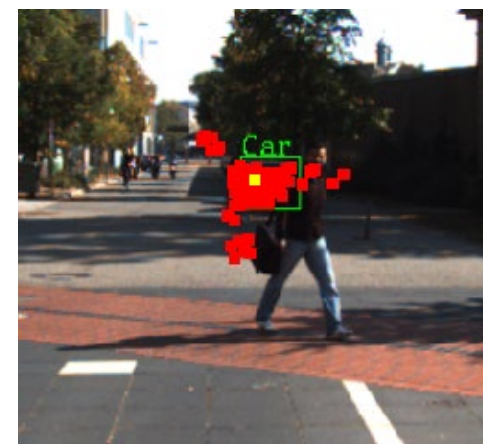
stsn not-detected



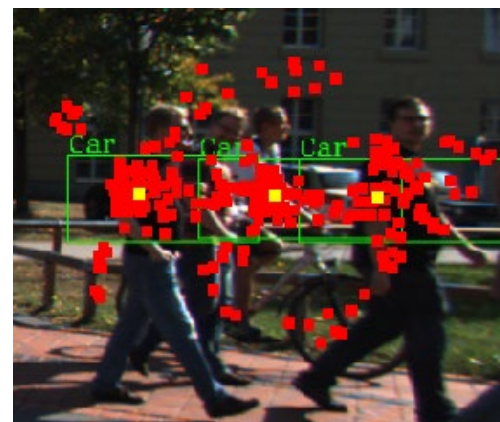
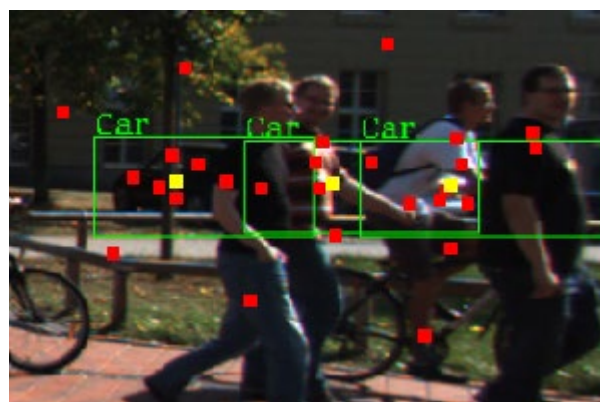
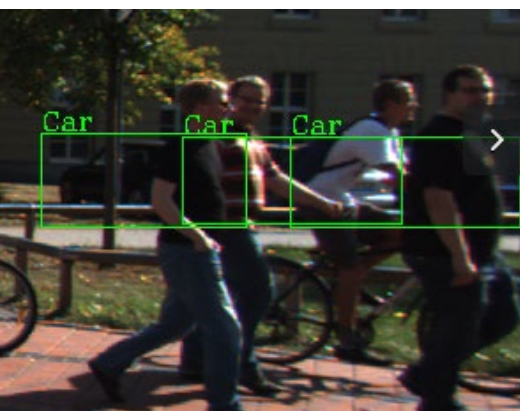
Reppoint offset



support1 offset



support2 offset



For the top case, the N+2 signal is right, then the problem could come from the aggregation, where we use N frame as anchor.

For the bottom case, signals are messed up, both the car and person signal is fused together, the reason might be the resnet and fpn have messed up the signal.

# Person -- not-detected bbox and offset

Green bbox is the ground bbox not-detected.

Yellow point is the center of the bbox

Red points are the offset of the yellow point.

The result is checked frame by frame, if the bbox only shows on frame 3 and frame 5, that means frame 4 detects all objects.

Person class consist of pedestrian and sitting person on kitti

Support1 is N-2 frame, support2 is N+2 frame

Reppoint+stsn is two dcn layers, so we get 81 offsets to show

	Vehicle	Person	Cyclist
Train	22543	8702	1354
test	8058	3444	584

- First show failure case of reppoint but can be detected by stsn
- Then the failure case of stsn

# Stsn get not-detected bbox of reppoint



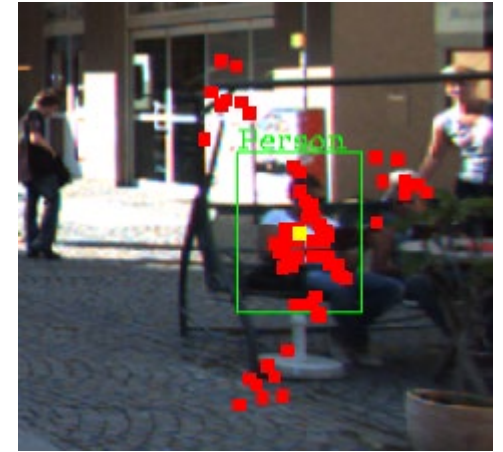
Reppoint not-detected



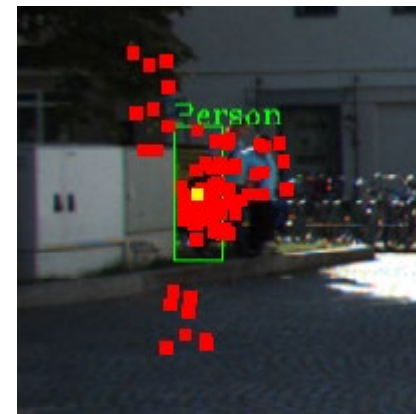
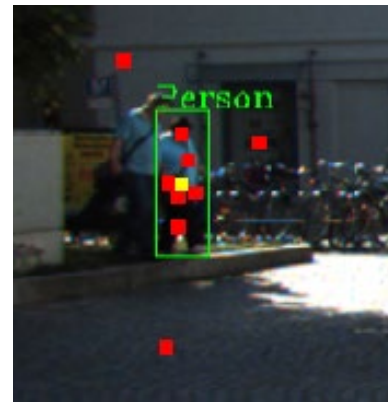
Reppoint offset



support1 offset



support2 offset



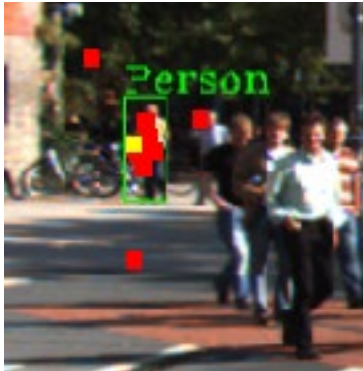
For the top case, N+2 frame brings in the good signal and the N frame is basically right on the person.

For the bottom case, the two persons are similar, the N-2 frame fuse the signal to make it more like a person.

# Stsn get not-detected bbox of reppoint



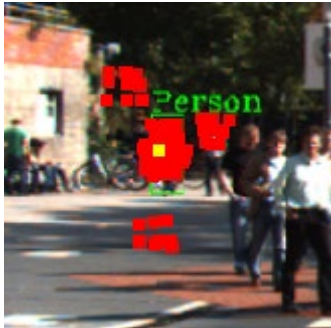
Reppoint not-detected



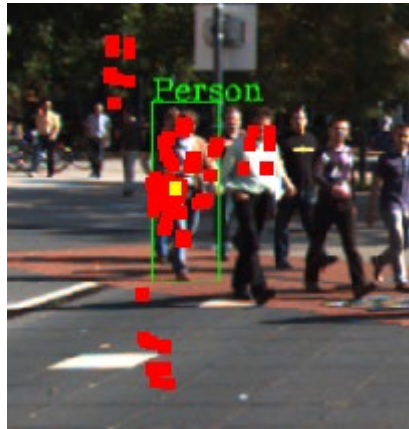
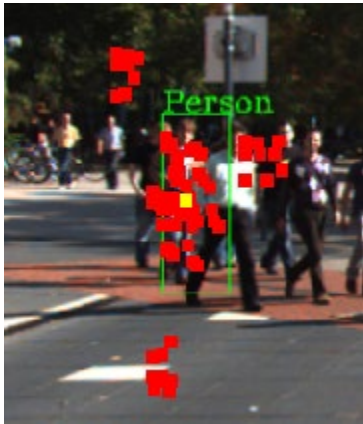
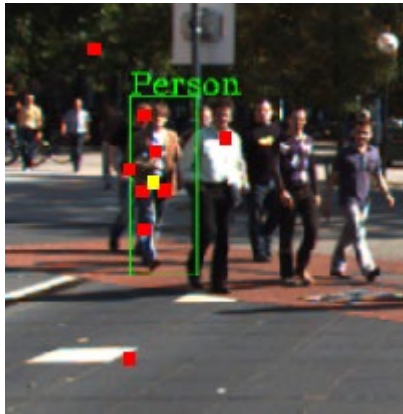
Reppoint offset



support1 offset



support2 offset



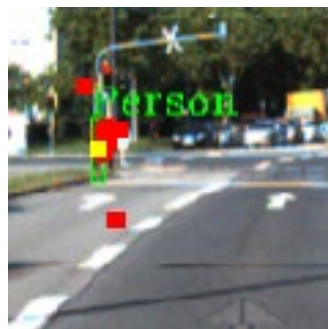
- For the top case, the stsn moves the person's head to the small person and enhance the body signal at the same time.
- For the bottom case, two persons all wear the bright shirt, N+2 frame enhances the signal and the head signal might be extremely important for the classification of person, so finding where the head is the both stsn and reppoint trying to do.

# fail case of stsn

Frame  
64



stsn not-detected



Reppoint offset



support1 offset

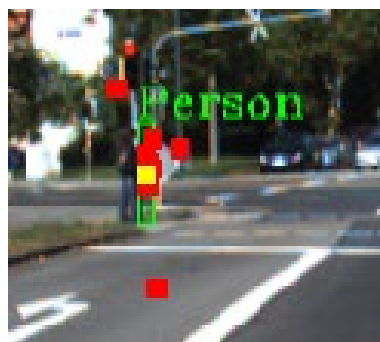


support2 offset

Frame  
82



stsn not-detected



Reppoint offset



support1 offset



support2 offset

It is interesting the offset looks like a shape of person, that's we can say the feature to infer the offset has recognize the person, but the signal and scale is not suitable for classifier.

As the offset also is a signal, we may use it as a feature too, rather than just treat it a motion signal.

# fail case of stsn



stsn not-detected



Reppoint offset



support1 offset



support2 offset



The top case is right might because of head is wrong and we can see the N+2 frame fuses the signal of two person but the similarity of the two person is high, this might confuse the aggregation.

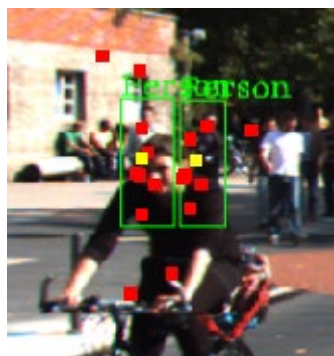
For the bottom case, the two person's signal is messed up. The other reason is that the sitting person only gets 500 samples on the benchmark, and the pose is obviously different from the pedestrian.



# fail case of stsn



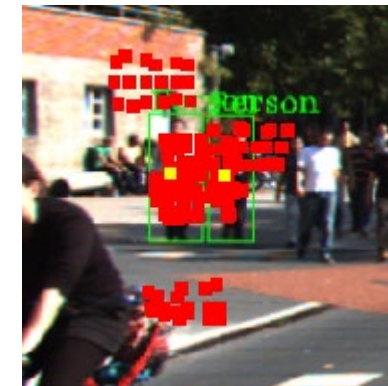
stsn not-detected



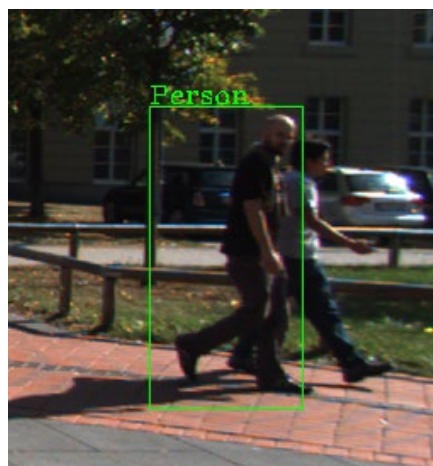
Repoint offset



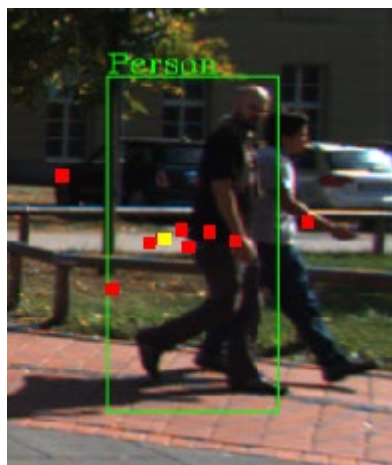
support1 offset



support2 offset



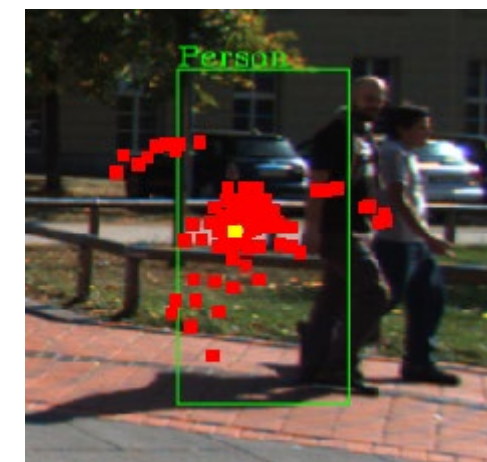
stsn not-detected



Repoint offset



support1 offset



support2 offset

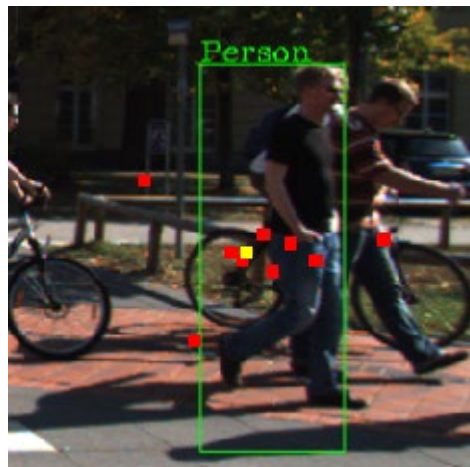
For the top case,  $N+2$  and  $N-2$  bring into right signal, the problem might be the  $N$  frame is not right, so the aggregation is failed because of the wrong anchor.

For the bottom case, I only visualize the offset of stage2, but we can see the offset on frame  $N$  is not right, but on frame  $N-2$  is right, this might require the fusion of signal between different scales.

# fail case of stsn



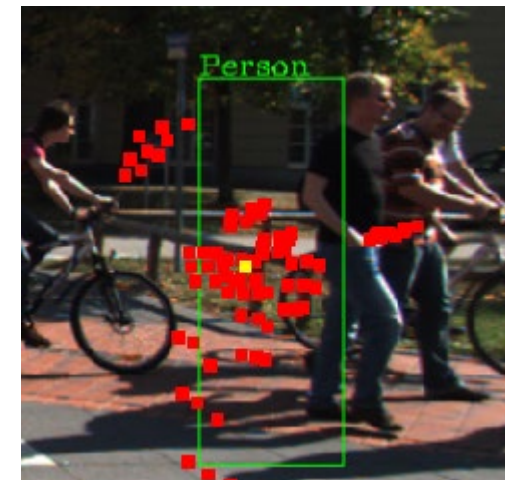
stsn not-detected



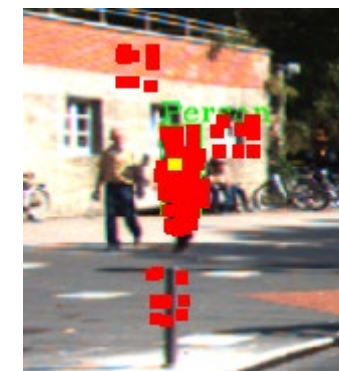
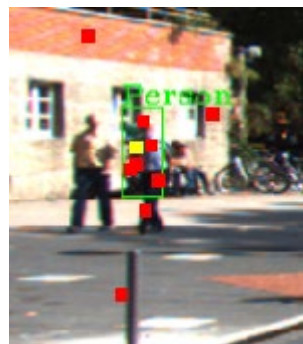
Reppoint offset



support1 offset



support2 offset



- For the top case, the offset on stage 2 is not right, the big object needs to visualize on the later stage. But on the N-2 frame, we can still see it is trying to find the right signal, using stsn may also explore the spatial space, so that we do not need to rely on the later stages.
- For the bottom case, too many signals are fused into the small objects.

# cyclist--not-detected bbox and offset

Green bbox is the ground bbox not-detected.

Yellow point is the center of the bbox

Red points are the offset of the yellow point.

The result is checked frame by frame, if the bbox only shows on frame 3 and frame 5, that means frame 4 detects all objects.

The training data for cyclist is real small, and the cyclist in the far distance is very easy to be classified as a person.

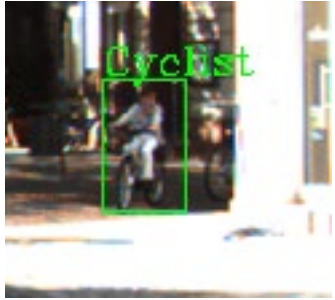
Support1 is N-2 frame, support2 is N+2 frame

Reppoint+stsn is two dcn layers, so we get 81 offsets to show

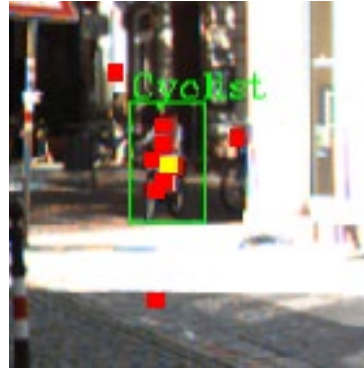
	Vehicle	Person	Cyclist
Train	22543	8702	1354
test	8058	3444	584

- First show failure case of reppoint but can be detected by stsn
- Then the failure case of stsn

# Fail case of stsn



stsn not-detected



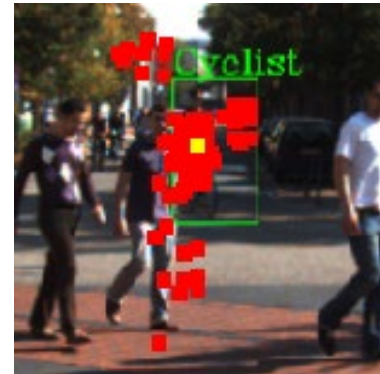
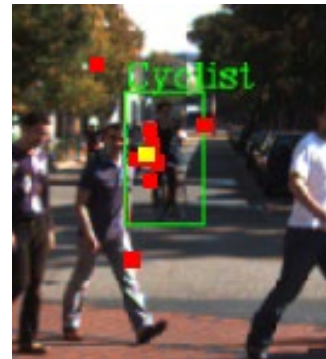
Reppoint offset



support1 offset



support2 offset

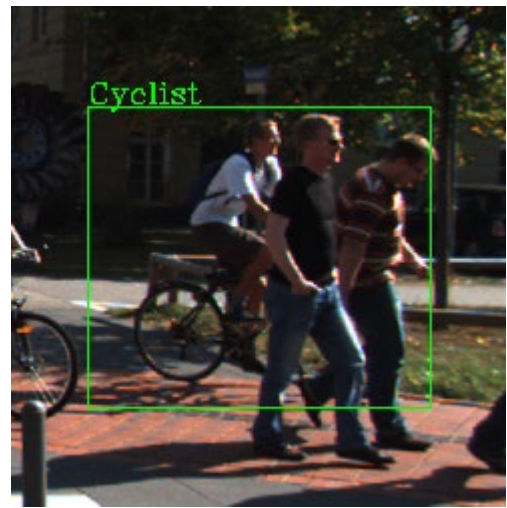


To be clear, most cyclist is classified as person because of the lack of training samples. They are detected but in the wrong class.

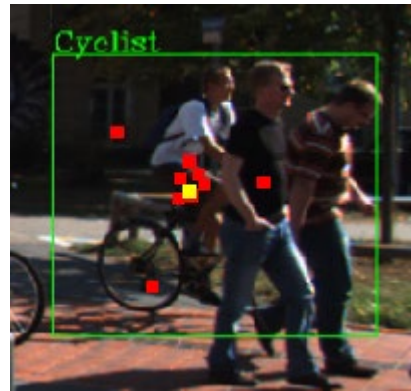
From bottom case, we can see the signal is easy to move on the person.

# fail case of stsn

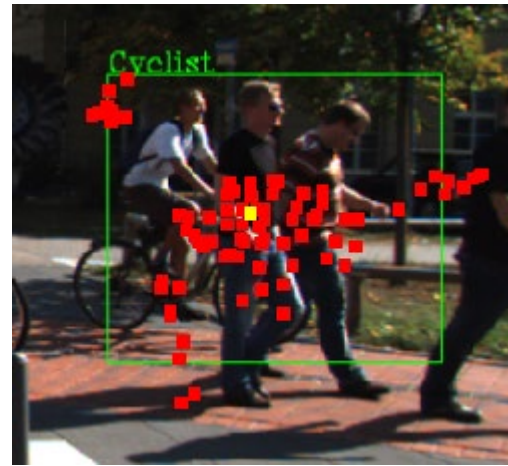
Cyclist occluded by person



stsn not-detected



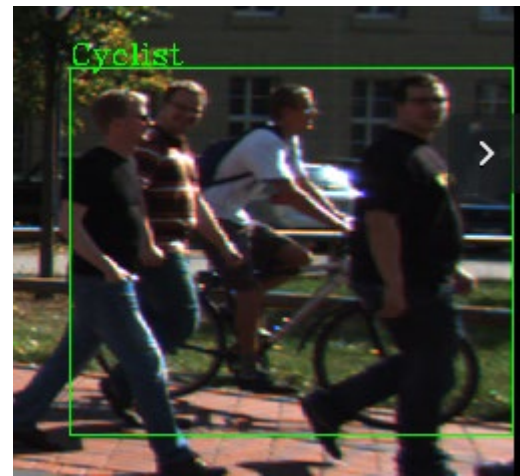
Reppoint offset



support1 offset



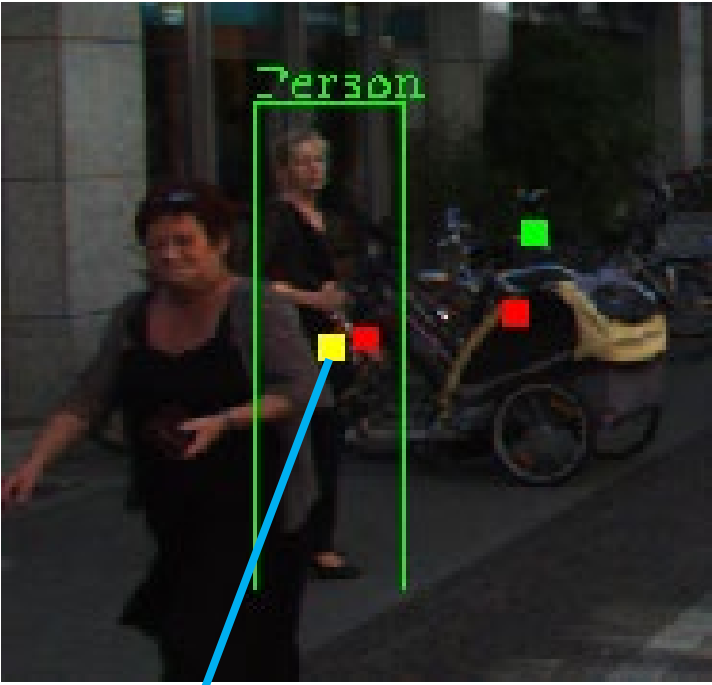
support2 offset



The stsn offset is wrong, it do not search for the person with white shirt but directly searching for person with black shirt. On the bottom base, even the signal of the rider on N-2 frame is more similar to frame N, the offsets are still on the black shirt person and the head is on the other person, the signal is confused for both aggregation and classification.

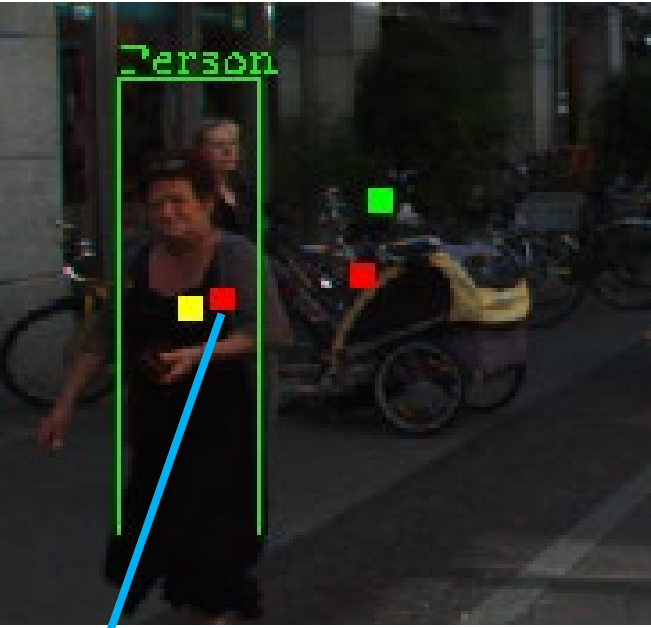
# Init offset Success case

Frame N



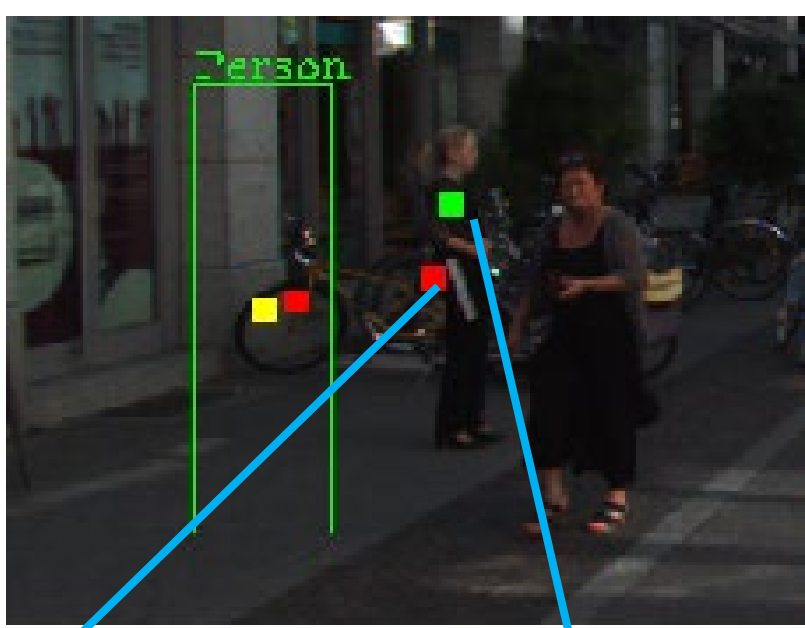
Occluded person at frame N

Frame N-1



Compute Stsn Offset on N-1  
Offset is the step A  
Compute the grad by linearization

Frame N-5



Init offset=5\*A  
• Warp reference feature by bilinear  
• Linearization  
feature=grad\*5A+reference

Final offset  
Computed  
between support  
and warped  
feature

# Quantity result of initialization

Training with random  
selected two frames  
among 8 frames,  
Test with N-1 and N-5

Stsn one  
Score 0.3, nms 0.5

Use the pre-trained stsn  
one model.  
Test with N-1 and N-5,  
Use  $4*(n-1)$  offset as init  
for N-5 frame  
Use bilinear to warp the  
reference to N-5

bilinear  
Score 0.3, nms 0.5

Test with N-1 and N-5,  
Use  $4*(n-1)$  offset as  
step length, use grad  
from n-1 frame to  
compute the reference  
feature at N-5

linear  
Score 0.3, nms 0.5

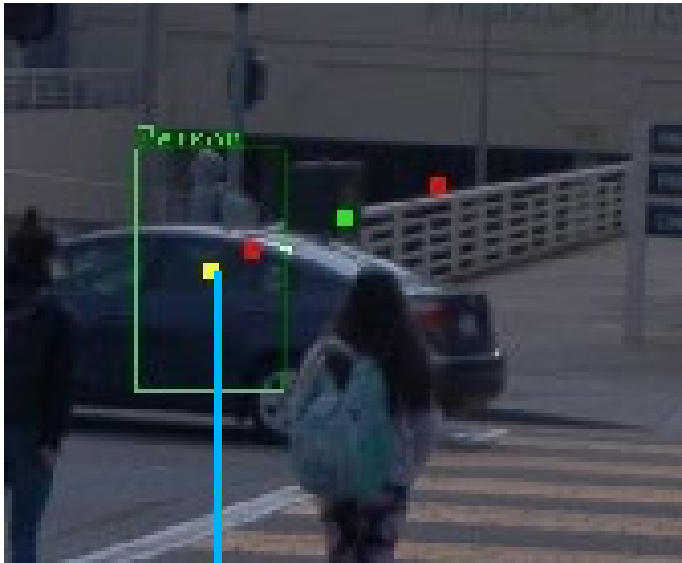
```
evaluating result of refer
+-----+-----+-----+-----+-----+-----+
| class   | gts   | dets  | recall | precision | ap   |
+-----+-----+-----+-----+-----+-----+
| Vehicle | 29289 | 34230 | 0.752  | 0.643     | 0.700 |
| Pedestrian | 19463 | 32328 | 0.765  | 0.460     | 0.682 |
+-----+-----+-----+-----+-----+-----+
| mAP     |       |       |        |           | 0.691 |
+-----+-----+-----+-----+-----+-----+

evaluating result of agg
+-----+-----+-----+-----+-----+-----+
| class   | gts   | dets  | recall | precision | ap   |
+-----+-----+-----+-----+-----+-----+
| Vehicle | 29289 | 45518 | 0.768  | 0.494     | 0.687 |
| Pedestrian | 19463 | 45564 | 0.782  | 0.334     | 0.649 |
+-----+-----+-----+-----+-----+-----+
| mAP     |       |       |        |           | 0.668 |
+-----+-----+-----+-----+-----+-----+

evaluating result of agg
+-----+-----+-----+-----+-----+-----+
| class   | gts   | dets  | recall | precision | ap   |
+-----+-----+-----+-----+-----+-----+
| Vehicle | 29289 | 42730 | 0.756  | 0.518     | 0.657 |
| Pedestrian | 19463 | 38378 | 0.752  | 0.381     | 0.573 |
+-----+-----+-----+-----+-----+-----+
| mAP     |       |       |        |           | 0.615 |
+-----+-----+-----+-----+-----+-----+
```

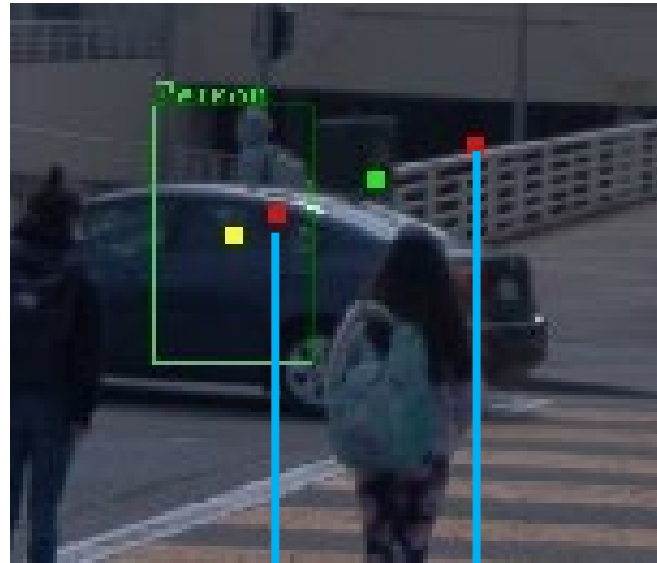
# Success case

Frame N



Occluded person

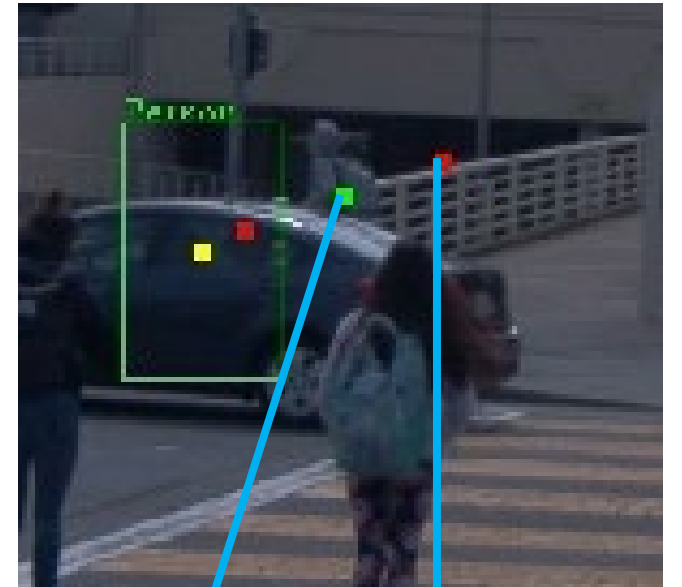
Frame N-1



Offset on N-1

Init offset

Frame N-5



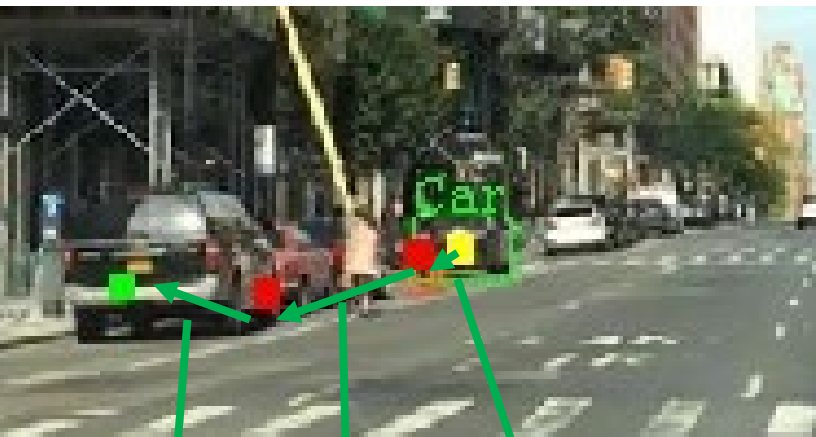
Final offset

Init offset



# Fail case: most offset do not work as flow

Frame N



stsn

Initial  
offset

stsn

Frame N-1



Offset A between frame N and N-1  
Compute grad and step=offset A

Frame N-5

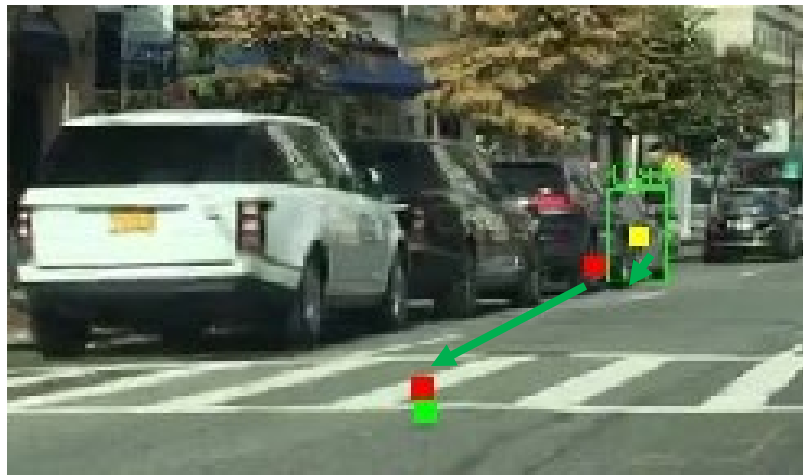


5\*times of A to get the initial  
offset on frame 1

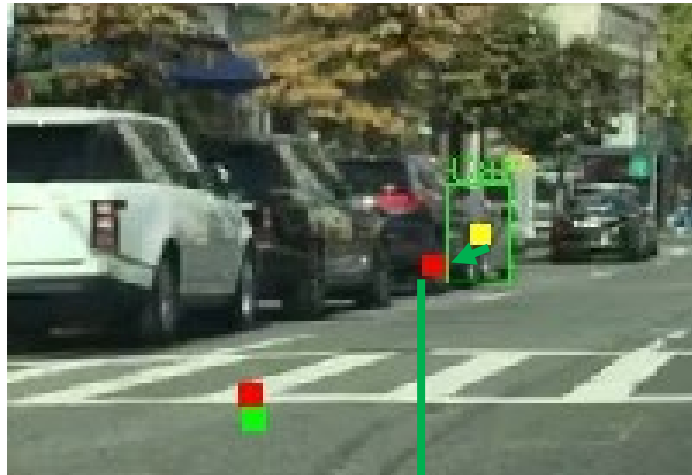
Offset found on frame 1  
warped feature computed from  
linearization by  $\text{grad} * 5 * A + \text{reference}$

The previous exps show the offset do not work as flow as we expected, so use the offset the initialize do not give the result as flow.

Frame N



Frame N-1



The offset is on the same class,  
the car itself is not moving

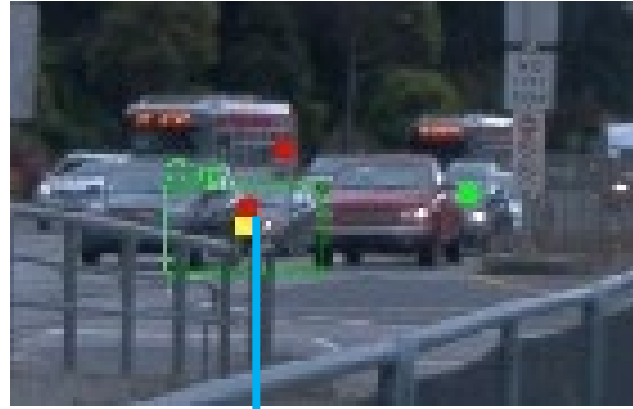
Frame N-5



Initialize the offset by the N-1  
result is failed as the offset is on  
the other object



Occluded car



Offset on N-1 frame



Init offset on N-5 frame

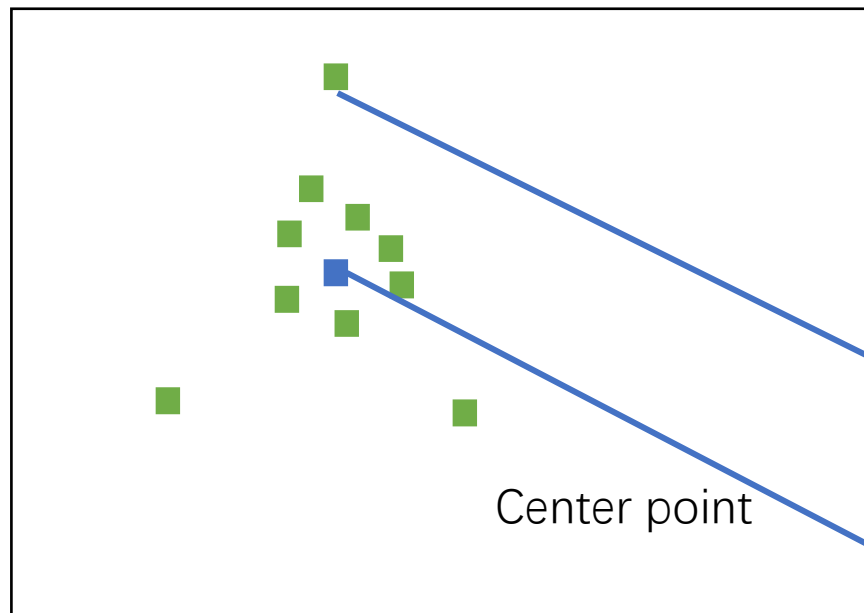
offset on N-5 frame

# RANSAC realization1

Step1: connect the center point by optical flow

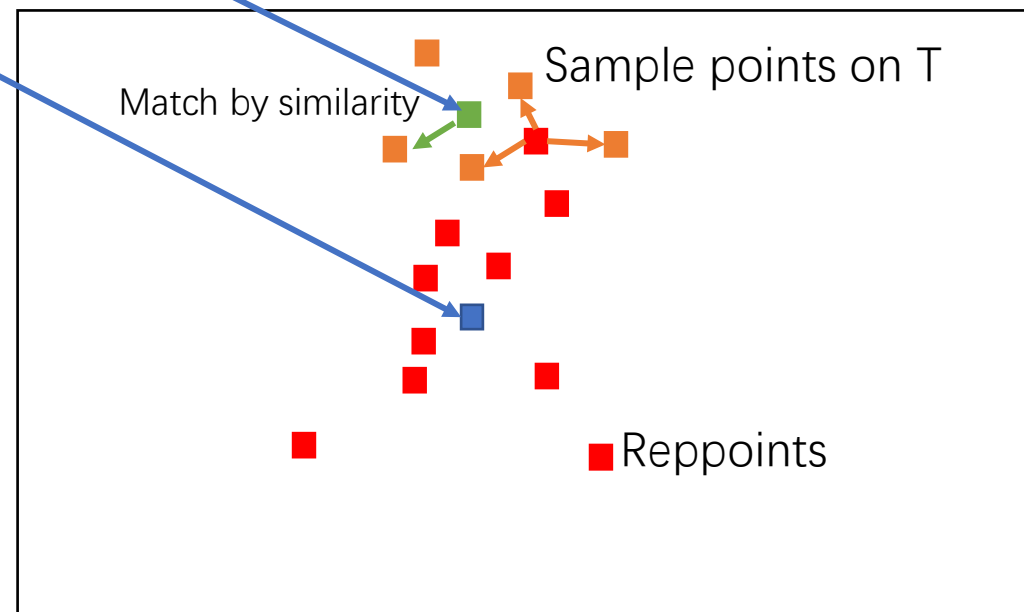
Step2: warp the reppoints at T-1 to T

Step3: sample around the reppoints on T to compute the correspondence between reppoints at T-1 and T



Frame T-1

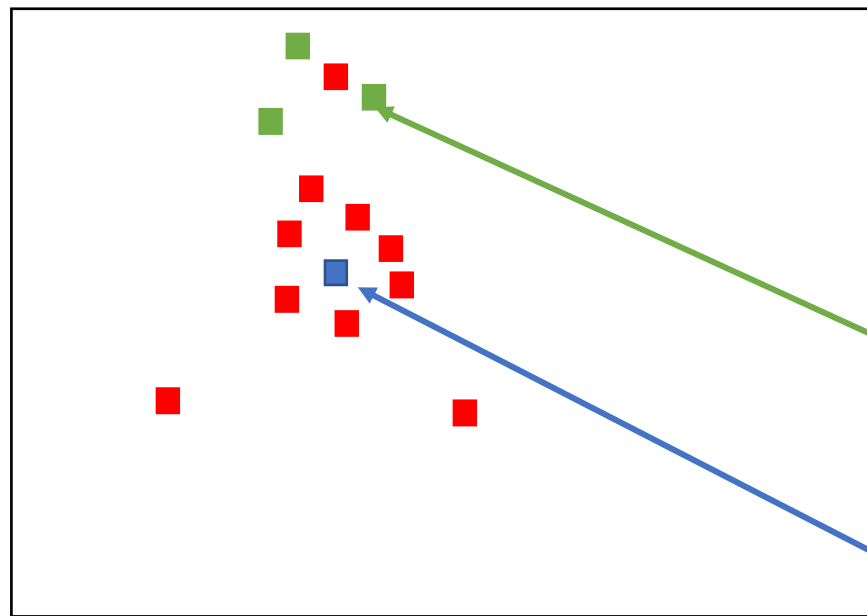
warp by flow



Step4: compute the flow between correspondence, do ransac on the computed correspondence, and find the inner points to keep as the not occluded parts

# RANSAC realization2

Step1: connect the center point by optical flow  
Step2: sample around the reppoints on T-1 and T  
Step3: warp the reppoints at T-1 to T and compute the correspondence of sampled points between N and N-1 by their similarity



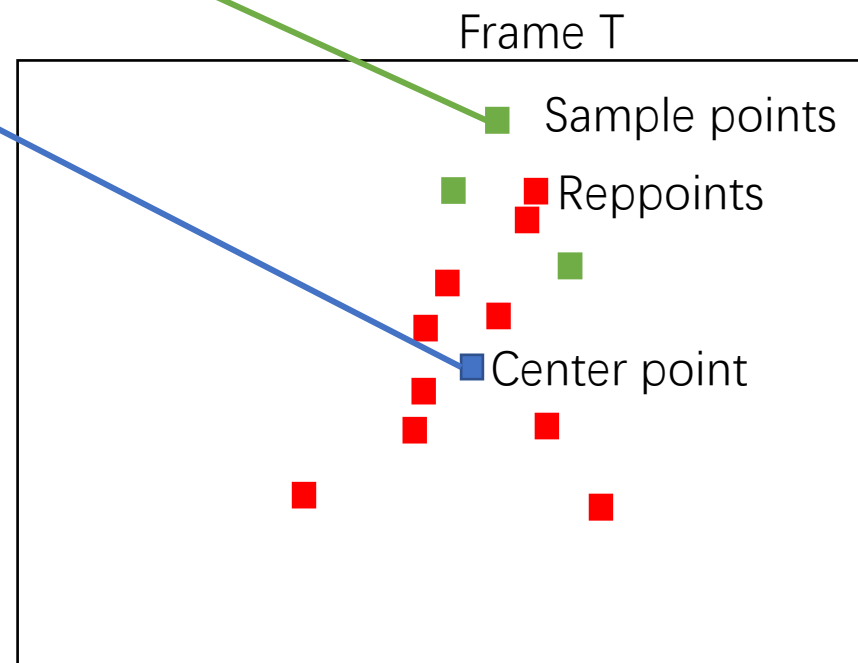
Frame T-1

Connect by flow

match by similarity

Step4: compute the flow between correspondence, do ransac among the sampled set of each reppoint, and keep the inner points as the corresponding set for one reppoint.

Step5: do ransac among the nine reppoint sampled sets and select the inner points and the corresponding reppoints not occluded points.



Frame T

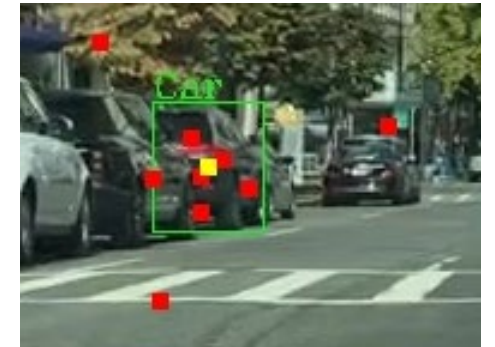
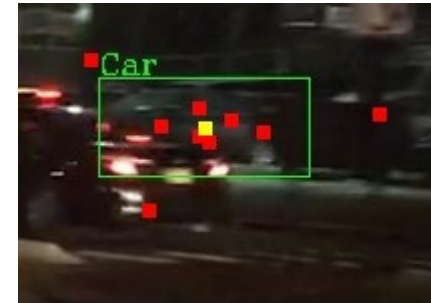
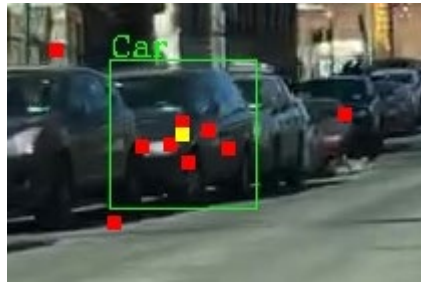
Sample points

Reppoints

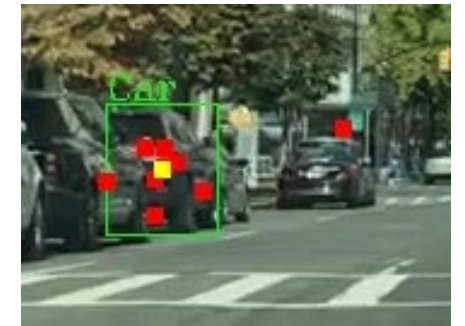
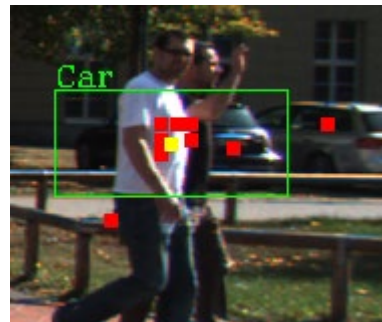
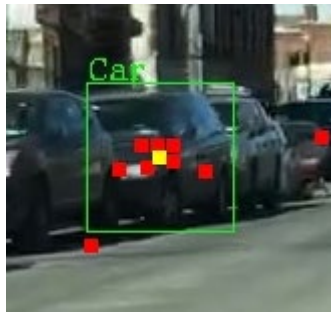
Center point

# ransac

before



after



Points at grid position is deactivate

# Iterate the deactivate and reactivate 10 times

baseline

```
evaluating result of /home/ld/RepPoints/ld_result/stsn_one_r
+-----+-----+-----+-----+-----+
| class   | gts    | dets   | recall | precision | ap    |
+-----+-----+-----+-----+-----+
| Vehicle | 29289  | 36823  | 0.748  | 0.595     | 0.694 |
| Pedestrian | 19463 | 30665  | 0.765  | 0.486     | 0.691 |
+-----+-----+-----+-----+-----+
| mAP     |        |        |        |          | 0.693 |
+-----+-----+-----+-----+-----+
```

Deactivate+  
reactivate

```
evaluating result of /home/ld/RepPoints/ld_result/stsn_one_r
+-----+-----+-----+-----+-----+
| class   | gts    | dets   | recall | precision | ap    |
+-----+-----+-----+-----+-----+
| Vehicle | 29289  | 35504  | 0.746  | 0.615     | 0.696 |
| Pedestrian | 19463 | 28945  | 0.764  | 0.513     | 0.698 |
+-----+-----+-----+-----+-----+
| mAP     |        |        |        |          | 0.697 |
+-----+-----+-----+-----+-----+
```

Iterate de-  
reactivate 5 times

```
single test
evaluating result of /home/ld/RepPoints/ld_result/stsn_one_r
+-----+-----+-----+-----+-----+
| class   | gts    | dets   | recall | precision | ap    |
+-----+-----+-----+-----+-----+
| Vehicle | 29289  | 35597  | 0.745  | 0.613     | 0.695 |
| Pedestrian | 19463 | 29038  | 0.764  | 0.512     | 0.698 |
+-----+-----+-----+-----+-----+
| mAP     |        |        |        |          | 0.696 |
+-----+-----+-----+-----+-----+
```

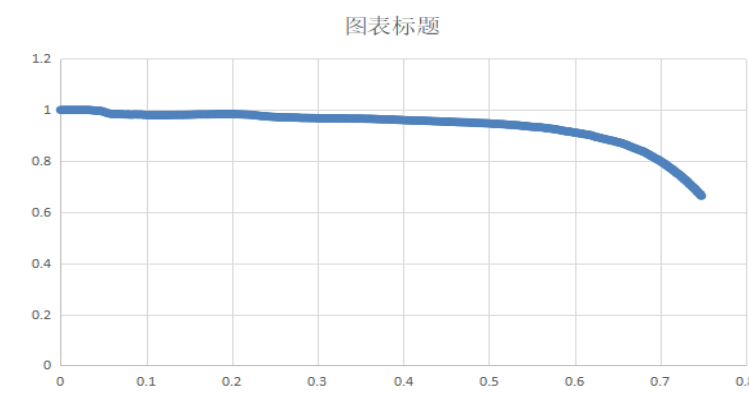
# Stsn one result trained end to end

Stsn one

baseline

```
In [1]: kitti_eval_base(result_record, data_selected)
```

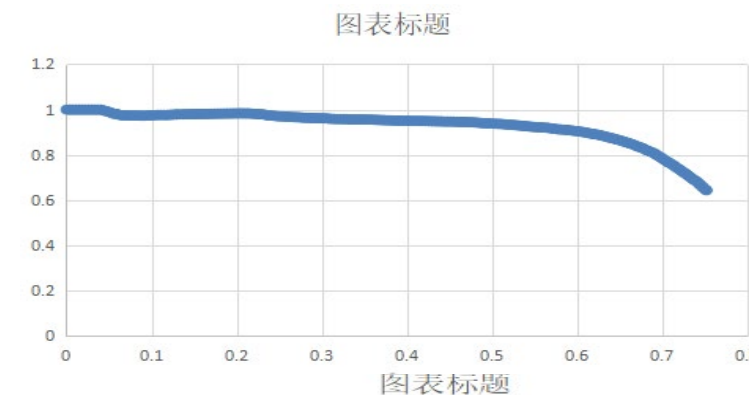
class	gts	dets	recall	precision	ap
Car	29289	32998	0.748	0.664	0.703
Person	19463	29309	0.760	0.505	0.693
mAP					0.698



Fix weight

```
evaluating result of refer
```

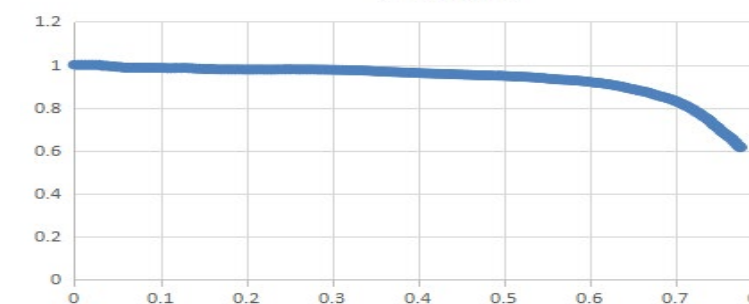
class	gts	dets	recall	precision	ap
Vehicle	29289	34230	0.752	0.643	0.700
Pedestrian	19463	32328	0.765	0.460	0.682
mAP					0.691



End to end

```
evaluating result of refer
```

class	gts	dets	recall	precision	ap
Vehicle	29289	36909	0.775	0.615	0.725
Pedestrian	19463	30404	0.754	0.483	0.679
mAP					0.702





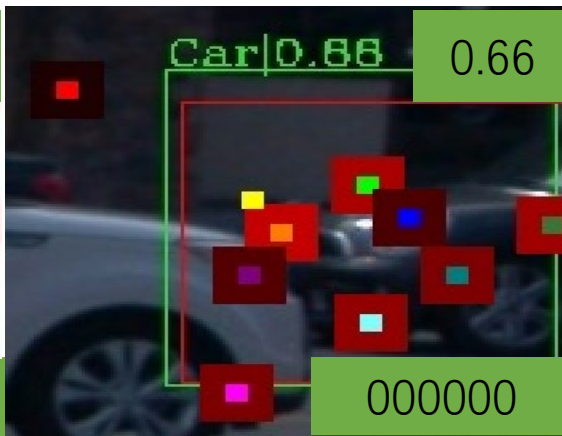
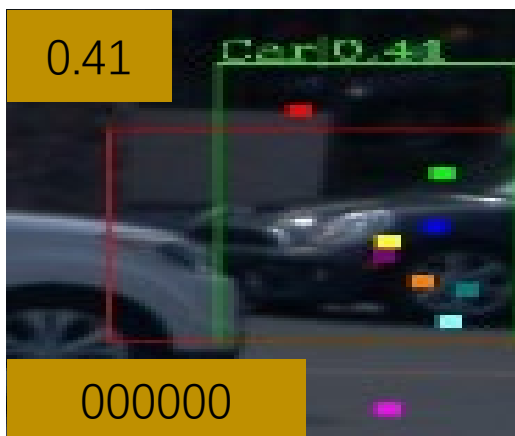
# Result on Waymo

Reppoints Baseline

class	gts	dets	recall	precision	ap
Car	20825	22403	0.700	0.651	0.670
Person	7652	5539	0.492	0.680	0.450
mAP					0.560

Reppoints with Mask

class	gts	dets	recall	precision	ap
Car	20825	20004	0.748	0.779	0.728
Person	7652	5797	0.565	0.745	0.527
mAP					0.628

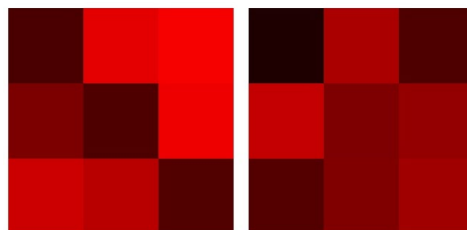
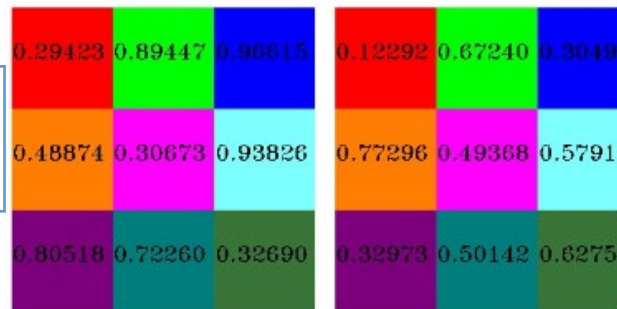


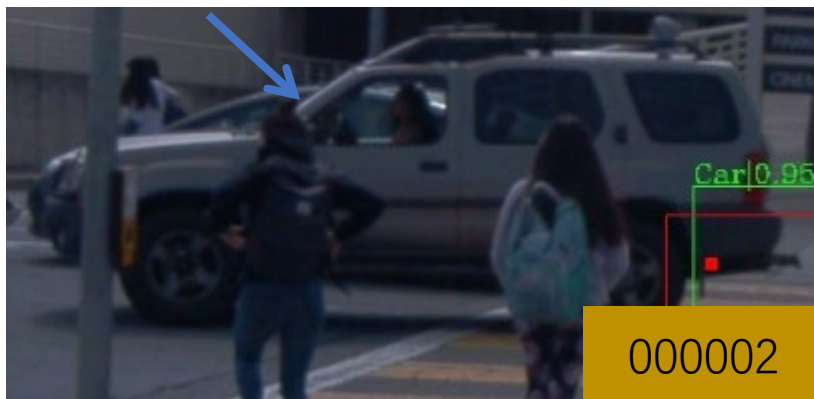
Reppoints  
Baseline

Reppoints  
With Mask

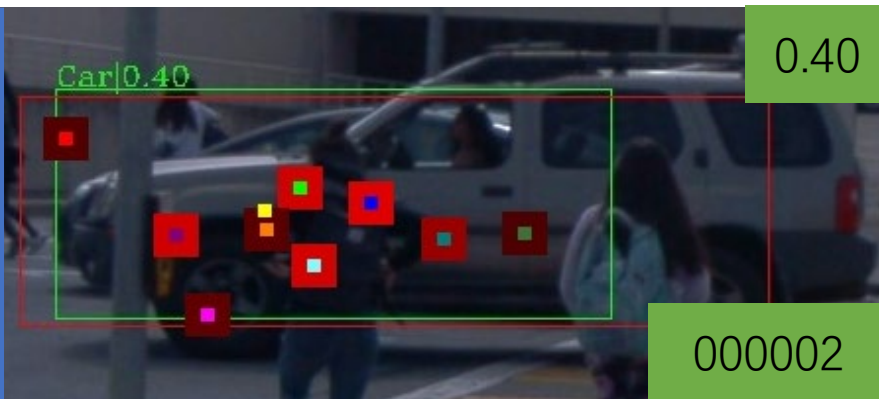
cls\_mask

reg\_mask





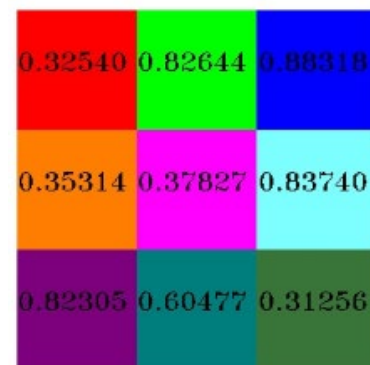
Reppoints  
Baseline



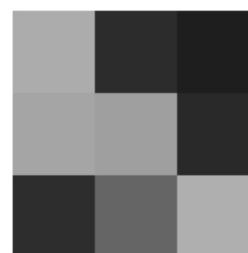
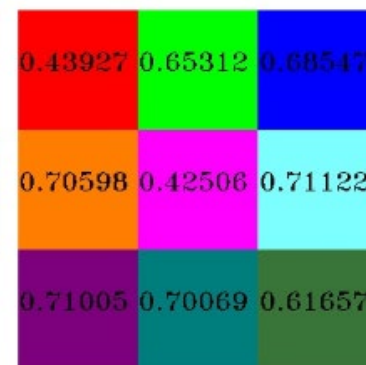
Reppoints  
With Mask



cls\_mask



reg\_mask



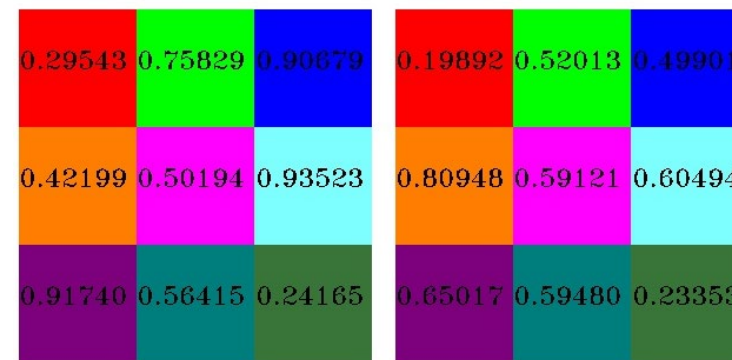


000187	000188	000189
000190	000191	000192
000193	000194	000195
000196	000197	000198

000187	000188	000189
000190	000191	000192
000193	000194	000195
000196	000197	000198

cls\_mask

reg\_mask



Reppoints  
Baseline

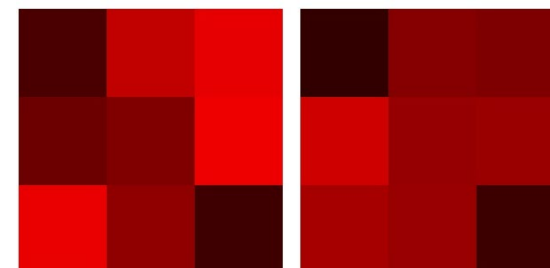
绿色代表能检测出红色车

红色代表能检测不出红色车

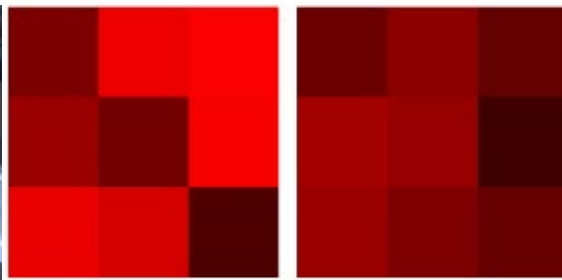
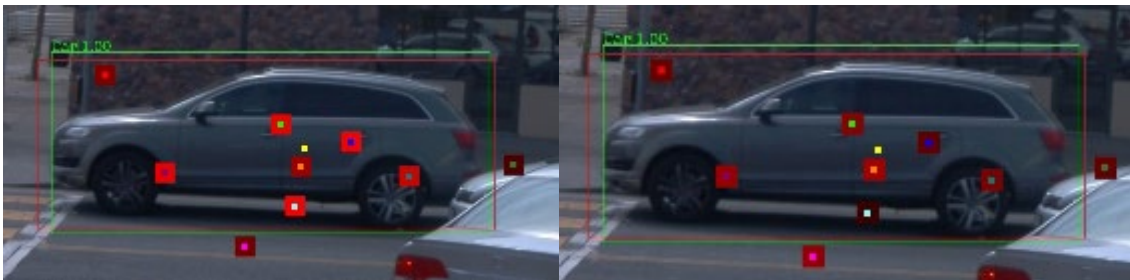
Reppoints  
With Mask

绿色代表能检测出红色车

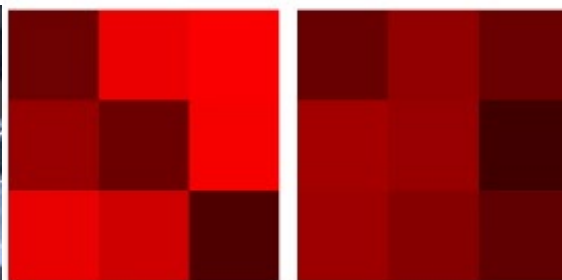
红色代表能检测不出红色车



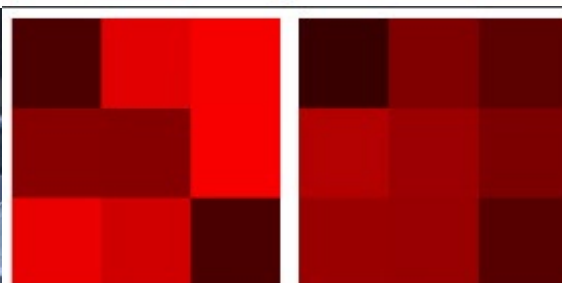
000066



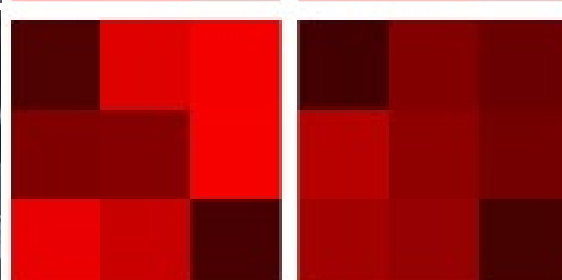
000068



000070



000072



# reppoints baseline VS reppoints with mask

✓ 1	minmax	baseline	<pre>In [1]: kitti_eval_base(result_record, data_selected) +-----+-----+-----+-----+-----+   class   gts    dets    recall   precision   ap    +-----+-----+-----+-----+-----+   Car     20825   22403   0.700    0.651      0.670     Person   7652    5539    0.492    0.680      0.450   +-----+-----+-----+-----+-----+   mAP                                 0.560   +-----+-----+-----+-----+-----+</pre>		<pre>In [1]: kitti_eval(result_record, dataset) +-----+-----+-----+-----+-----+   class     gts    dets    recall   precision   ap    +-----+-----+-----+-----+-----+   Vehicle   20825   22126   0.744    0.700      0.717     Pedestrian   7652    5807    0.537    0.708      0.478   +-----+-----+-----+-----+-----+   mAP                                   0.597   +-----+-----+-----+-----+-----+</pre>
✓ 2	minmax	mask	<pre>In [1]: kitti_eval_base(result_record, data_selected) +-----+-----+-----+-----+-----+   class   gts    dets    recall   precision   ap    +-----+-----+-----+-----+-----+   Car     20825   20004   0.748    0.779      0.728     Person   7652    5797    0.565    0.745      0.527   +-----+-----+-----+-----+-----+   mAP                                 0.628   +-----+-----+-----+-----+-----+</pre>		<pre>+-----+-----+-----+-----+-----+   class     gts    dets    recall   precision   ap    +-----+-----+-----+-----+-----+   Vehicle   20825   21921   0.756    0.718      0.726     Pedestrian   7652    7784    0.628    0.618      0.571   +-----+-----+-----+-----+-----+   mAP                                   0.648   +-----+-----+-----+-----+-----+</pre>
✓ 3	minmax	mask	<pre>evaluating result of refer +-----+-----+-----+-----+-----+   class     gts    dets    recall   precision   ap    +-----+-----+-----+-----+-----+   Vehicle   20825   23240   0.758    0.679      0.728     Pedestrian   7652    7852    0.631    0.614      0.574   +-----+-----+-----+-----+-----+   mAP                                   0.651   +-----+-----+-----+-----+-----+</pre>		<pre>evaluating result of refer +-----+-----+-----+-----+-----+   class     gts    dets    recall   precision   ap    +-----+-----+-----+-----+-----+   Vehicle   20825   22529   0.756    0.699      0.729     Pedestrian   7652    8301    0.644    0.594      0.580   +-----+-----+-----+-----+-----+   mAP                                   0.655   +-----+-----+-----+-----+-----+</pre>
✓ 4	moment	baseline			
✓ 5	moment	mask			
✓ 6	moment	mask	prediction		

# reppoints baseline VS reppoints with mask

<input checked="" type="checkbox"/> 1	minmax	baseline	<pre> +-----+-----+-----+-----+-----+   class   gts    dets    recall   precision   ap    +-----+-----+-----+-----+-----+   Vehicle   102251   118899   0.772   0.664   0.741     Pedestrian   63222   39830   0.434   0.690   0.392   +-----+-----+-----+-----+-----+   mAP           0.567   +-----+-----+-----+-----+-----+ </pre>
<input checked="" type="checkbox"/> 2	minmax	mask	<pre> In [1]: kitti_eval(result_record, dataset) +-----+-----+-----+-----+-----+   class   gts    dets    recall   precision   ap    +-----+-----+-----+-----+-----+   Vehicle   102251   103006   0.784   0.778   0.764     Pedestrian   63222   45809   0.537   0.741   0.501   +-----+-----+-----+-----+-----+   mAP           0.632   +-----+-----+-----+-----+-----+ </pre>

<input checked="" type="checkbox"/> 3	moment	baseline	<pre> +-----+-----+-----+-----+-----+   class   gts    dets    recall   precision   ap    +-----+-----+-----+-----+-----+   Vehicle   102251   121211   0.798   0.673   0.769     Pedestrian   63222   46881   0.531   0.716   0.483   +-----+-----+-----+-----+-----+   mAP           0.626   +-----+-----+-----+-----+-----+ </pre>
<input checked="" type="checkbox"/> 4	moment	mask	<pre> In [1]: kitti_eval(result_record, dataset) +-----+-----+-----+-----+-----+   class   gts    dets    recall   precision   ap    +-----+-----+-----+-----+-----+   Vehicle   102251   112999   0.800   0.724   0.773     Pedestrian   63222   60256   0.610   0.640   0.553   +-----+-----+-----+-----+-----+   mAP           0.663   +-----+-----+-----+-----+-----+ </pre>

# reppoints baseline VS reppoints with mask

## Baseline

```
DONE (t=8.33s).
Average Precision (AP) @[ IoU=0.50:0.95 | area= all | maxDets=100 ] = 0.416
Average Precision (AP) @[ IoU=0.50 | area= all | maxDets=100 ] = 0.620
Average Precision (AP) @[ IoU=0.75 | area= all | maxDets=100 ] = 0.453
Average Precision (AP) @[ IoU=0.50:0.95 | area= small | maxDets=100 ] = 0.245
Average Precision (AP) @[ IoU=0.50:0.95 | area=medium | maxDets=100 ] = 0.463
Average Precision (AP) @[ IoU=0.50:0.95 | area= large | maxDets=100 ] = 0.541
Average Recall (AR) @[ IoU=0.50:0.95 | area= all | maxDets= 1 ] = 0.342
Average Recall (AR) @[ IoU=0.50:0.95 | area= all | maxDets= 10 ] = 0.547
Average Recall (AR) @[ IoU=0.50:0.95 | area= all | maxDets=100 ] = 0.582
Average Recall (AR) @[ IoU=0.50:0.95 | area= small | maxDets=100 ] = 0.379
Average Recall (AR) @[ IoU=0.50:0.95 | area=medium | maxDets=100 ] = 0.631
Average Recall (AR) @[ IoU=0.50:0.95 | area= large | maxDets=100 ] = 0.742
(mmlab2)
```

## epoch5

```
Average Precision (AP) @[ IoU=0.50:0.95 | area= all | maxDets=100 ] = 0.374
Average Precision (AP) @[ IoU=0.50 | area= all | maxDets=100 ] = 0.569
Average Precision (AP) @[ IoU=0.75 | area= all | maxDets=100 ] = 0.405
Average Precision (AP) @[ IoU=0.50:0.95 | area= small | maxDets=100 ] = 0.210
Average Precision (AP) @[ IoU=0.50:0.95 | area=medium | maxDets=100 ] = 0.420
Average Precision (AP) @[ IoU=0.50:0.95 | area= large | maxDets=100 ] = 0.497
Average Recall (AR) @[ IoU=0.50:0.95 | area= all | maxDets= 1 ] = 0.323
Average Recall (AR) @[ IoU=0.50:0.95 | area= all | maxDets= 10 ] = 0.516
Average Recall (AR) @[ IoU=0.50:0.95 | area= all | maxDets=100 ] = 0.546
Average Recall (AR) @[ IoU=0.50:0.95 | area= small | maxDets=100 ] = 0.344
Average Recall (AR) @[ IoU=0.50:0.95 | area=medium | maxDets=100 ] = 0.592
Average Recall (AR) @[ IoU=0.50:0.95 | area= large | maxDets=100 ] = 0.708
(mmlab2)
```

## epoch15

```
Average Precision (AP) @[ IoU=0.50:0.95 | area= all | maxDets=100 ] = 0.370
Average Precision (AP) @[ IoU=0.50 | area= all | maxDets=100 ] = 0.561
Average Precision (AP) @[ IoU=0.75 | area= all | maxDets=100 ] = 0.399
Average Precision (AP) @[ IoU=0.50:0.95 | area= small | maxDets=100 ] = 0.214
Average Precision (AP) @[ IoU=0.50:0.95 | area=medium | maxDets=100 ] = 0.410
Average Precision (AP) @[ IoU=0.50:0.95 | area= large | maxDets=100 ] = 0.488
Average Recall (AR) @[ IoU=0.50:0.95 | area= all | maxDets= 1 ] = 0.318
Average Recall (AR) @[ IoU=0.50:0.95 | area= all | maxDets= 10 ] = 0.511
Average Recall (AR) @[ IoU=0.50:0.95 | area= all | maxDets=100 ] = 0.543
Average Recall (AR) @[ IoU=0.50:0.95 | area= small | maxDets=100 ] = 0.339
Average Recall (AR) @[ IoU=0.50:0.95 | area=medium | maxDets=100 ] = 0.587
Average Recall (AR) @[ IoU=0.50:0.95 | area= large | maxDets=100 ] = 0.695
```



Reppoints minmax

```
In [1]: kitti_eval_base(result_record, data_selected)
```

class	gts	dets	recall	precision	ap
Car	20825	22403	0.700	0.651	0.670
Person	7652	5539	0.492	0.680	0.450
mAP					0.560

Reppoints minmax Mask

```
In [1]: kitti_eval_base(result_record, data_selected)
```

class	gts	dets	recall	precision	ap
Car	20825	20004	0.748	0.779	0.728
Person	7652	5797	0.565	0.745	0.527
mAP					0.628

moment

```
In [1]: kitti_eval(result_record, dataset)
```

class	gts	dets	recall	precision	ap
Vehicle	20825	22126	0.744	0.700	0.717
Pedestrian	7652	5807	0.537	0.708	0.478
mAP					0.597

epoch 20

Moment+mask

```
In [1]: kitti_eval(result_record, dataset)
```

class	gts	dets	recall	precision	ap
Vehicle	20825	19813	0.734	0.771	0.708
Pedestrian	7652	8693	0.624	0.549	0.559
mAP					0.633

epoch 9

```
evaluating result of refer
```

class	gts	dets	recall	precision	ap
Vehicle	20825	23240	0.758	0.679	0.728
Pedestrian	7652	7852	0.631	0.614	0.574
mAP					0.651

```
evaluating result of refer
```

class	gts	dets	recall	precision	ap
Vehicle	20825	22529	0.756	0.699	0.729
Pedestrian	7652	8301	0.644	0.594	0.580
mAP					0.655

epoch 20

waymo8

minmax+mask+prediction

Moment+mask+prediction

Reppoints minmax

```
In [1]: kitti_eval_base(result_record, data_selected)
```

class	gts	dets	recall	precision	ap
Car	20825	22403	0.700	0.651	0.670
Person	7652	5539	0.492	0.680	0.450
mAP					0.560

Reppoints minmax  
Mask

```
In [1]: kitti_eval_base(result_record, data_selected)
```

class	gts	dets	recall	precision	ap
Car	20825	20004	0.748	0.779	0.728
Person	7652	5797	0.565	0.745	0.527
mAP					0.628

```
In [1]: kitti_eval(result_record, dataset)
```

class	gts	dets	recall	precision	ap
Vehicle	20825	21921	0.756	0.718	0.726
Pedestrian	7652	7784	0.628	0.618	0.571
mAP					0.648

moment

```
In [1]: kitti_eval(result_record, dataset)
```

class	gts	dets	recall	precision	ap
Vehicle	20825	22126	0.744	0.700	0.717
Pedestrian	7652	5807	0.537	0.708	0.478
mAP					0.597

epoch  
20

Moment+mask

```
In [1]: kitti_eval(result_record, dataset)
```

class	gts	dets	recall	precision	ap
Vehicle	20825	19813	0.734	0.771	0.708
Pedestrian	7652	8693	0.624	0.549	0.559
mAP					0.633

epoch  
9

```
evaluating result of refer
```

class	gts	dets	recall	precision	ap
Vehicle	20825	22529	0.756	0.699	0.729
Pedestrian	7652	8301	0.644	0.594	0.580
mAP					0.655

epoch  
20

Moment+mask+prediction

Moment+mask

class	gts	dets	recall	precision	ap
Vehicle	20825	21921	0.756	0.718	0.726
Pedestrian	7652	7784	0.628	0.618	0.571
mAP					0.648

epoch  
20

# waymo54

<input checked="" type="checkbox"/> 1	minmax	baseline	<pre>+-----+-----+-----+-----+-----+   class     gts      dets     recall   precision   ap      +-----+-----+-----+-----+-----+   Vehicle   102251   118899   0.772    0.664       0.741     Pedestrian   63222   39830    0.434    0.690       0.392   +-----+-----+-----+-----+-----+   mAP  0.567   +-----+-----+-----+-----+-----+</pre>
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<input checked="" type="checkbox"/> 2	minmax	mask	<pre>In [1]: kitti_eval(result_record, dataset) +-----+-----+-----+-----+-----+   class     gts      dets     recall   precision   ap      +-----+-----+-----+-----+-----+   Vehicle   102251   103006   0.784    0.778       0.764     Pedestrian   63222   45809    0.537    0.741       0.501   +-----+-----+-----+-----+-----+   mAP  0.632   +-----+-----+-----+-----+-----+</pre>
---------------------------------------	--------	------	---

<input checked="" type="checkbox"/> 3	moment	baseline	<pre>+-----+-----+-----+-----+-----+   class     gts      dets     recall   precision   ap      +-----+-----+-----+-----+-----+   Vehicle   102251   121211   0.798    0.673       0.769     Pedestrian   63222   46881    0.531    0.716       0.483   +-----+-----+-----+-----+-----+   mAP  0.626   +-----+-----+-----+-----+-----+</pre>
---------------------------------------	--------	----------	--

<input checked="" type="checkbox"/> 4	moment	mask	<pre>In [1]: kitti_eval(result_record, dataset) +-----+-----+-----+-----+-----+   class     gts      dets     recall   precision   ap      +-----+-----+-----+-----+-----+   Vehicle   102251   112999   0.800    0.724       0.773     Pedestrian   63222   60256    0.610    0.640       0.553   +-----+-----+-----+-----+-----+   mAP  0.663   +-----+-----+-----+-----+-----+</pre>
---------------------------------------	--------	------	---

```
evaluating result of refer
+-----+-----+-----+-----+-----+
| class   | gts    | dets   | recall | precision | ap    |
+-----+-----+-----+-----+-----+
| Vehicle | 102251 | 119940 | 0.813  | 0.693     | 0.787 |
| Pedestrian | 63222 | 61177  | 0.583  | 0.603     | 0.520 |
+-----+-----+-----+-----+-----+
| mAP     |        |        |        |           | 0.654 |
+-----+-----+-----+-----+-----+
```

Prediction minmax

```
evaluating result of refer
+-----+-----+-----+-----+-----+
| class   | gts    | dets   | recall | precision | ap    |
+-----+-----+-----+-----+-----+
| Vehicle | 102251 | 117039 | 0.801  | 0.700     | 0.776 |
| Pedestrian | 63222 | 63292  | 0.610  | 0.609     | 0.552 |
+-----+-----+-----+-----+-----+
| mAP     |        |        |        |           | 0.664 |
+-----+-----+-----+-----+-----+
```

Prediction moment

# Stsn waymo8

Reppoints minmax

```
In [1]: kitti_eval_base(result_record, data_selected)
+-----+-----+-----+-----+-----+
| class | gts  | dets  | recall | precision | ap   |
+-----+-----+-----+-----+-----+
| Car    | 20825 | 22403 | 0.700  | 0.651    | 0.670 |
| Person | 7652  | 5539  | 0.492  | 0.680    | 0.450 |
+-----+-----+-----+-----+-----+
| mAP    |      |      |      |      | 0.560 |
+-----+-----+-----+-----+-----+
```

Reppoints minmax  
Mask

```
In [1]: kitti_eval_base(result_record, data_selected)
+-----+-----+-----+-----+-----+
| class | gts  | dets  | recall | precision | ap   |
+-----+-----+-----+-----+-----+
| Car    | 20825 | 20004 | 0.748  | 0.779    | 0.728 |
| Person | 7652  | 5797  | 0.565  | 0.745    | 0.527 |
+-----+-----+-----+-----+-----+
| mAP    |      |      |      |      | 0.628 |
+-----+-----+-----+-----+-----+
```

evaluating result of refer

```
+-----+-----+-----+-----+-----+
| class | gts  | dets  | recall | precision | ap   |
+-----+-----+-----+-----+-----+
| Vehicle | 20825 | 20625 | 0.725  | 0.732    | 0.702 |
| Pedestrian | 7652  | 7923  | 0.637  | 0.615    | 0.568 |
+-----+-----+-----+-----+-----+
| mAP    |      |      |      |      | 0.635 |
+-----+-----+-----+-----+-----+
/home/ld/RepPoints/ld_result/stsn_minmax_waymo/epoch_97_thres0.3_nms0.5_-1
(mmlab) ld@cci-deployment-20202211-5957d5bc76-nlcpu:/RepPoints$
```

evaluating result of refer

```
+-----+-----+-----+-----+-----+
| class | gts  | dets  | recall | precision | ap   |
+-----+-----+-----+-----+-----+
| Vehicle | 20825 | 20629 | 0.725  | 0.732    | 0.702 |
| Pedestrian | 7652  | 7868  | 0.635  | 0.618    | 0.566 |
+-----+-----+-----+-----+-----+
| mAP    |      |      |      |      | 0.634 |
+-----+-----+-----+-----+-----+
/home/ld/RepPoints/ld_result/stsn_minmax_waymo/epoch_97_thres0.3_nms0.5_-5
(mmlab) ld@cci-deployment-20202211-5957d5bc76-nlcpu:/RepPoints$
```

evaluating result of refer

```
+-----+-----+-----+-----+-----+
| class | gts  | dets  | recall | precision | ap   |
+-----+-----+-----+-----+-----+
| Vehicle | 20825 | 20647 | 0.725  | 0.731    | 0.702 |
| Pedestrian | 7652  | 7854  | 0.636  | 0.620    | 0.567 |
+-----+-----+-----+-----+-----+
| mAP    |      |      |      |      | 0.635 |
+-----+-----+-----+-----+-----+
/home/ld/RepPoints/ld_result/stsn_minmax_waymo/epoch_97_thres0.3_nms0.5_-10
(mmlab) ld@cci-deployment-20202211-5957d5bc76-nlcpu:/RepPoints$
```