

# Large-scale Traffic Prediction using 3DResNet and Sparse-UNet

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Presenter: Bo Wang

10<sup>th</sup> Dec 2021

# Outline



**1. Background**

**2. Methodology**

**3. Conclusion**

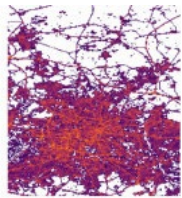


# Background



Inputs **X** (12 Timesteps)

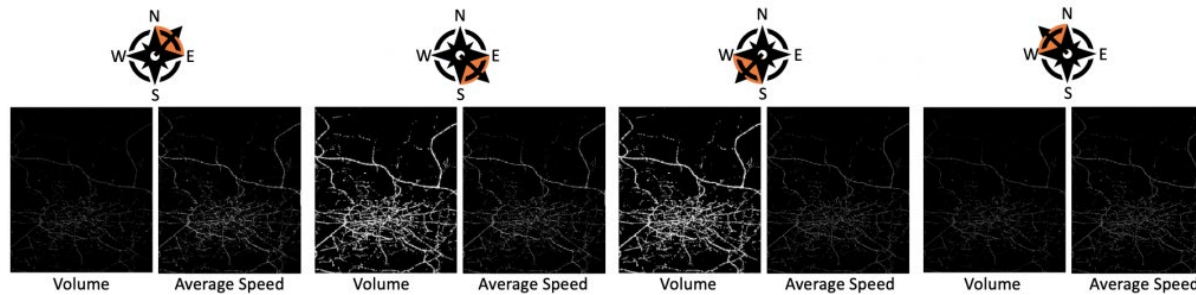
Targets **Y** (6 Timesteps)



$1 \times W \times H \times I$

Timestep

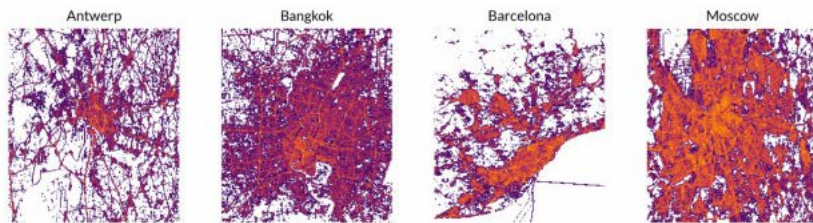
Expand I



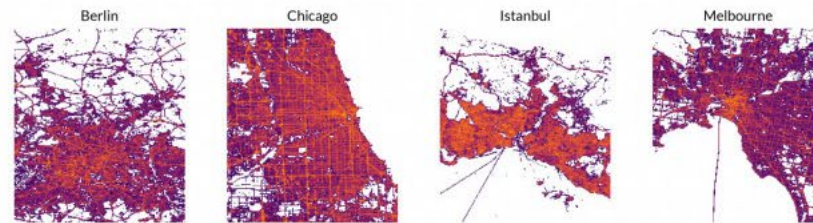
8 Traffic Channels

# Background

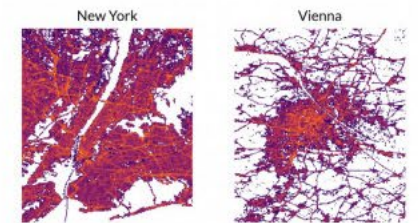
## TRAINING



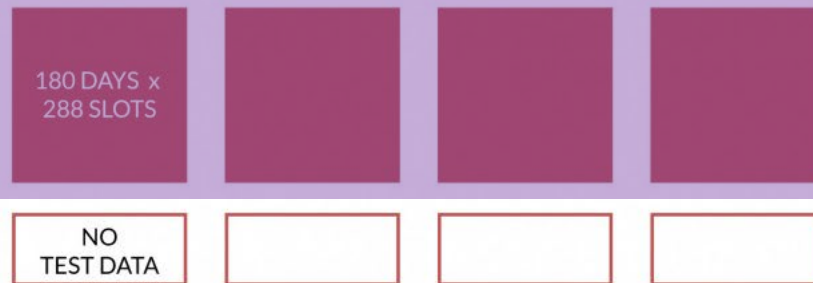
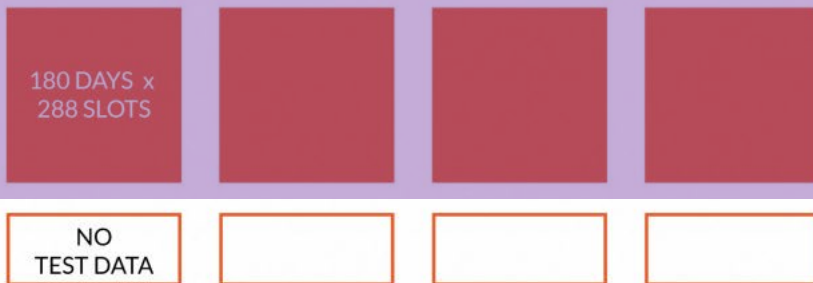
## CORE CHALLENGE



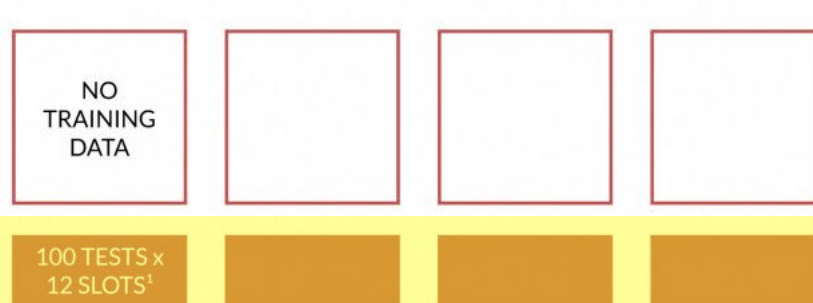
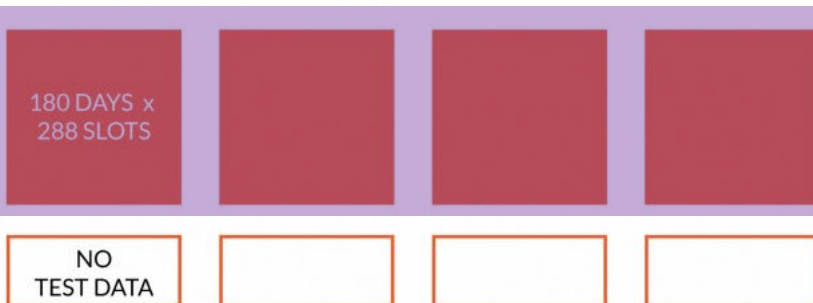
## EXTENDED CHALLENGE



2019



2020



# Methodology – Data Loader

Select days from Monday to Sunday

```
def get_train_file(df, config):  
    diff_dofw = None  
    while diff_dofw != 7:  
        df_temp = pd.concat([df[df. yeartype == k].sample(n=config[k]) for k in df. yeartype. unique()])  
        diff_dofw = len(df_temp. dayofweek. unique())  
    files_train = [Path(f) for f in df_temp. file. values]  
    random. shuffle(files_train)  
    return files_train
```

Select all cities

```
"MOSCOW2019": 1,  
"MOSCOW2020": 1,  
"ANTWERP2020": 1,  
"ANTWERP2019": 1,  
"BERLIN2019": 2,  
"CHICAGO2019": 2,  
"MELBOURNE2019": 3,  
"BARCELONA2019": 1,  
"BARCELONA2020": 1,  
"BANGKOK2020": 2,  
"BANGKOK2019": 2,  
"ISTANBUL2019": 2,
```

```
[PosixPath('/home/bwan131/scratch/nips/data/raw/BANGKOK/training/2020-03-23_BANGKOK_8ch.h5'),  
PosixPath('/home/bwan131/scratch/nips/data/raw/MELBOURNE/training/2019-06-23_MELBOURNE_8ch.h5'),  
PosixPath('/home/bwan131/scratch/nips/data/raw/ANTWERP/training/2019-03-13_ANTWERP_8ch.h5'),  
PosixPath('/home/bwan131/scratch/nips/data/raw/BERLIN/training/2019-02-17_BERLIN_8ch.h5'),  
PosixPath('/home/bwan131/scratch/nips/data/raw/MOSCOW/training/2019-03-02_MOSCOW_8ch.h5'),  
PosixPath('/home/bwan131/scratch/nips/data/raw/BANGKOK/training/2020-05-15_BANGKOK_8ch.h5'),  
PosixPath('/home/bwan131/scratch/nips/data/raw/ANTWERP/training/2020-01-09_ANTWERP_8ch.h5'),  
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PosixPath('/home/bwan131/scratch/nips/data/raw/MELBOURNE/training/2019-05-28_MELBOURNE_8ch.h5'),  
PosixPath('/home/bwan131/scratch/nips/data/raw/BANGKOK/training/2019-01-02_BANGKOK_8ch.h5'),  
PosixPath('/home/bwan131/scratch/nips/data/raw/ISTANBUL/training/2019-04-10_ISTANBUL_8ch.h5'),  
PosixPath('/home/bwan131/scratch/nips/data/raw/CHICAGO/training/2019-01-18_CHICAGO_8ch.h5'),  
PosixPath('/home/bwan131/scratch/nips/data/raw/CHICAGO/training/2019-01-03_CHICAGO_8ch.h5'),  
PosixPath('/home/bwan131/scratch/nips/data/raw/BARCELONA/training/2020-02-11_BARCELONA_8ch.h5')]
```

DataLoader -> input and output batches

X (batch size, 12, 495, 436, 8)  
Y (batch size, 6, 495, 436, 8)

# Methodology – Data Loader

File 1 0, 1, 2, 3, 4

File 2 5, 6, 7, 8, 9

File 3 10, 11, 12, 13, 14

`Dataloader(dataset, batch_size=3, shuffle=True)`

Shuffle all index first  
Get batch index -> (1, 7, 13)

If the indices are coming from multiple files, it will need multiple IO operations.

File 1 12, 13, 11, 14, 15

File 2 2, 3, 1, 4, 0

File 3 9, 6, 5, 7, 10, 8



Custom Sampler + File Cache

From 1.89 batches/s

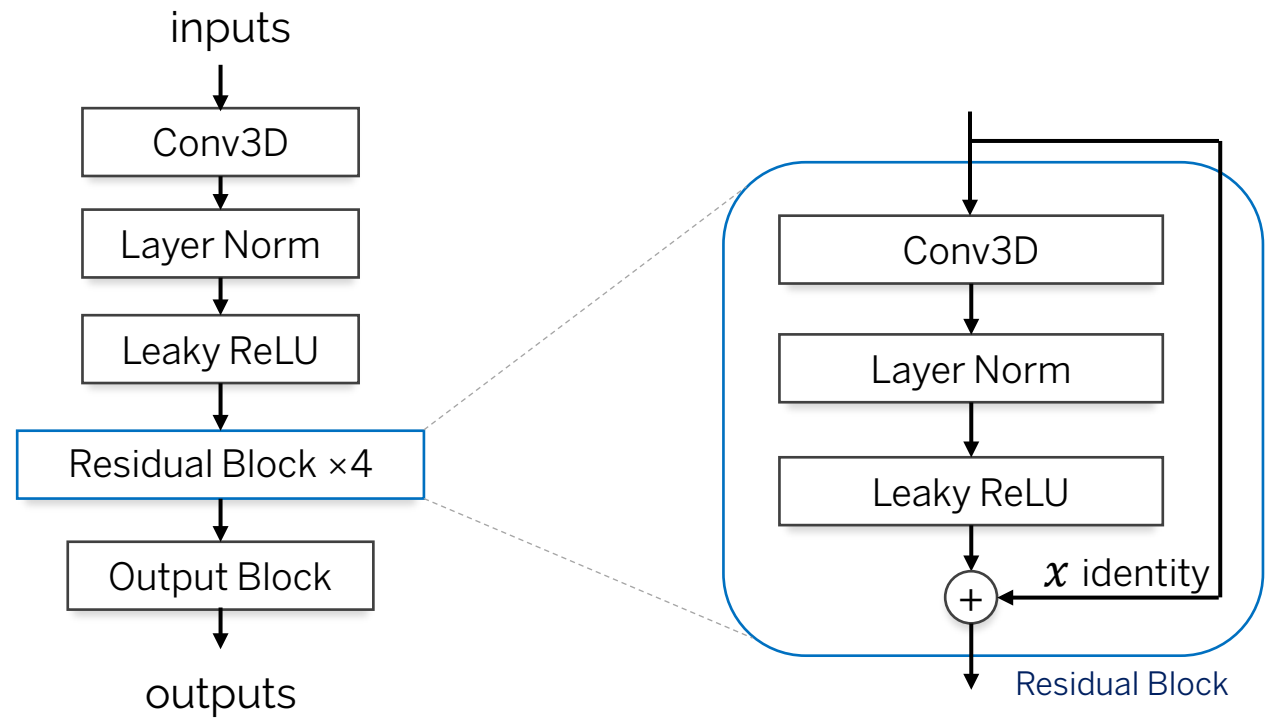
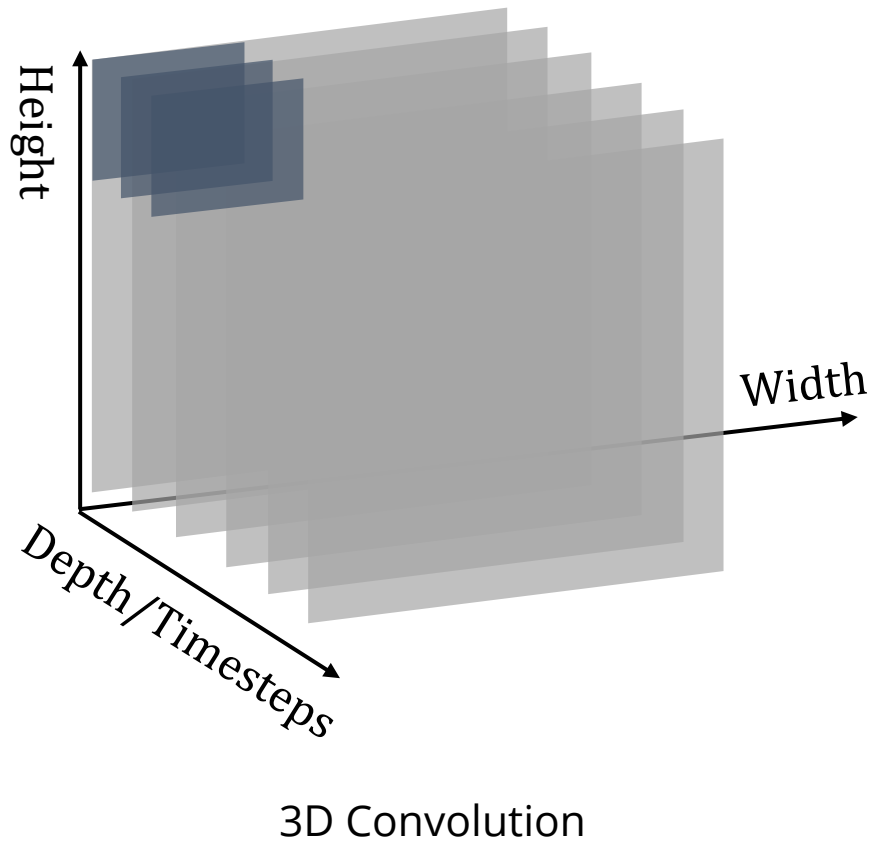
Up to 60 batches/s

30X faster



# Methodology – 3DResNets

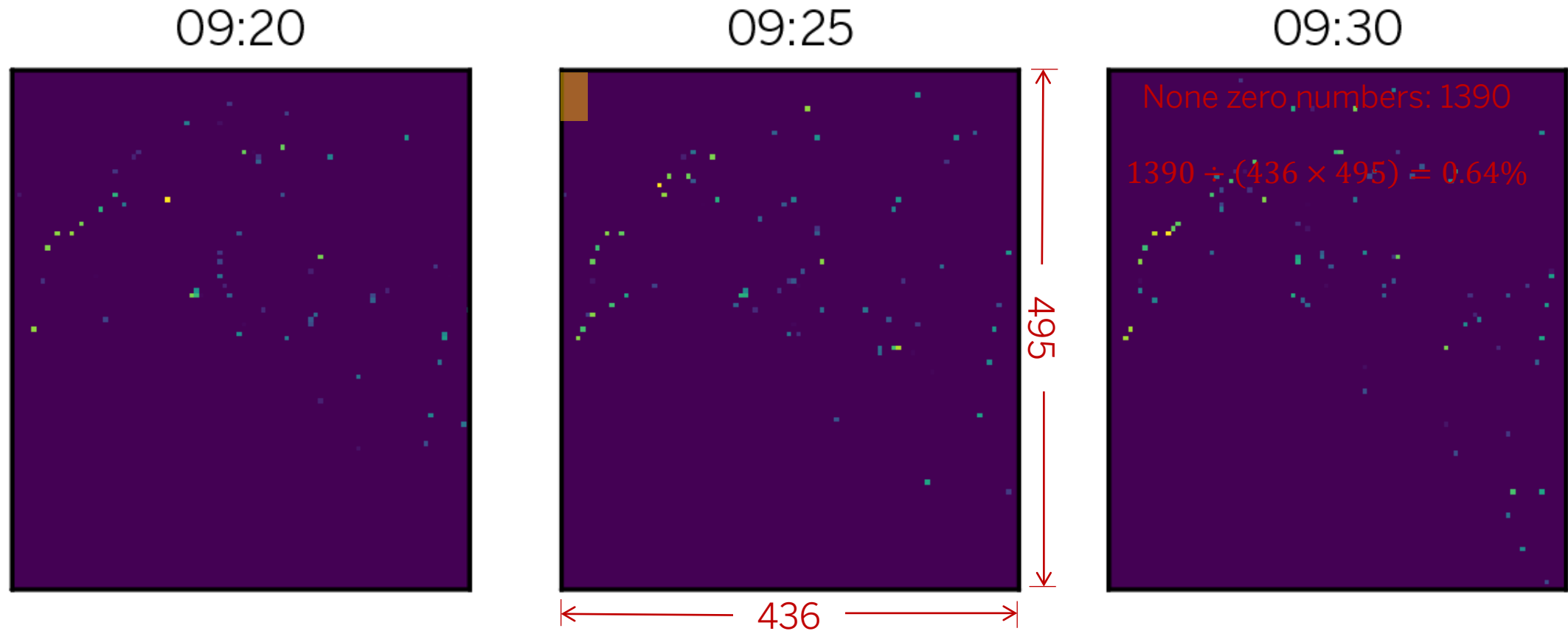
## 3DResNets



MSE 50.219 (5th place) on the Core Challenge

# Methodology – Sparse-UNet

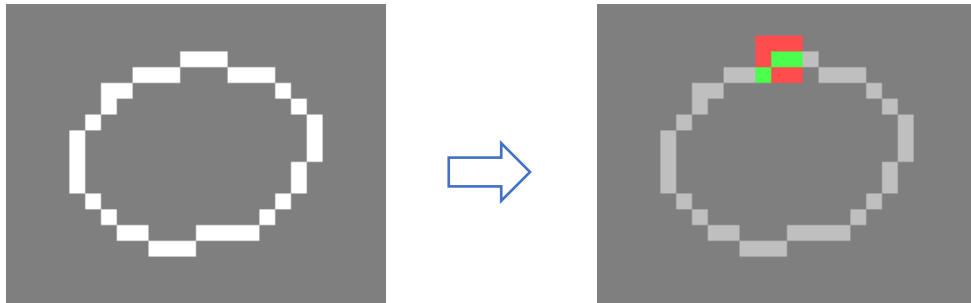
## Motivation





# Methodology – Sparse-UNet

## Sparse Convolution



Regular Convolution

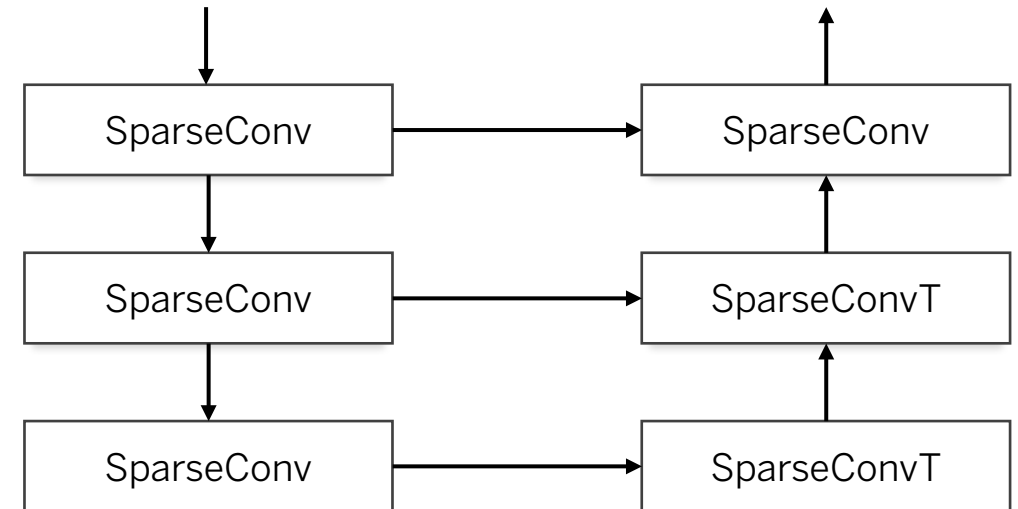
Sparse Convolution

Source:

<https://github.com/facebookresearch/SparseConvNet>

<https://github.com/NVIDIA/MinkowskiEngine>

## Sparse-UNet



# Methodology – Sparse-UNet

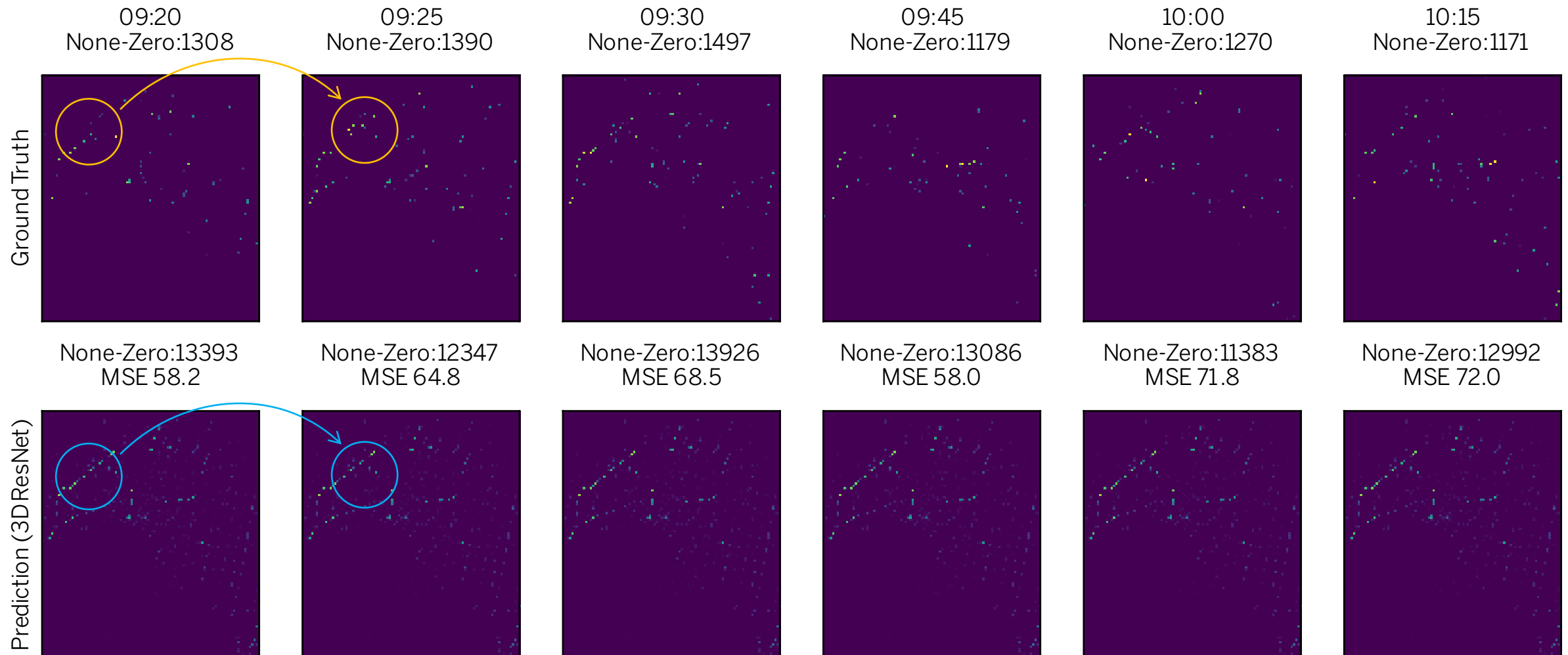
## Speed Comparison

Table 1. Different training times using Conv3D and sparse convolution (batch size = 2 with one P100 GPU). City names from left to right: Antwerp, Bangkok, Barcelona, Berlin, Chicago, Istanbul, Melbourne, Moscow.

City	ANT	BAN	BAR	BER	CHI	IST	MEL	MOS
Non-zero Rate (per batch)	0.0079	0.0072	0.0023	0.0303	0.0085	0.0481	0.0039	0.0758
Conv3D-UNet (batches/s)	0.98	0.98	0.98	0.98	0.98	0.98	0.98	0.98
Sparse-UNet (batches/s)	7.76	6.82	8.30	5.56	7.87	3.53	8.45	2.18

MSE 61.59 for Extended Challenge

# Performance Bottleneck



# Conclusion



We have proposed two new approaches, **3DResNet** (more accurate than 2D Convolution) and **Sparse-UNet** (much faster in modelling), for the challenges in traffic4cast 2021.

In the future:

- (1) Study further on sparse convolution in regression problems
- (2) Combine different feature learning approaches together
- (3) Consider the transport domain knowledge in designing neural networks
- (4) Improve the spatial accuracy in local area.

# Thank you!

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## Large-scale Traffic Prediction using 3DResNet and Sparse-UNet

Authors: Bo Wang, Reza Mohajerpoor, Chen Cai, Inhi Kim, Hai Vu

Presenter: Bo Wang

Paper: <https://arxiv.org/abs/2111.05990>

Code: <https://github.com/resuly/Traffic4Cast-2021>

10<sup>th</sup> Dec 2021