

## Basic Information

Chinese name: 成泽森

Pinyin: Cheng Ze Sen

English name: akko

Place of Birth: Hunan Yongzhou (湖南永州)

Birth Date: 1999.11.10

Major: Electronic Science and Technology

CET6: 522 (Listen:182 Read:211 Writing:129)

Scores: 91.12      GPA: 3.92      Rank: 1/88

I'm an undergraduate student of South China University of Technology (华南理工大学)



Now, I'm fatter...



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## Awards

1. National Scholarship (2017 ~ 2018) [Intellectual education: 1st, Comprehensive assessment: 1st]
2. The Second Prize Scholarship (2018 ~ 2019) [Intellectual education: 1st, Comprehensive assessment: 3st]
3. Cstro 16th Conference -- AI and Big Data Part -- ROI Segmentation Challenge -- Head and Neck OAR track --  
Champion
4. National College Students Mathematical Modeling Competition Provincial Third Prize
5. Taidi Cup 8th Data Mining Challenge Second Prize



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## Research Experiences

1. Multi-media Emotion Recognition (SRP project From 2019.6 to 2020.4)
2. Sun Yat-sen University Cancer Center Research Assistant (From 2019.8 to now)
3. SCUT-Robot lab Member of vision group (From 2019.9 to now)

### Detailed:

1. In this project, I'm mainly in charge of implementation. We try to use neural network to extract the emotion feature representation from Multi-media data like image, video etc. Then we use regression and classification model to model the emotion state.
2. I'm the research assistant of MIACA research group Sun Yat-sen University Cancer Center. I focus on medical image segmentation especially multi-modality medical image segmentation. Besides, I am also interested in medical image registration and medical image generation.
3. I'm a member of Vision Group. I'm mainly in charge of base station. Base station can use object detection, object track and motion forecasting to strategy calculation so that it can provide some strategy support for other robots.

## Publications

《基于深度学习的医学图像配准综述》

p.s. [**rejected**]

《Auto Segmentation of Pelvic OARs On MRI Multi-Sequence Using An Fused-Unet》

p.s. [**received**] (AAPM 2020 **oral presentation** & **BLUE RIBBON ePOSTER**)

《A Novel Hybrid Network for H&N Organs At Risk Segmentation》

p.s. [**received**] (ICBIP 2020 **oral presentation**) -- Corresponding SCI Under Review

《Attention V-Net: A Residual U-Net with Attention Gate Block for Lung Organ At Risk Segmentation》

p.s. [**received**] (CSAE 2020) -- Corresponding SCI Under Review

《ZigZag U-Net: Multi-stage medical segmentation network》

p.s. [**to be submitted**] (inspired by CB-Net)



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## Research Interest

Because of my past research, I may be interested in medical image segmentation, object detection and object track. Actually, I'm not limited to specific computer vision tasks. I'm interested in computer vision tasks which are meaningful and practical. For instance, the tumors and organs at risk automatic contouring on MRI Multi-sequence. MRI Multi-sequence tumors and organs at risk manual contouring is really time consuming and clinical knowledge intensive. So, we use Multi-modality model to perform MRI Multi-sequence automatic contouring which largely reduce the contouring time.

## The implementation of VectorNet

1. Representing trajectories and maps
2. Constructing the polyline subgraphs
3. Global graph for high-order interactions

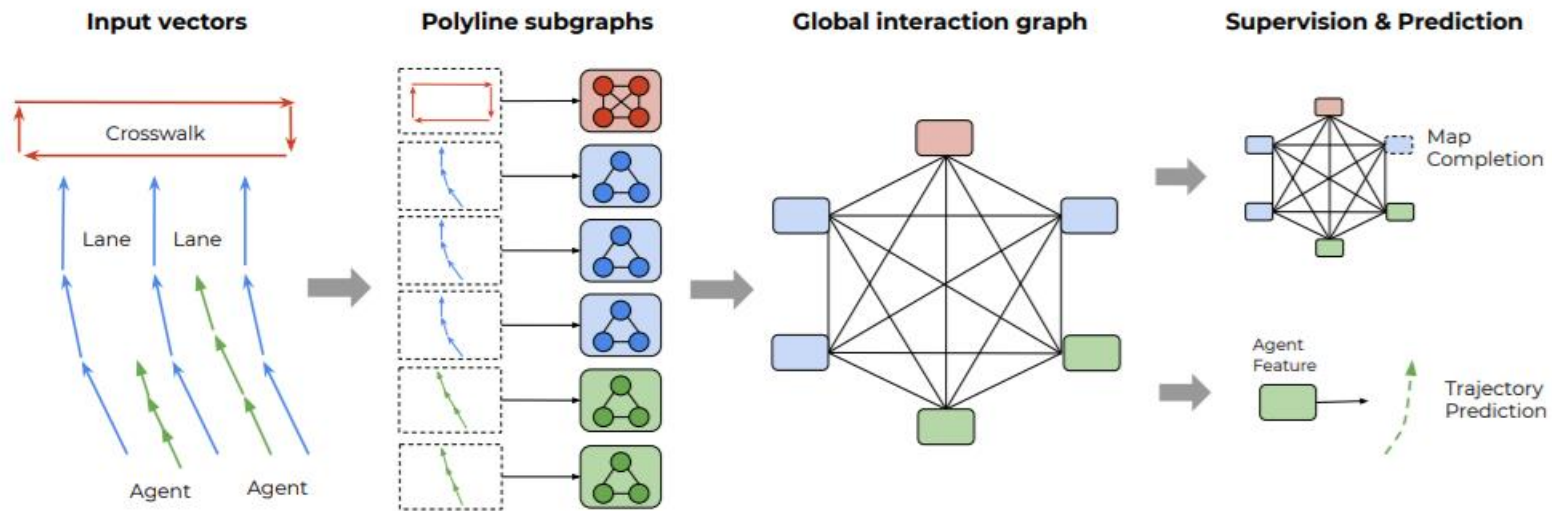


Figure 1. The total workflow of VectorNet

## Representing trajectories and maps

There are only the agent trajectory and lane needed to represent according to the requirements. In order to represent lane better, I use average lane width to calculate lane two-side edge line. Besides, I split the trajectory into observed trajectory [0s, 2s) and future trajectory [2s, 5s).

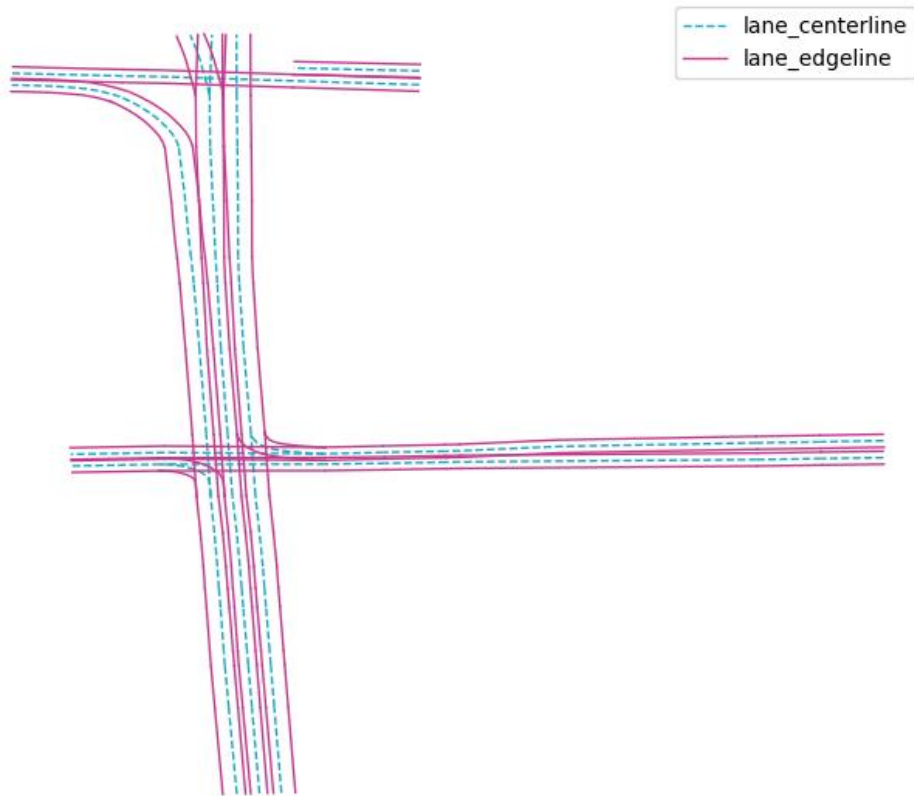


Figure 1. lane centerline and edgeline representations

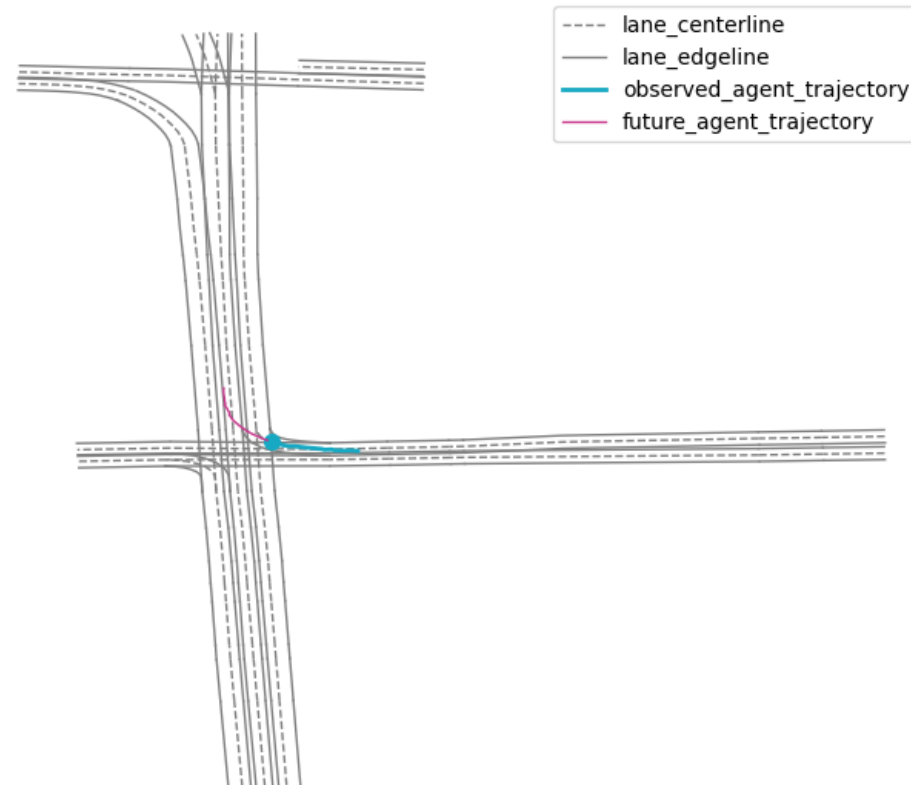


Figure 2. lane centerline and edgeline representations, agent observed trajectory, agent future agent trajectory

## Representing trajectories and maps

Because all of the lanes and trajectories are represented by point sets initially. So, we need to incorporate neighboring two point to vectorize these points and construct vector sets. Then we keep vector sets and remove point sets. Because the future trajectory sample points are our predict target, so we keep them.

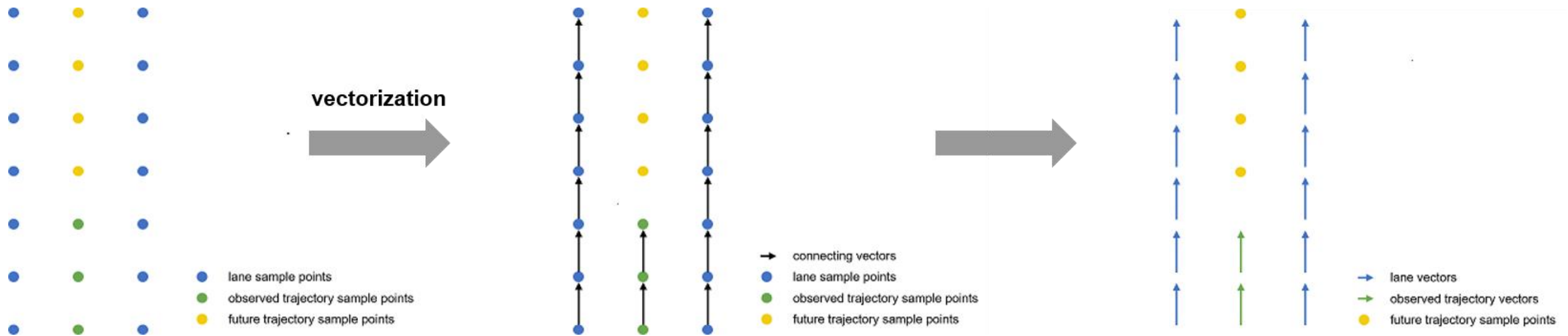


Figure 3. The process of vectorization of agent and observed trajectory sample points



## Constructing the polyline subgraphs

According to the paper, one vector set can be represented a polyline. As stated before, one lane represent a vector set and the observed is also a vector set. So, we need to see lane vector sets and observed trajectory vector sets as polylines. Then constructing the polyline subgraphs to convert them to polyline level features.

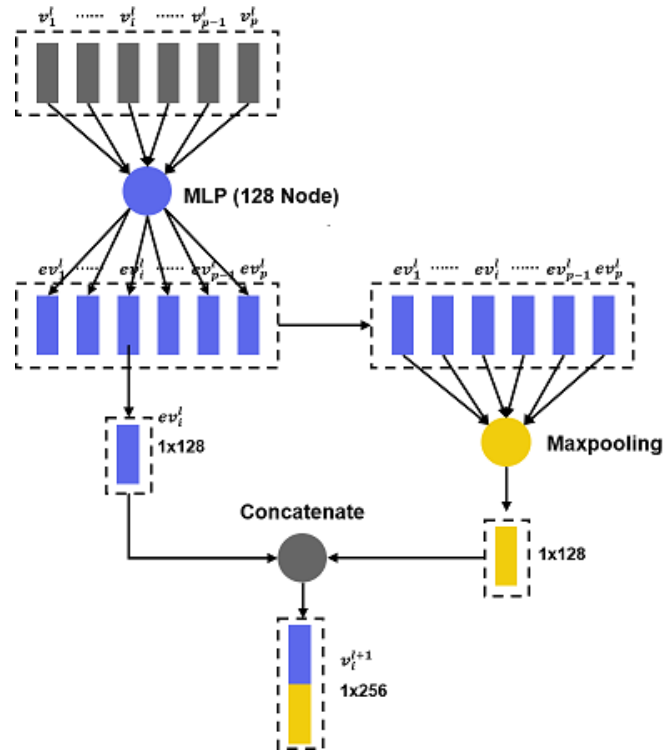


Figure 4. The detailed operation flow about how to get  $v_i^{l+1}$  from  $\{v_1^l, v_2^l, \dots, v_i^l, \dots, v_{p-1}^l, v_p^l\}$

$\{v_1^0, v_2^0, \dots, v_i^0, \dots, v_{p-1}^0, v_p^0\}$  representing raw vector sets.  
 $\{v_1^0, v_2^0, \dots, v_i^0, \dots, v_{p-1}^0, v_p^0\}$  representing vector sets which has been operated  $l$  time.

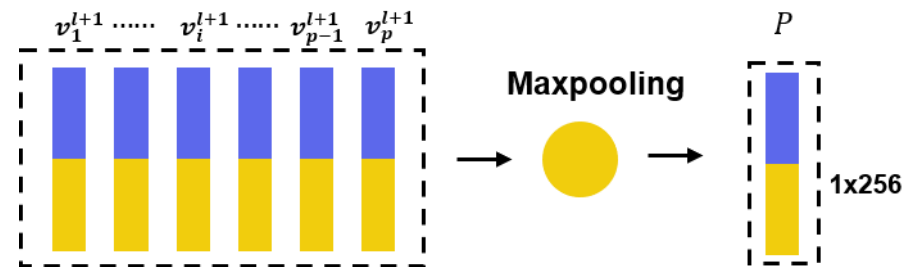


Figure 5. The detail operation flow of aggregating  $\{v_1^{l+1}, v_2^{l+1}, \dots, v_i^{l+1}, \dots, v_{p-1}^{l+1}, v_p^{l+1}\}$  to polyline level feature  $P$

## Global graph for high-order interactions

We refer to the self-attention mechanism and analyze the detailed calculation flow of self-attention mechanism. The global graph neural network is related to the multi-head attention to a great extent. They copy raw feature map into three copies and call them query, key, value. Their aim is to use matrix multiplication to get weight matrix from query and key. Then using matrix multiplication to assign weight matrix on value matrix.

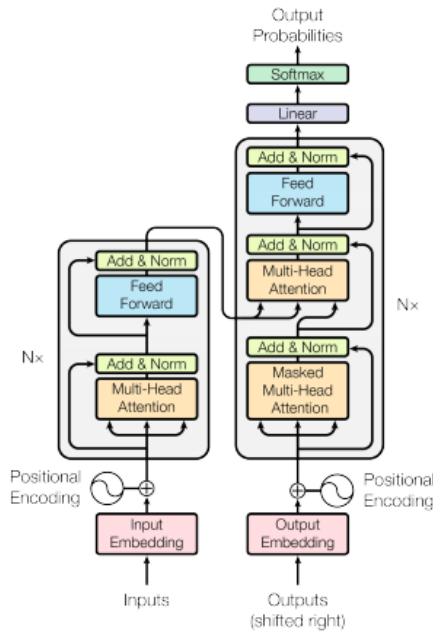
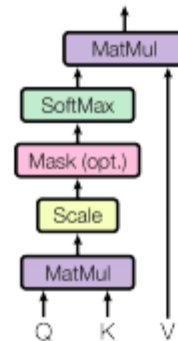


Figure 6. The total pipeline of self-attention mechanism

Scaled Dot-Product Attention



Multi-Head Attention

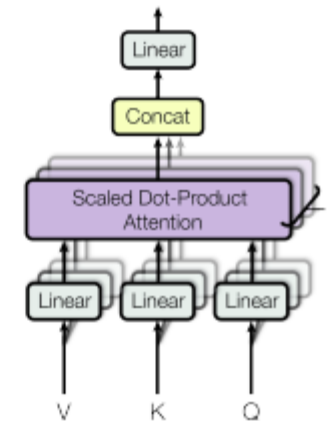


Figure 7. The detailed architecture of multi-head attention which is the reference of global graph of VectorNet

## Global graph for high-order interactions

It seems that it has nothing to do with the graph network. In fact, the matrix multiplication operation is just a type of fully connected interaction between different graph node. The graph node features shape as  $[b, L^f, n^p]$  where  $b$  denotes batch size,  $L^f$  denotes feature length or the width of global graph,  $n^p$  denotes polylines count. After interacting with other graph node, agent polyline feature is decoded by MLP and output agent prediction trajectory points.

