

# LYFT 3D OBJECT DETECTION FOR AUTONOMOUS VEHICLES

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TEAM

SPRINT 1



# PRODUCT DEFINITION

*The statement of the  
meaning of the product*

# PRODUCT MISSION

Find out the classes and positions of objects on the road

- Sensor data from 7 cams and 3 LiDARs
- Calibrations and Geographic data are available as well
- Using these data to find out the 3D volume (box) of every object and classifying the categories of objects



# USER STORIES

*How does the user think*

*Lyft, who is the sponsor of the competition, as a ridesharing company, want to use this algorithm in their autonomous vehicles system.*

# TARGET USERS

*Who will use our product*

## Ridesharing Companies

- Lyft, as the sponsor of the Kaggle Competition
- Uber, who uses LiDAR data as well in their system
- Other Ridesharing Companies who will be interested in autonomous vehicles.

## Startups Companies

- Since our works are open source, it could inspire some startups
- Even the data and parameters in the model could cause copyright problem, the algorithm and model themselves are free to use

## Other Researchers

- We will open source our code and basically every document. So it is more than welcome to check the code and algorithm out

# MINIMUM VIABLE

*At least, we need ...*

## Algorithms and Models

- The results from our method have to be at least better than from white noise

## Implementations

- A fully functional implementation of our algorithm

## Documentations

- A report about our method and the working theory
- Docs to help people to build and use our product and code

# PRODUCT ANALYSIS

*Detailed examination  
of the product*

# SIMILAR PRODUCTS

*What already exist*

## Waymo

- Google's self-driving car project
- Strong, powerful, but only dataset is available.

## Autopilot

- Tesla's self-driving car system
- Already in use as a commercial application.

## Others

- There are a lot of open source framework in the field of self-driving cars for us to study.
- Like Apollo Auto, etc.



# PATENT RELATED

*Is there any limitation*

## TOOLS

- Programming language
- Widely used library like numpy, OpenCV
- We only choose open source tools which are under liberal license

## ALGORITHMS

- The widely used ML and DL algorithms like UNet, GCN, GAN don't have a patent problem in our use case.

# ALGORITHM DESIGN

*Thoughts behind the  
algorithm*

# MAJOR COMPONENTS

*The major submodules of the algorithm*

## 3D Projection

- Lyft, as the sponsor of the Kaggle Competition
- Uber, who uses LiDAR data as well in their system
- Other Ridersharing Companies who will be interested in autonomous vehicles.

## Classification

- Lyft, as the sponsor of the Kaggle Competition
- Uber, who uses LiDAR data as well in their system
- Other Ridersharing Companies who will be interested in autonomous vehicles.

## Objec Detection

- Lyft, as the sponsor of the Kaggle Competition
- Uber, who uses LiDAR data as well in their system
- Other Ridersharing Companies who will be interested in autonomous vehicles.

# TECHNOLOGY SELECTION

*Which technology is the best, to us*

## Languages

- Python
- All the team members are familiar with Python
- The ecosystem (community) of Python is better than others in this field
- The native performance of Python is bad. But most of the performance sensitive works in our code are just a sort of DSL. The framework will handle the performance problems.

## DL Frameworks

- Since DL is one of the most necessary and important part in our algorithm, we have to carefully choose DL frameworks.
- Keras, to run simple model tests and demos.
- PyTorch, to complete more complicate DL task

## Other Libs

- Classic ML lib: scikit-learn
- Data Utils:
  - Official: nusenes-devkit
  - 3<sup>rd</sup>-Party: pandas, numpy, scipy
- CV lib: OpenCV (cv2 binding for Python)
- Viz lib: seaborn, matplotlib
- Others: some scripts from GitHub

# TEST AND VALIDATION

*How to evaluate our work*

*The Kaggle Competition has provided the way they evaluate the algorithm.*

## Threshold

- Intersection over Union (IoU):

$$IoU(A, B) = \frac{A \cap B}{A \cup B}$$

## Average Precision

$$\frac{1}{|thresholds|} \sum_t \frac{TP(t)}{TP(t) + FP(t) + FN(t)}$$

**THANK YOU**