FORMULA STUDENT

Institution of MECHANICAL ENGINEERS



Making a car drive itself

How to prepare for FS-AI

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Agenda

- The challenge
- Why you should do it
- Founding a team
- Technical aspects
- Software tools, development strategies and testing
- Tips for competition



Self-driving cars today

- Emerging industry set to revolutionise transport in the next decade
- Blend between engineering and computer science







The FS-AI challenge

- Create a FS car that can drive itself
- Split into two categories:

DDT



ADS





Why should you join?

- Exciting and innovative
- Actively contribute to the field
- Get hands-on experience
- Make a career out of it
- Fun



"Self-driving cars industry in the UK will be worth £28 billion in the next 17 years." - WIRED UK

https://www.wired.co.uk/article/driverless-cars-uk-self-driving-cars



Founding a team

- Multi-disciplinary team
- Bridge the gap between your Engineering and CS departments
- Autonomous cars aren't cheap start with a simulation
- Find academics to support you
- Start small (<10 people) until you figure out what you need to do
- ADS and DDT can be developed in parallel. Software can be made to be applicable to both



Integrating to your existing FS team

- Share your old platform
- Share knowledge and skills
 - But set clear boundaries at the beginning of the year
- Share resources tools, workshops, transportation, etc.
- Things to consider
 - Recruitment conflicts
 - Sponsors conflicts
 - Operations and events
 - Budgeting and sharing resources
 - Team identity social media, newsletters, etc.



Management

- Keep developers focused have people to deal solely with operations, events, financials
- Tight communication between subteams
- Plan the year ahead
- Regular review of progress
- Balance between engagement and hard work

Technical

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ADC DU

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Lidars

- High accuracy
- High output rate
- Output point cloud
- Expensive
- Fragile

Types:

- Planar
- Multi-beam
- Image-like
- Solid-state





Cameras

- High output rate
- Lots of data
- Versatile
- Cheap
- Computationally expensive
- Huge variety choose carefully

Types:

- Monocular
- Stereo







IMU and GPS

- Huge variety!
- Estimate location in 3D space
- From 5\$ to 5,000\$
- Quality varies greatly
- You can get standalone IMU and GPS
- You can get integrated INS (IMU+GPS)
- You can also get RTK GPS





Computing

- Process intensive tasks
- Desktop PC vs industrial computers
- Arm64 vs x86
- Difficult to estimate how much power you need. To be safe go with more than you need right now
- Distributed computing
- Dedicated computers
- Strap a laptop first before you buy PCs















Perception

- Extracting meaningful information from sensors
- ie. from sensor data figure out where the track is
- What's a cone and where are they relative to the car (in 3D)?
- Could also be pedestrian detection as safety feature







Perception (LiDAR)

- Process LiDAR:
 - Down sample
 - Extract ground plane
 - Limit interested region eg. Heights
- Cone detection in 3D:
 - Euclidean clustering basic approach
 - Feature matching eg. corners, edges
 - CNNs for pointclouds get the algorithm to learn the features









Perception (Camera)

- Process camera:
 - Compression?
 - Grayscale?
- Detect cones:
 - Colour thresholding and tracking
 - Feature matching eg. SIFT
 - CNN object detection (YOLO) get the algorithm to learn the feature matching
- 3D location:
 - Structure from stereo
 - Structure from motion
 - LiDAR Camera calibration
- Drivable area / track segmentation







Localisation and Mapping

- First time we go around the unknown track we use only sensor data
- Let's use this to build up a map or model of the track and determine car's position
- Helps us to plan what's coming up when we see it again







Localisation and Mapping

- What to record in our map?
- SLAM Simultaneous Localisation and Mapping
- Dense SLAM eg. ORB-SLAM
 - Maps more general features eg. Edges, lines
 - More information stored for localisation
- Landmark based SLAM eg. EKF SLAM
 - Maps only what we need cone locations in 3D
 - Depends on cone extraction eg. From LiDAR
- Also 2D approaches mapping in ground plane





Localisation and Mapping

- Localisation Have to locate car in the map in order to plan
- Effector noise eg. friction
- Odometry sensor noise
- Better localisation estimates by fusing IMU, GPS, odometry etc.
- But still error prone
- Solution: relocalise by matching sensor data to that expected based on map
- Probabilistic framework formalises uncertainty and belief states







Path planning and Control

Local planning

Plan a trajectory and follow it only on immediate input data



Global planning

Once you have a full map of the track, can you optimise to go faster?



https://autorally.github.io/



Local planning

- No view of the whole track must be conservative
- Can use traditional robotics map planners
 - Based on map
 - Usually safe and non-dynamic
 - Eg-A*
- Constrained environment can make a specialised local planner
 - Eg. midpoints of track







Global planning

Now that you have a map:

- Can you optimise a better trajectory?
- How can you process your map data?
- Hit all apexes?
- Take into account vehicle dynamics?
- How feasible is that path for the car to follow?





Control

- Once you have a trajectory, how can the car follow it?
- Software engineers can assume the car is a black box:
 - Steering
 - Speed
 - Torque
- Hardware engineers should:
 - Simplify control
 - Make it reliable



Control algorithms

- PID Controller
 - Trivial
 - Doesn't forward sample
 - Doesn't take into account dynamics model
- MPC Controller
 - Forward samples and optimises control
 - Based on costing
 - Requires knowledge of the dynamics of the car





Al and Machine Learning

- Use breakthroughs in deep and reinforcement learning to control the car
- End to end camera image input directly to control outputs
- Imitation learning
 - Record a human driving, model learns to copy the expert
- Reinforcement learning
 - Model has a go and given feedback on it's performance
 - eg. How far it got round track



Software tools, development and testing

- Middleware software eg. ROS
 - Message-passing between processes
 - Implementation of commonly used functionality
 - Package management
 - "Nodes" (programs) in most programming languages C++, Python, MATLAB etc.
- Repo management eg. GitLab
 - Code reviews
 - Software testing continuous integration
 - Version control



Software tools, development and testing

- Simulation for developing algorithms / ideas eg. Gazebo
 - Ours (with FSUK DDT car model) is open sourced:
 - https://github.com/eufsa/eufs_sim
 - <u>https://github.com/Microsoft/AirSim/wiki/technion</u>
- You can develop a small-scale testing platform that can be tested anywhere(eg. a wheelchair)
- Datasets for verifying and evaluation on more realistic data:
 - Ours on wheelchair and from 2018 FSUK competition is shared:
 - https://github.com/eufsa/datasets
 - <u>https://github.com/AMZ-Driverless/fsd-resources</u>





To prepare for the competition

- There are a lot of cool things you can do but always justify why you are doing them!
- Always think of how you can generalise your solution! 90% of the things you do on the car is also being done in industry right now
- Have clearly defined tasks for all members attending
- Have a cool Plan A
 - But if that fails be prepared with Plan B



Problems

- It's FS problems are always around the corner
- Good idea to already have redundancy
- Draw up a list of things that can go wrong they probably will!



Sponsors & QA







