Geospatial Data Analytics Projects undertaken by Ryan Watson Consulting, Melbourne, Australia

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Overview of Presentation

- Company profile
- Current importance of geospatial data analytics
- Victorian road accident data analysis project
- Road accident hotspot prediction using machine learning
- Canopy coverage analysis using 2D and 3D data
- Whitehorse City Council Project
- Conclusions



Company Profile - Ryan Watson Consulting Pty Ltd

- Established in October 2020
- Based in eastern suburbs of Melbourne
- Built on research with Swinburne University in:
 - Internet of Things
 - Big Data
 - Smart Cities







• Open to collaboration with other universities



Current Importance of Geospatial Data Analytics

- Using geospatial technologies to gather location information provides us with data that can give a deeper understanding of many issues. This spatial data can answer questions, help with predictions, analysis and decision making
- Spatial data can be combined with other data from various sources weather data, census data, satellite imagery, aerial photography, social media data, etc
- Geospatial data can be used in a wide variety of different industries urban planning, transportation, natural disasters, health care, agriculture, and many

more





Victorian Road Accident Data Analysis Project

- Victorian road accidents and fatalities datasets analysed using python spatial libraries
- Various statistical tests employed to establish spatial autocorrelation and distribution of road accidents
- Geospatial factors contributing to road accidents such as land use, built infrastructure, road characteristics and population density found from open datasets and their connection to road accidents established
- Street view images of road accident sites analysed and included in a machine learning model of road accident hotspots

Reference: Geospatial Factors Applied to Road Accidents: A Review *Journal of Advances in Information Technology*, **15**, 3, March 2024



Victorian Road Accident Data Analysis Project

- Data sources
 - Federal, state, local open geospatial data for Australia
 - Open Street Map, Google Earth for World
- Geospatial Analysis
 - Vector, raster and point cloud data
 - Spatial join, coordinate projection, etc
- Geospatial statistics
 - Kernel Density Estimation, Getis-Ord, Geographically Weighted Regression, etc
- Machine Learning
 - sklearn, keras, SHAP, etc
- Visualization
 - Maps, Choropleths, heatmaps, etc



Victorian Road Accident Data Analysis Project











Some examples of data visualization from this project



Road Accident Hotspot Prediction using Machine Learning

- It is important to understand the contributing factors of road accidents and their impacts in order to design safer roads
- Contributing factors can be categorised based on different characteristics such as road, vehicle, human, speed limit, and others
- Machine learning models use accident and geospatial environmental data to predict hotspots
- Model predictions need to be explained by techniques such as Shapley Additive Explanations (SHAP)





Environment and Road Accidents

- A road can also be defined in the form of its environment features such as presence of trees, open space, buildings, etc
- Identifying association between environmental factors and road crashes can be crucial for making roads safe
- Deep learning can be used to extract environmental features from street view images and feed these features along with other contributing features that lead to road accidents





Environmental Features

- Street environmental features can be extracted using Semantic Image Segmentation such as PSPNet – A very popular architecture
- Segments in these images are obtained using Detectron2 – A system developed by Facebook AI Research which implements state-of-the-art object detection algorithms
- Segment information can be used for machine learning algorithms and feed these features





Feature Extraction



From: Uncovering the association between traffic crashes and street-level built-environment features using street view images, Sheng Hu et al, <u>https://doi.org/10.1080/13658816.2023.2254362</u>



Feature Extraction

- Instance is the desired environmental feature
- Area is the number of pixels for each instance in that image

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Image Name	Instance	Area	
0016E5_05460.png	traffic light	0	
0016E5_05460.png	car	0	
0016E5_05460.png	person	0	
0016E5_05460.png	road	63828	
0016E5_05460.png	tree	7441	
0016E5_05460.png	sky	18902	
0016E5_05460.png	building	63200	
0016E5_06570.png	car	0	
0016E5_06570.png	road	50617	
0016E5_06570.png	sky	14952	
0016E5_06570.png	pavement	4865	
0016E5_06570.png	building	82717	





What Next?

Develop a small model to identify the relevance of environmental features in road crashes using Deep Neural Network

Limitations:

- Street View Images
- Relevance of captured Images
- Redesigned roads and developed areas





Canopy coverage analysis using 2D and 3D data

- Local Councils are trying to increase tree canopy coverage to reduce urban heat islands
- 2D (RGB imagery) and 3D (lidar point cloud and stereo vision) data is being collected by remote sensing to monitor changes in canopy coverage over time
- Lidar point cloud data is obtained from 3D scanners that project light at objects and interpret the scattered ray
- It can be analysed to categorise objects using deep learning.
- Our Interns have been outplaced to Whitehorse City Council GIS Team to analyse RGB imagery and lidar datasets



Whitehorse City Council Project

- Tree canopy cover
- Whitehorse Local Government Area in Victoria
- Cooling and Greening Melbourne







https://map.whitehorse.vic.gov.au/index.html Ryan Watson Consulting Pty Ltd

Background to Study

- Cooling and Greening Melbourne (2019)
 - <u>https://www.planning.vic.gov.au/guides-and-resources/strategies-and-initiatives/plan-melbourne/cooling-and-greening-melbourne</u>
- DELWP, RMIT University, CSIRO, Clean Air and Urban Landscapes Hub
 - supported by Melbourne Water
- CSIRO's Urban Monitor create first high resolution map that records baseline of urban vegetation across Melbourne, mapping imagery in 2014 and 2018



WCC Targets

- Cooling and Greening reported that canopy cover across Whitehorse was measured at 18%
- This represents a 2.3% canopy cover decrease from 2014
- Urban Forest Strategy Targets
 - Increase tree canopy to 27% by 2031
 - Increase tree canopy cover to 30% by 2050





Project Deliverables

- Benchmarking tree cover change, urban heat and heat vulnerability
 - 2014 and 2018 tree cover
 - 2014 \rightarrow 2018 tree cover change
 - 2018 urban heat
 - 2018 heat vulnerability index

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WCC Dataset

- LIDAR data from 2018
- Aerial Imagery
- 52.8 GB
- .las file format 3D point cloud
- Over 100 .las files to cover WCC
 - Each >350MB



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Analysis Techniques

- QGIS: open source, many plugins eg LAS tools, Qgis2threejs
- Python libraries: detectron2, detectree



https://qgis.org

Python detectree



Bosch M. 2020. "DetecTree: Tree detection from aerial imagery in Python". *Journal of Open Source Software, 5(50),* 2172. doi.org/10.21105/joss.02172



DEM, DTM, CHM for Canopy Cover

• DSM: Digital Surface Model

• DTM: Digital Terrain Model



- CHM: Canopy Height Model
 - CHM = DSM DTM





Canopy Height Analysis (QGIS)





Point Classification

Classification of LIDAR Points: en331n5809



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Canopy Height Analysis (Python)



Canopy Height Model Classification



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Summary and Conclusions

- RWC registered in 2020 in Melbourne, Australia
- Focus on research into IoT, Smart Cities, Big Data presented in conferences and journal papers
- Actively engaged in GIS projects
 - Road accident analysis and hotspot determination
 - Canopy cover to Victorian Local Government Area
- Collaboration with Swinburne University
- Open to collaboration with other universities



Questions?



