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UNSW Built Environment



City Analytics for Sustainable Mobility





City Analytics



City Housing



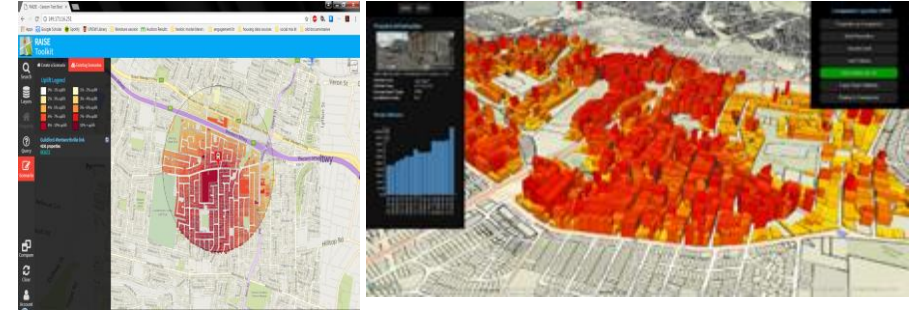
City Shaping



City Wellbeing



**Masters of
City Analytics**



**City Analytics
Research Program**

**City
Analytics**



**City Analytics
Laboratory
CAL**



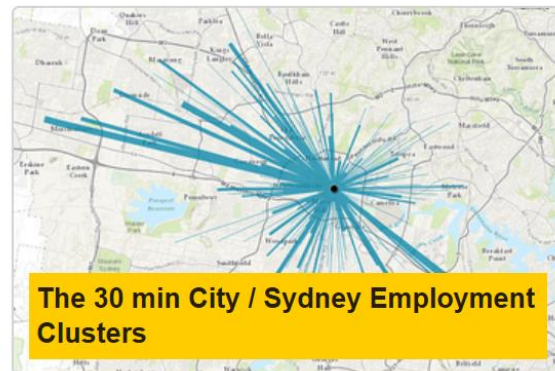
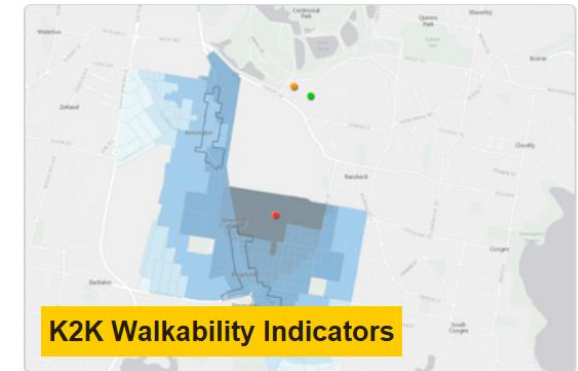
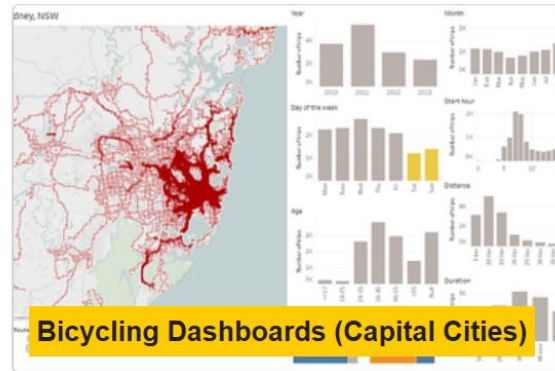
CityViz – Making urban big data visible

CityViz is a new data visualisation and analytics initiative by City Futures capitalising on our unique access to urban big data for Sydney. Over the next few months a range of data will be presented on this site that will start to build a comprehensive and integrated visual depiction of our changing city.

To start, we have assembled new data on Sydney's emerging housing market – affordability, strata development and our 'million dollar property' map.

In coming months we will be adding to and updating these data with new data on, for example, urban wellbeing, transport and bike use, health services and other newly available datasets. Watch this space!

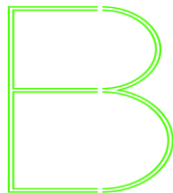
City Movement Indicators



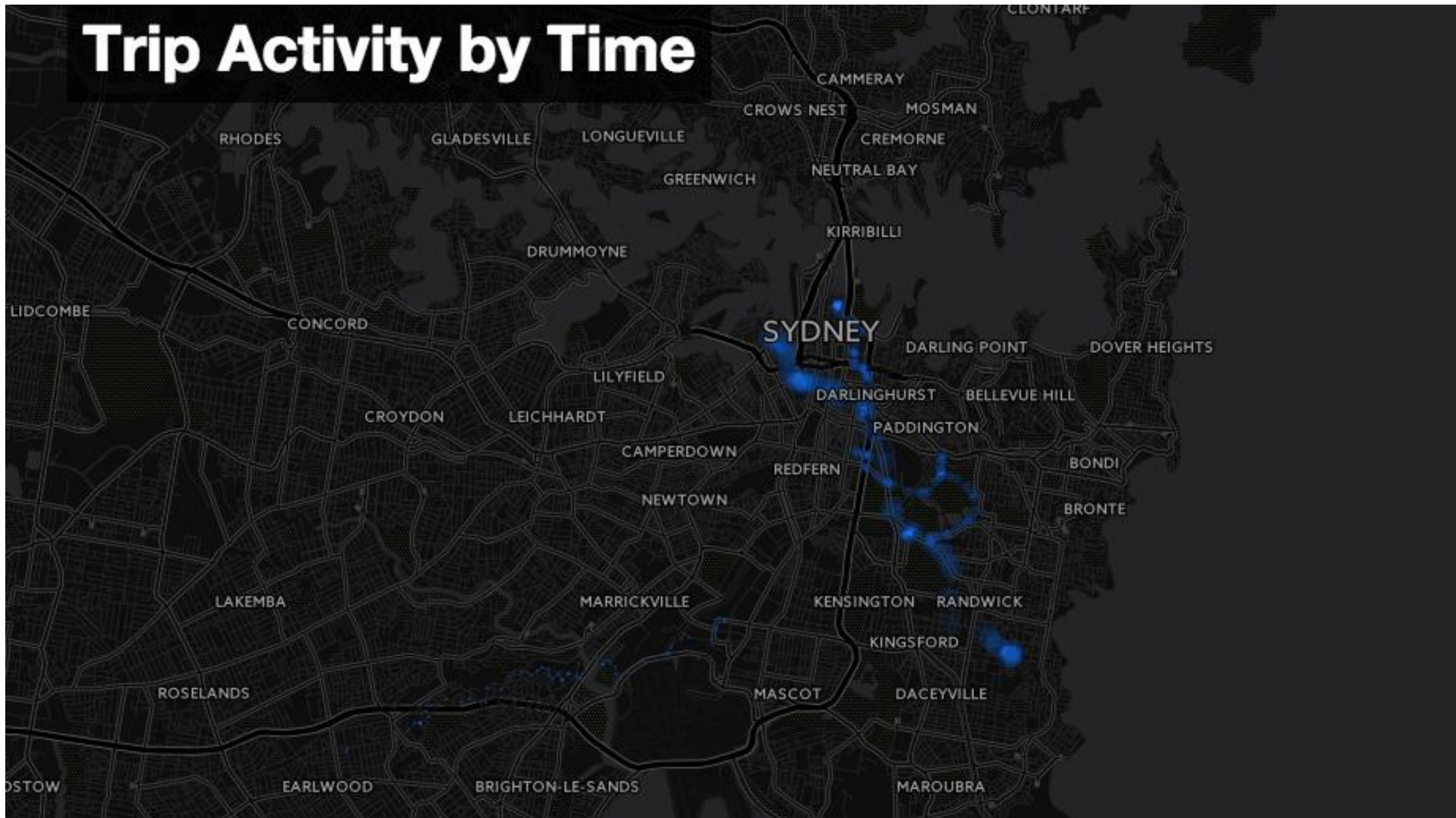
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CityViz – Cycling in Sydney over a 24 hour period



Trip Activity by Time



Communicating and visualising cycling data

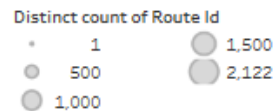
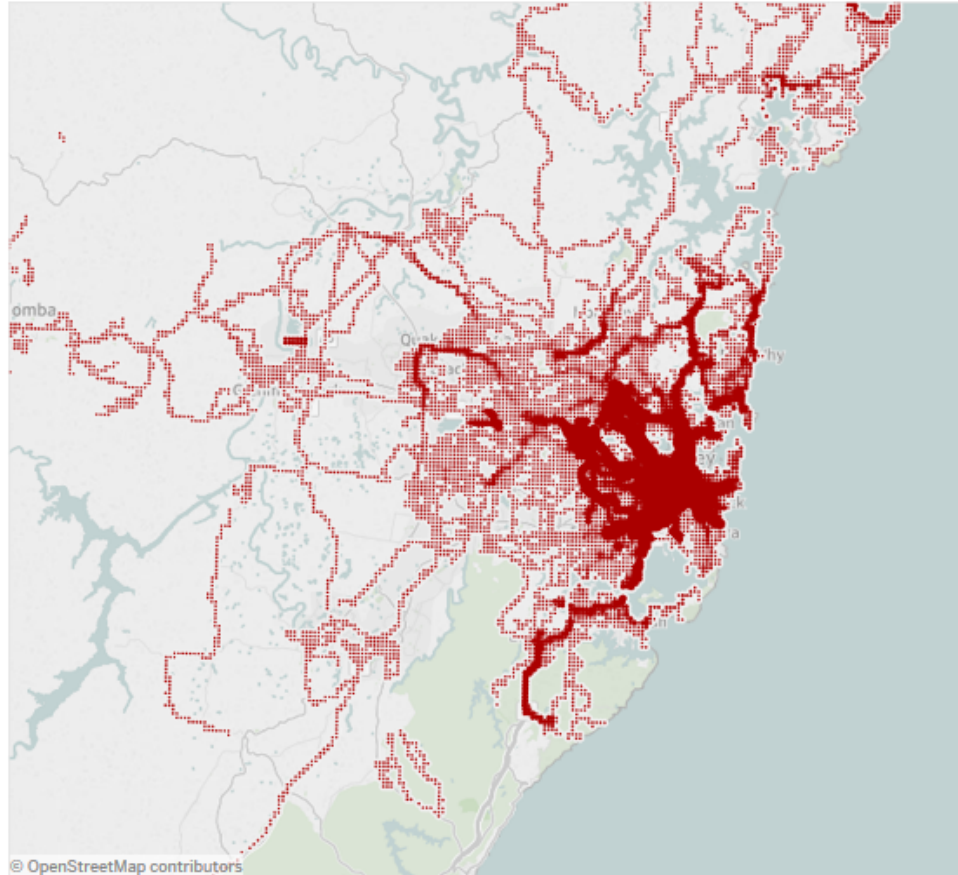


Bicycling Dashboards (Capital Cities)

CITY FUTURES RESEARCH CENTRE



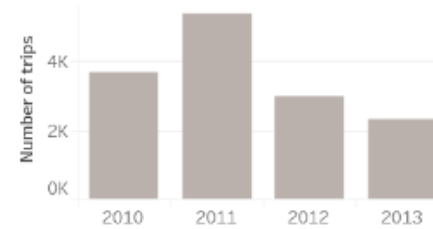
Greater Sydney, NSW



Number of trips:
14,313

Number of cyclists:
1,164

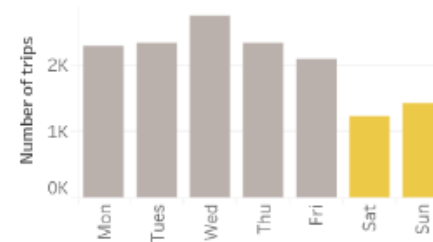
Year



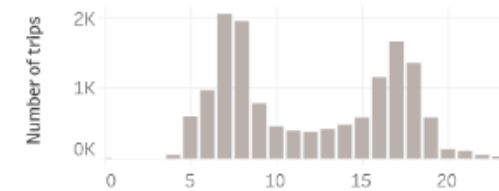
Month



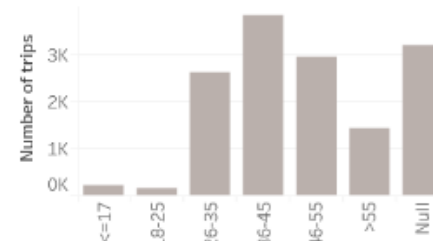
Day of the week



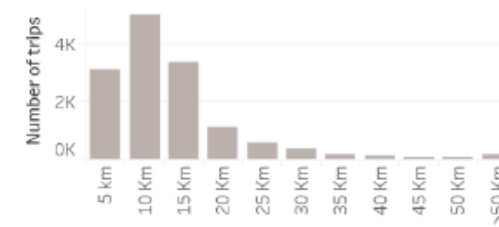
Start hour



Age



Distance



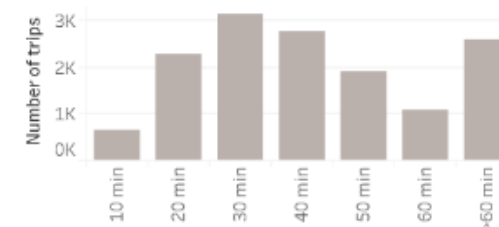
Gender



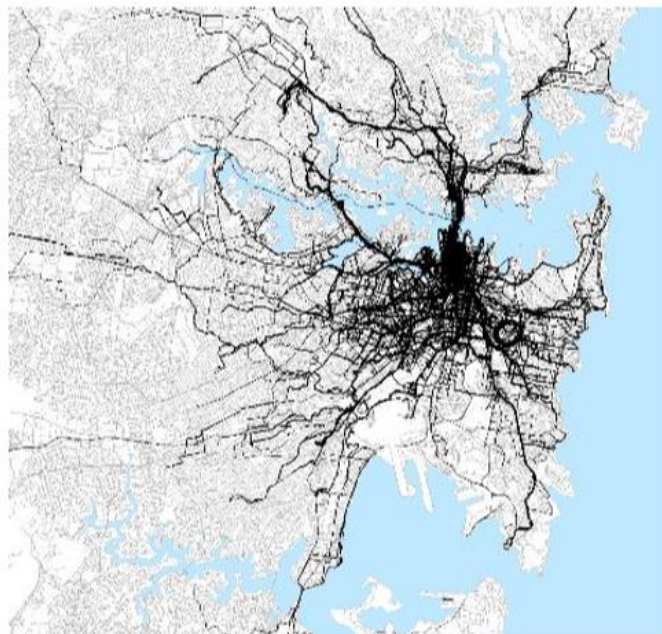
Purpose



Duration



Integrating cycling data to varied contextual information



RiderLog Tracks

Roads

a. GPS tracked cycling routes map



Conservation & Recreation Zones

Commercial Centres

b. Slope, parks and commercial centres



Cycling lanes shared with Traffic

Cycling lanes separated from Traffic Roads

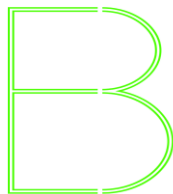
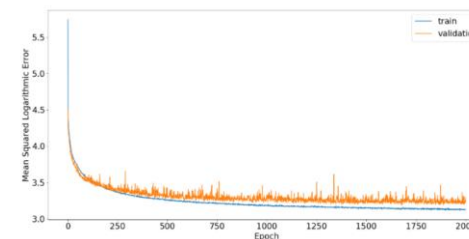
c. Cycling infrastructure

Table 2. Summary of multiple regression analysis – Dependent variable: Route_Distance

Predictor Variable	B	SE _B	β	Part-Correlation
Intercept	245.896	243.338		
Gender	298.284	74.855	.028	.027
Distance to commercial areas (DistComm)	1.040	.022	.343	.318
Distance to Parks and coastal areas (DistParks)	1.253	.018	.525	.470
% of the overall track with slope <= 2%	3551.991	200.030	.150	.121
% of the overall track with slope >2 and <= 5%	-669.270	333.528	-.021	-.014
% of the overall track with slope > 10%	12469.117	504.025	.258	.219
% of the overall track with mixed traffic lanes	5822.758	180.877	.227	.168
% of the overall track with no cycling infrastructure	-1129.253	225.874	-.035	-.034

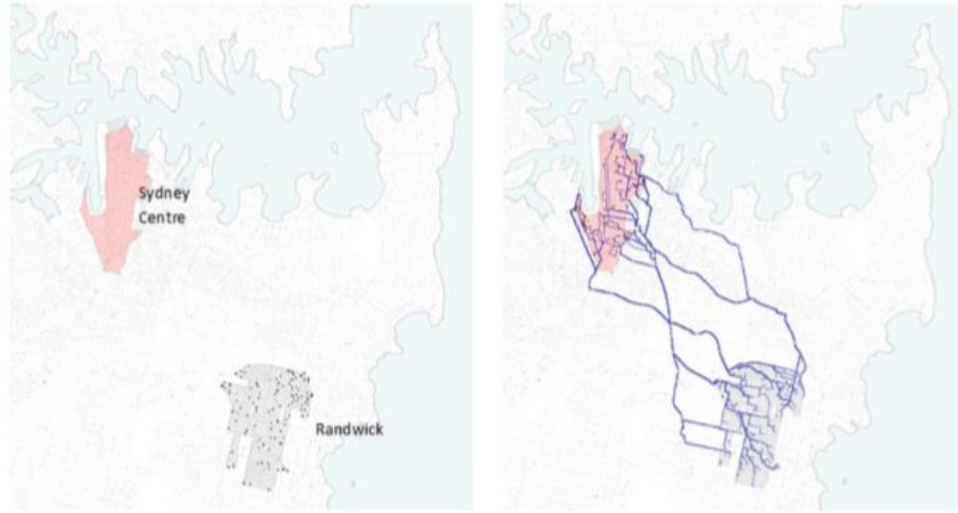
Note: * $p < .05$; B= unstandardized regression coefficient; SE_B= standard error of the coefficient; β =standardized coefficient

	Multiple Linear Regression	Deep Learning
4-bin Accuracy	50.43%	75.61%



Modelling and simulation of cycling patterns (agent-based modelling)

Ruled-based simulations (current context)



(a) Initial condition of a simulation



(b) Result of Simulation A

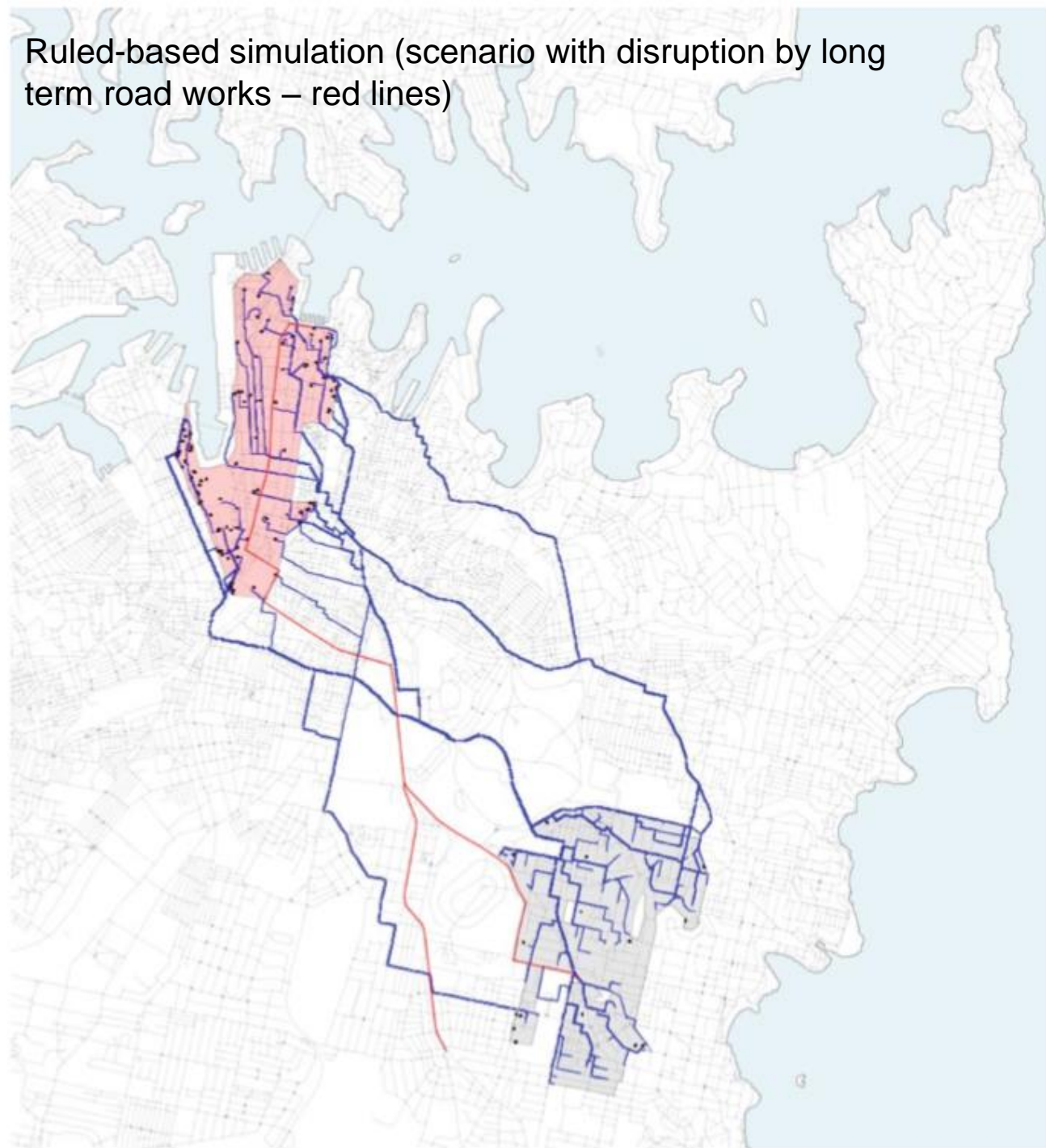


(c) Result of Simulation B

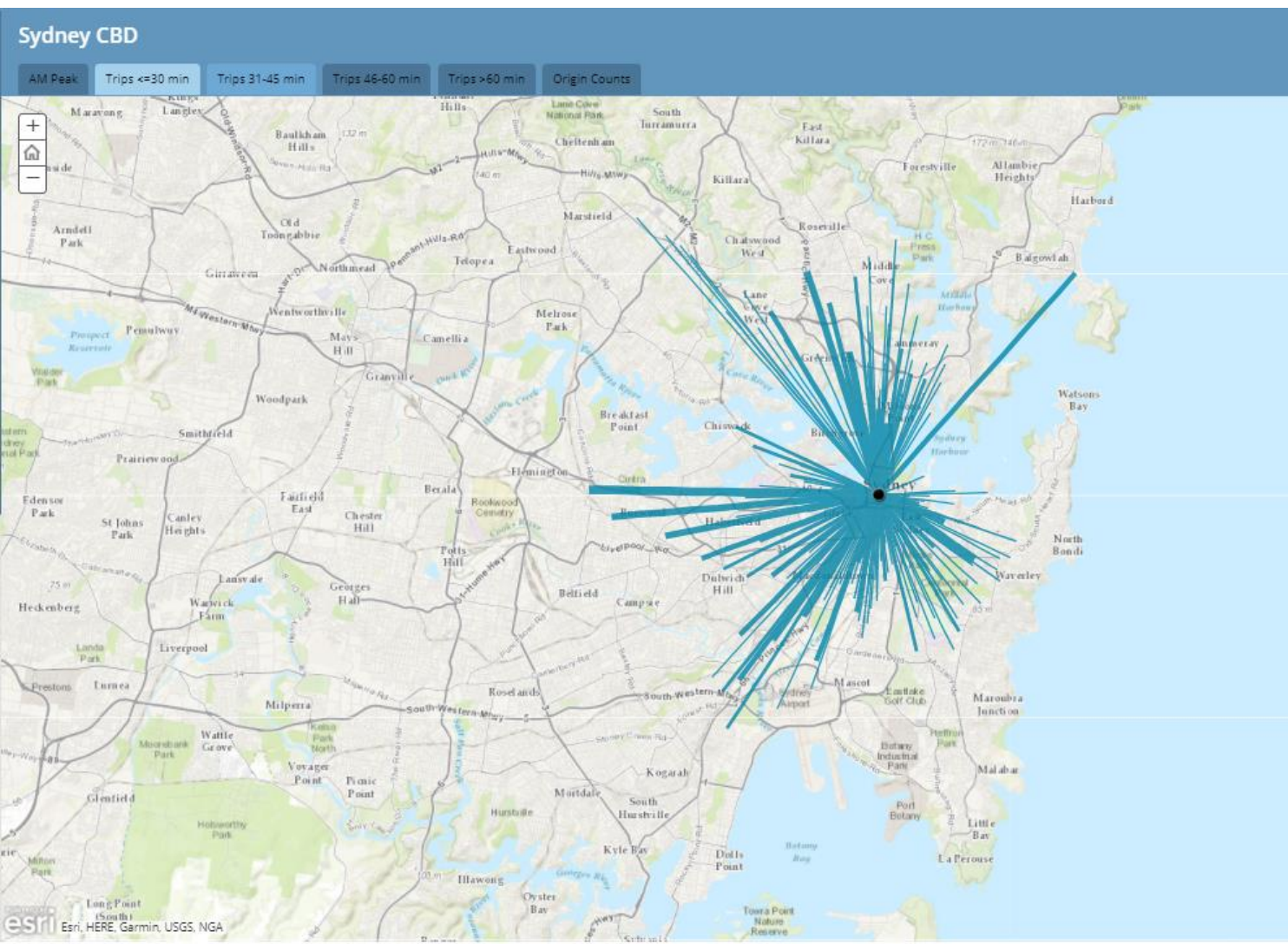


(d) Result of Simulation C

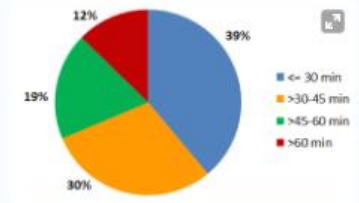
Ruled-based simulation (scenario with disruption by long term road works – red lines)



30 minute city, Sydney CBD

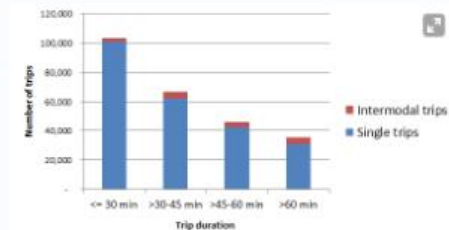


Total trips to Sydney CBD AM peak = 244,999 trips
39% = % trips with duration <= 30 min



Click on the image to enlarge view.

98% of trips with duration <= 30 min are single trips



Click on image to enlarge view.

Click on the map features for additional information; zoom in/out for better visualisation.

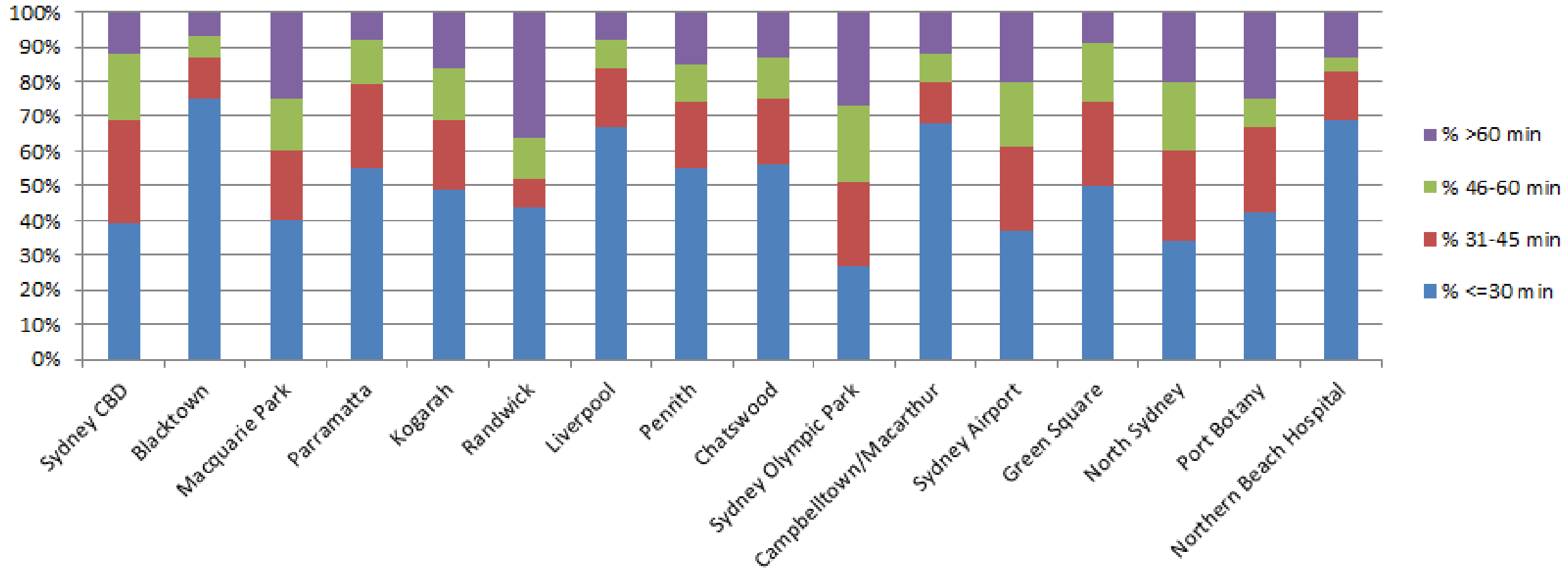
SydneyCBD_centroid



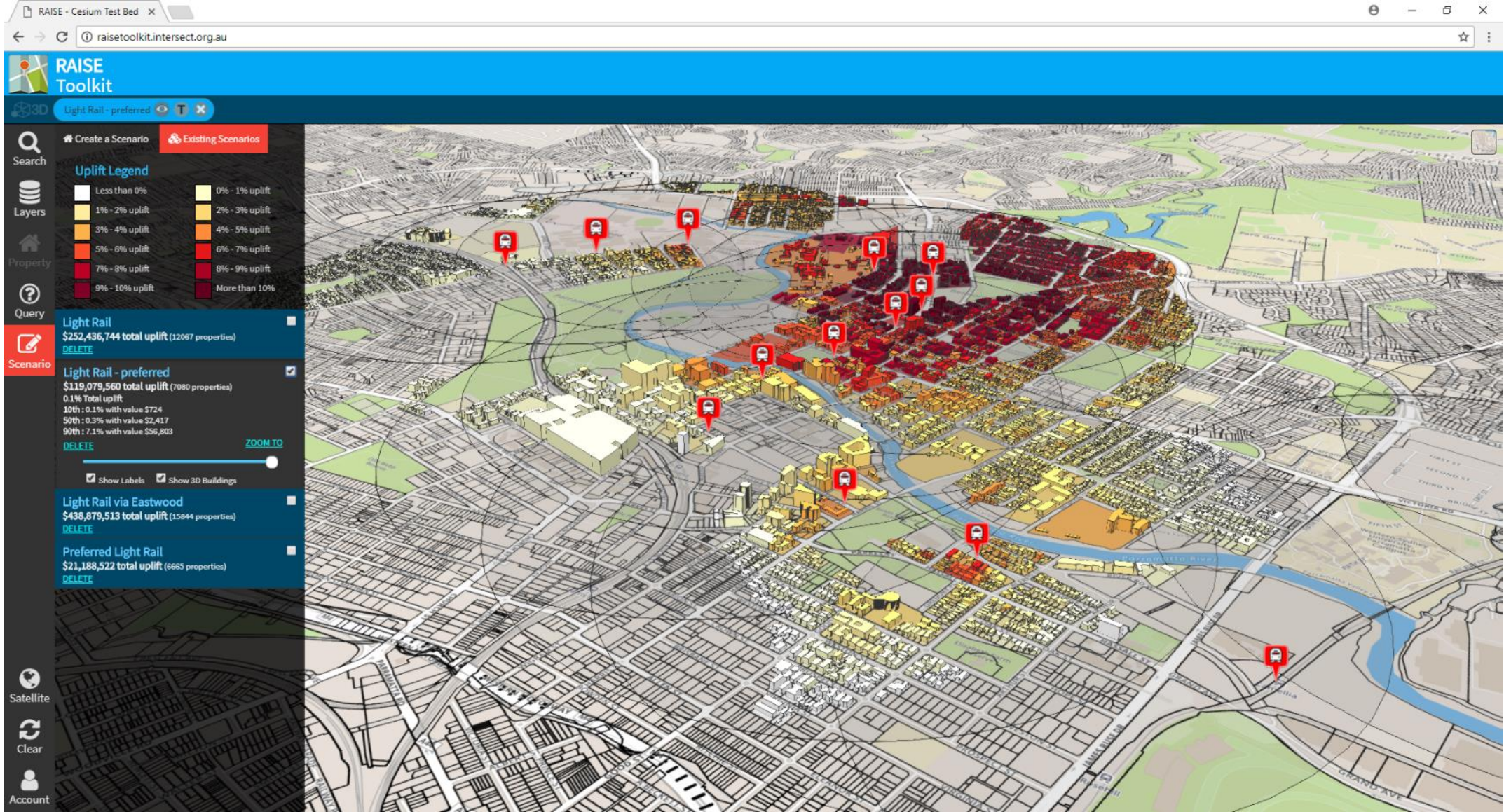
SydneyCBD_30min

- Number of trips:
- > 4,000 to 5,200
 - > 3,000 to 4,000
 - > 2,000 to 3,000

% Duration of trips to Destination, AM Peak typical weekday



Parramatta Light Rail options



Concluding thoughts

- Big data is providing information about active and public transport at a resolution never available before.
- Handling and integrating big data is not an ordinary task, and more development is needed.
- Big data can be used for understanding transport patterns and dynamics, and monitoring change at high spatial and temporal resolutions.
- Tools are needed to assist urban and transport planning to incorporate big data analytics into routine procedures.
- Digital planning tools need to be co-designed with end-users for full usability and impact on urban planning and management.
- Partnerships with academia, government and industry have been essential for our research development.

