Mining Google Street View for Urban Climate Micro-Climate Human Thermal Comfort Modelling Parameters

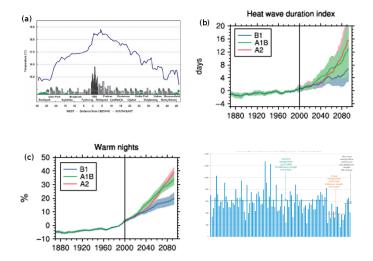
> Kerry A. Nice, Jingcheng Wang, Jasper S. Wijnands, Jason Thompson, Gideon D.P.A. Aschwanden, Mark Stevenson

Transport, Health, and Urban Design Hub, Faculty of Architecture, Building and Planning, University of Melbourne

ICUC10 2018, New York, 9 August 2018

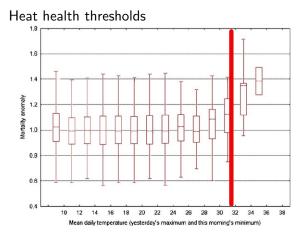


#### Context - Urban heat, climate trends, water supply



Urban heat island effects; predicted increasing extremes for Australia; Melbourne's water supply (Coutts et al., 2010; Alexander and Arblaster, J, 2009; Melbourne Water, 2016)

# Context - Heat / Health Risks

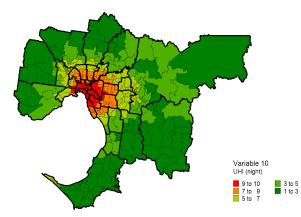


(Nicholls et al., 2008)

#### Low resolution urban heat data

#### How to identify priority areas for heat/health interventions?

Variable 10 Urban heat island



69

#### High resolution thermal imagery

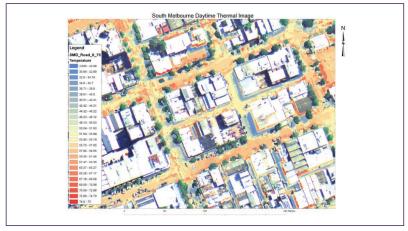


Figure 23: Example of road surface temperatures in South Melbourne during the day.

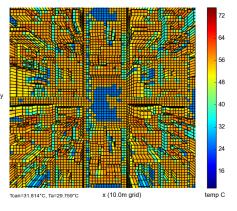
Difficult to obtain at high resolution. Can only make observations of surface temperatures of existing locations at single point in time.

Coutts and Harris (2013)

# VTUF-3D, a tool to model the cooling effects of trees at a microscale



#### LincolnSqRun3-400m-30Days - Tsfc 2014-01-13-1600

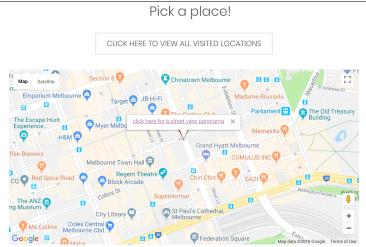


#### Lincoln Square, Melbourne

(Nice et al., 2018)

# Micro-climate modelling simple enough for non-experts





Nice, KA et al. Mining Google Street View

# Micro-climate modelling simple enough for non-experts

The panorama picture



START

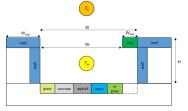
And the modelling begins...

# HTC modelling using TARGET and VTUF-3D

#### Automated generation of domains for TARGET and VTUF-3D.

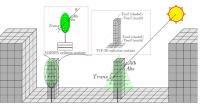
#### The Air-temperature Response to Green/blue-infrastructure Evaluation Tool (TARGET v1.0): an efficient and user-friendly model of city cooling.

Ashley M. Broadbent<sup>1,2,3,4</sup>, Andrew M. Coutts<sup>3,4</sup>, Kerry A. Nice<sup>3,4,5</sup>, Matthias Demuzere<sup>6,7</sup>, E. Scott Krayenhoff<sup>8,1,2</sup>, Nigel J. Tapper<sup>3,4</sup>, and Hendrik Wouters<sup>7,6</sup>



Development of the VTUF-3D v1.0 urban micro-climate model to support assessment of urban vegetation influences on human thermal comfort

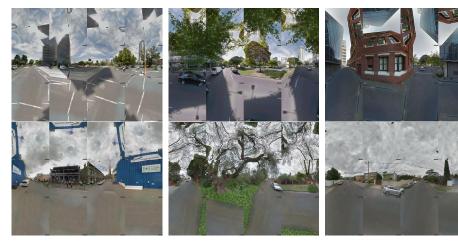
Kerry A. Nice<sup>a,b,c,\*</sup>, Andrew M. Coutts<sup>a,c</sup>, Nigel J. Tapper<sup>a,c</sup>



TARGET available at: (Documents) https://mothlight.github.io/target\_model/ (Python) https://doi.org/10.5281/zenodo.1300023 or (Java) https://zenodo.org/record/1310138

VTUF-3D available at: (Fortran) http://dx.doi.org/10.5281/zenodo.260064 or (Java) https://bitbucket.org/mothlight/vtuf-3d-java.

#### Urban data



Urban imagery is widely available but how to translate to modelling domains? Necessary parameters: sky view factor, surface types and amounts, building locations/heights, vegetation locations/types.

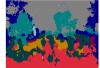
# Classification with convolutional neural network Segnet

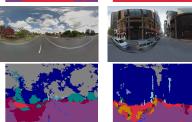
#### Average results, high training and computational costs







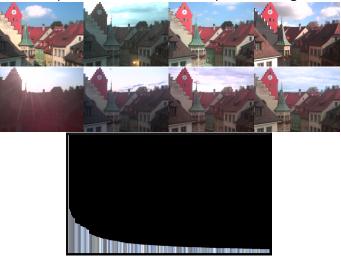




Segnet - Badrinarayanan et al. (2017)

#### Sky view factor - what color is the sky?

Just pick the blue bit at the top of the image?



Color distribution of sky calculated from 8894 daytime images.

(Skyfinder dataset - Mihail et al. (2016))



Approach based on mean shift (Comaniciu and Meer, 2002). Panels arranged to allow sky colors to flow together. Utilize which colors are above and below image center. Preprocessing step to analyse sky colors and amounts to pick mean shift parameters.

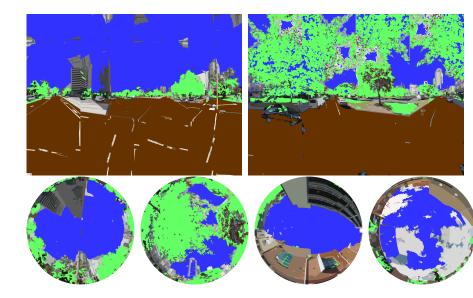
#### Vegetation



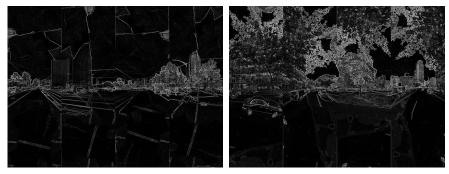
#### Vegetation detection through Otsu thresholding method.

Li et al. (2015)

# Output



#### Edge detection with Sobel filter



Strategies to handle special cases (i.e. high contrast blue/white sky/clouds, sky broken into small pieces by vegetation). Edge detection can be problematic in panoramas as the top of the sky can often contain building/trees, complicating finding a sky border (i.e. Zhijie et al. (2015); Shen and Wang (2013)).

# Hough Line Transform and blurring



Strategies to handle special cases (i.e. high contrast blue/white sky/clouds, sky broken into small pieces by vegetation). Find locations of probable non-sky regions.

Canny threshold 1=50, 2=200, Aperture Size=3, Blur size 10x10.

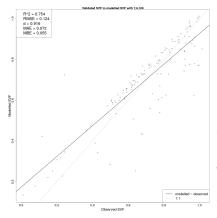
Use preprocessing step (brightness and color distribution) to find and predict special cases and use these techniques instead of mean shift.

#### Validation



Validation of SVF against hand marked images, using a modified Steyn's method (Steyn, 1980), 36 rings.

#### Validation metrics



Mean error	0.072
Mean accuracy	0.915
Mean precision	0.938
Mean recall	0.868
RMSE	0.124

Middel et al. (2018) - Validated against CNN classified 98.5% lateral and 99.6% upward facing images. Results: mean SVF difference of 0.002,  $R^2$ =0.880, and RMSE of 0.045. Gong et al. (2018) - using CNN classification: Verified by observations- $R^2$  >0.95, by GIS  $R^2$ =0.40

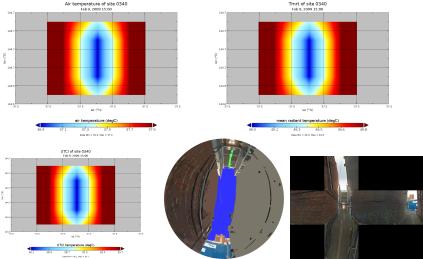
Segnet reports 82-86% global accuracy and 62-81% class accuracy (11 classes) (Badrinarayanan et al., 2017)

# Street view panorama imagery in greater Melbourne



# HTC modelling using detected domains

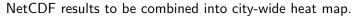
#### Also using satellite imagery and Open Street Map in future versions.

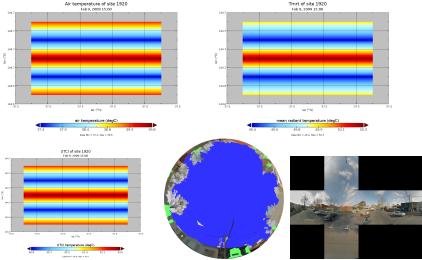


SVF (building and tree locations/heights), green view index, and surface types converted into 25x25m domain and

modelled with TARGET (Broadbent et al., 2018) (and VTUF-3D (Nice et al., 2018))

#### HTC modelling using detected domains





SVF (building and tree locations/heights), green view index, and surface types converted into 25x25m domain and

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Nice, KA et al.	Mining Google Street View
	winning Google Street view

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#### Research team

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https://mothlight.github.io/ 💙 @mothlight

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