

The impact of urban geometry on the radiant environment in outdoor spaces

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A black silhouette of a city skyline is positioned at the bottom of the slide, spanning the entire width. The skyline features various building shapes, including several tall, rectangular skyscrapers and smaller, more varied structures.

Outline of presentation

- Introduction
- Methodology
- Results
- Conclusions

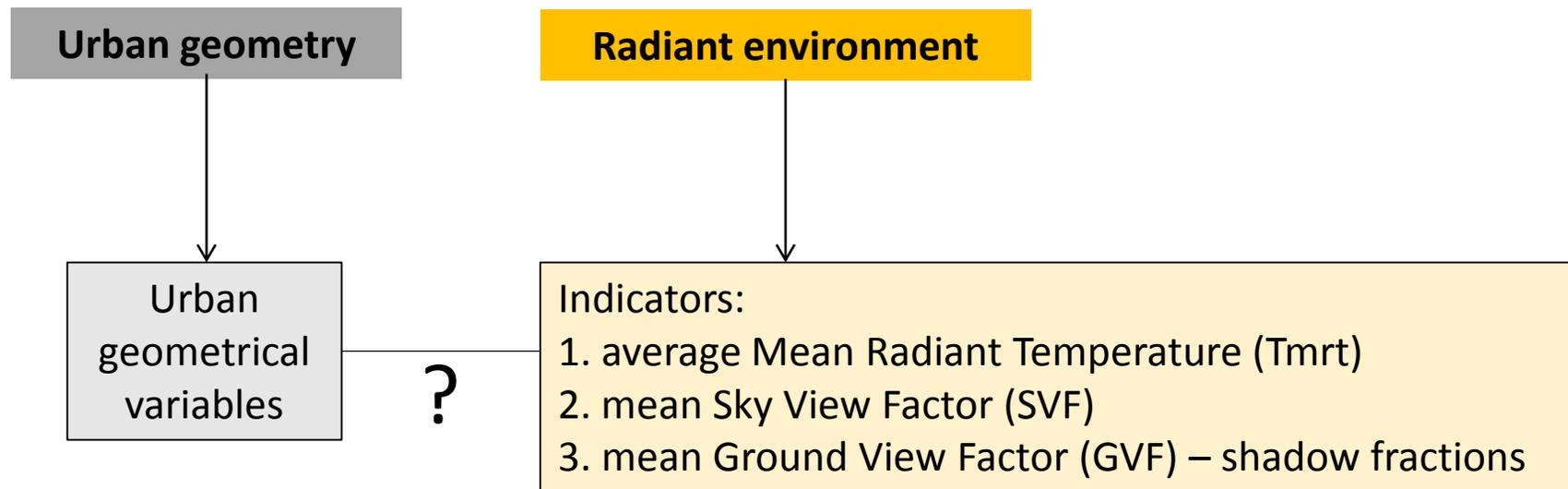
References



Introduction

Research questions

To what extent simple urban geometrical variables can predict the radiant environment?
Which urban variables affect it the most?

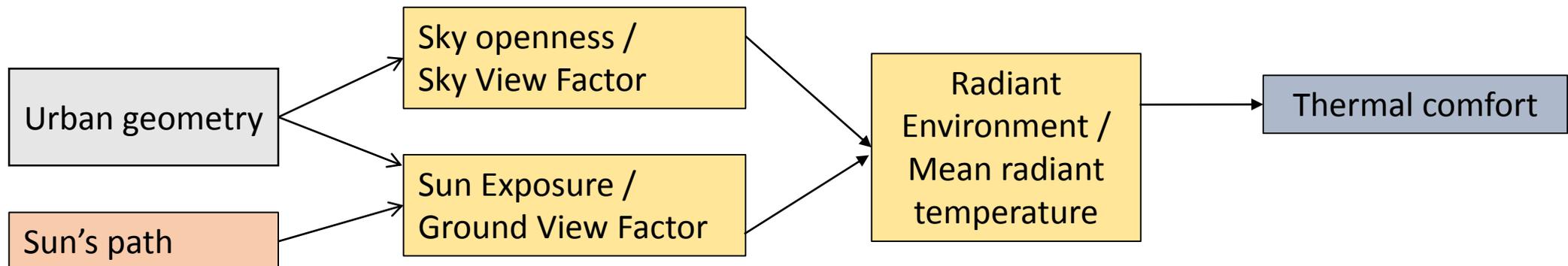


Introduction

Indicators of radiant environment

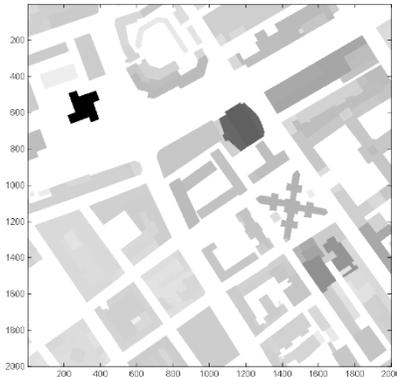
- **Mean radiant temperature** (T_{mrt}), [°C] – is the sum of all radiation fluxes to which a human body is exposed – governs the human thermal comfort
- **Sky View Factor** (SVF) – is a measure of the openness to the sky vault (related to the diffuse solar and sky component). For a given point, its value is constant and ranges between 0 (completely obstructed) and 1.
- **Ground View Factor** (GVF) – is a measure of the exposure to the sun (related to the direct solar component). For a point, its value varies in time and can be either 0 (in shade) or 1 (sunlit).

Diagram of 3 indicators' relation to each other and urban geometry



Introduction

3D urban geometry

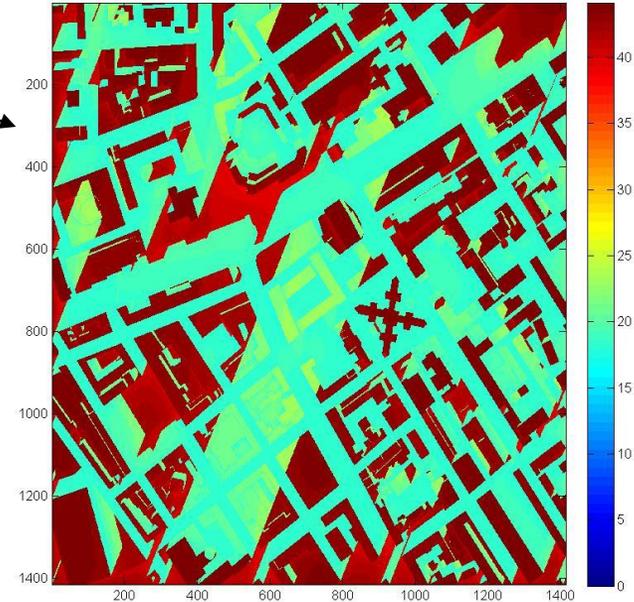


Sun's path

SVF map



GVF map – shadow pattern
(13 p.m. on 23 April)

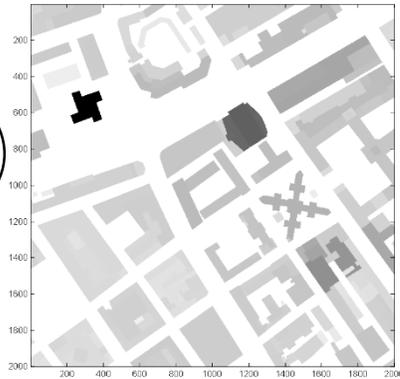


Mean Radiant Temperature map
(13 p.m. on 23 April)

SUNNY CONDITIONS

Introduction

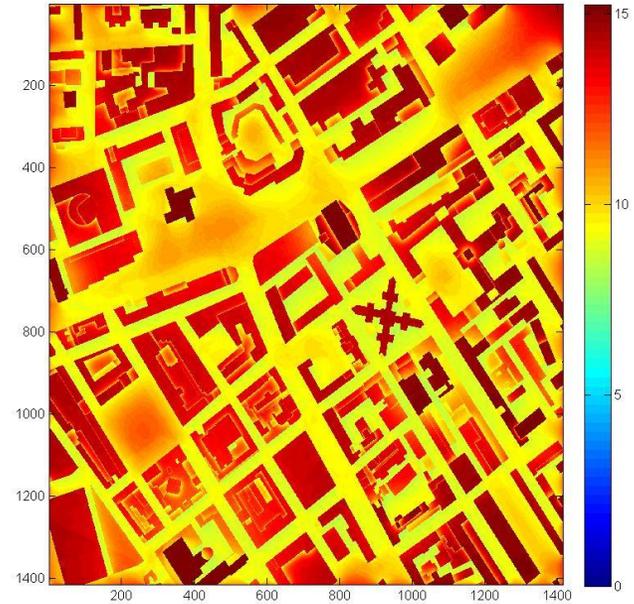
3D urban geometry



SVF map



GVF map – shadow pattern
(9 a.m. on 19 October)



Mean Radiant Temperature map
(9 a.m. on 19 October)

CLOUDY CONDITIONS

Introduction

What is special in this study?

Unique features of the study and relatively new in the literature:

- It uses **real urban forms** (areas of London);
- It has 72 urban forms as case studies which allows for the **statistical exploration of the topic**;
- And, the **spatial scale** (i.e. 500 x 500m) at which the particular topic is studied.

Methodology

3 stages of the study

Three distinct stages:

- (i) the **morphological analysis** using image processing techniques in Matlab software (Ratti and Richens, 2004),
- (ii) **radiation simulations** with the use of SOLWEIG software (Lindberg and Grimmond, 2008), and
- (iii) the **statistical analysis** of the results of the two previous stages investigating potential correlations.

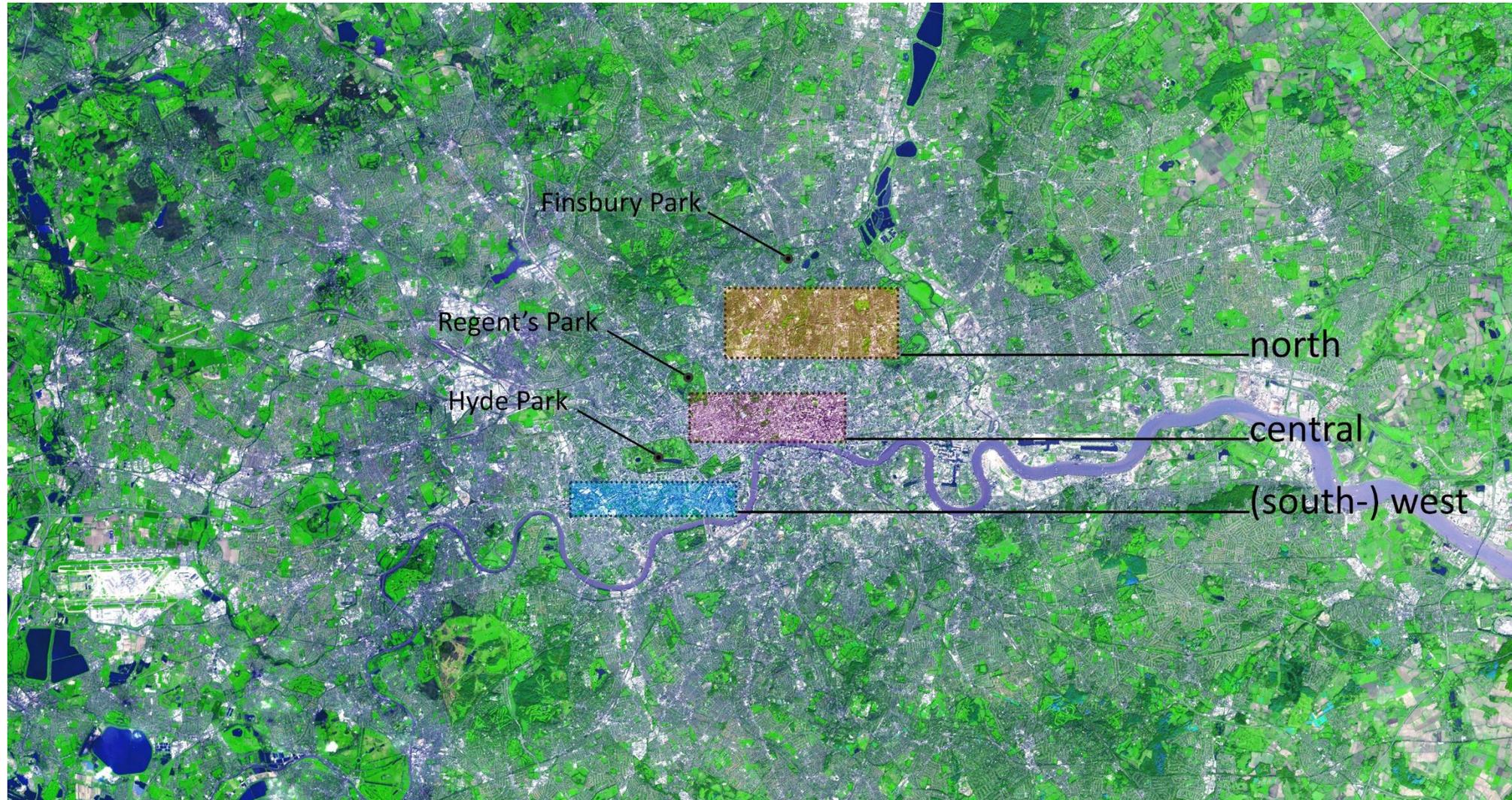
In the morphological analysis and SOLWEIG simulations, urban geometry is represented in DEM format (Digital Elevation Model) which is a compact way of storing urban 3D information using a 2D matrix of building height values.



Example of an urban DEM image

Methodology

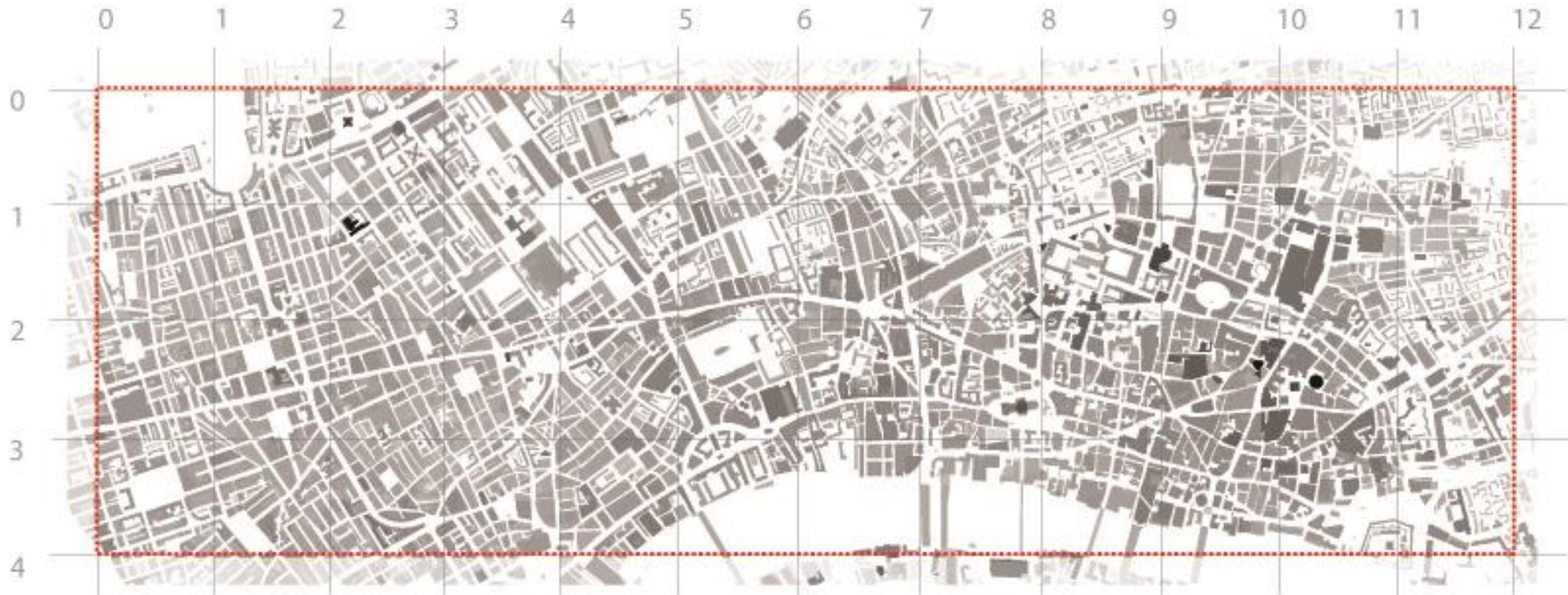
3 areas of London studied



Satellite view of London (Created by NASA, Source: Wikipedia Commons)

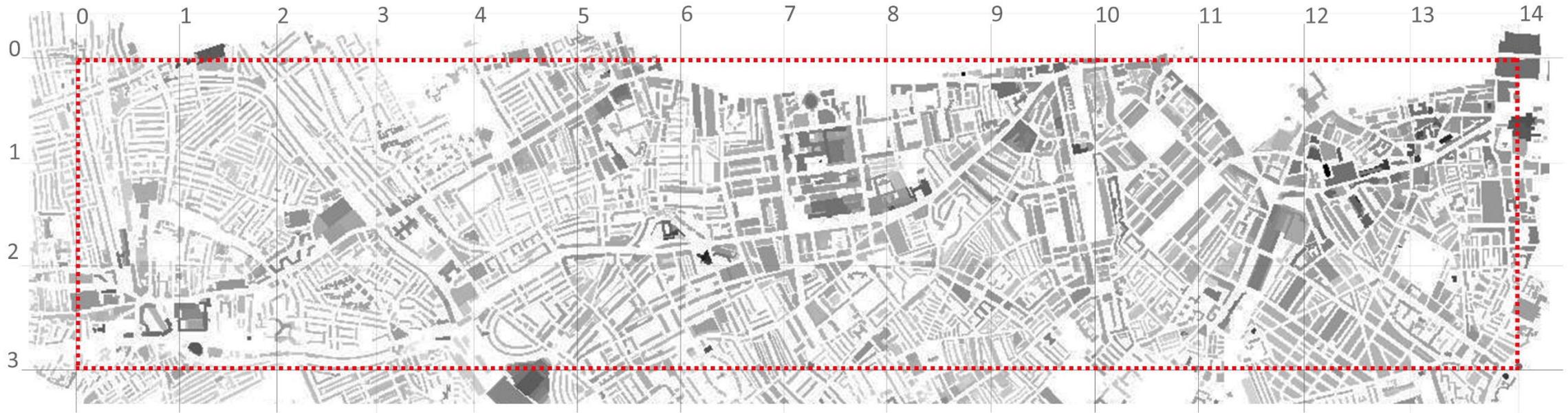
Methodology

DEM of the area in **central London**, divided into squares of 500 x 500m



Methodology

DEM of the area in **west London**, divided into squares of 500 x 500m



Methodology

DEM of the area in **north London**, divided into squares of 500 x 500m



Methodology

Selection of squares

From the studied areas, **72 squares** were selected to be studied:

- 28 squares of central London – high density (values: 9-33 m³/m²)
- 25 squares of west London – medium density (values: 4-14 m³/m²)
- 19 squares of north London – low density (values: 3-6 m³/m²)

Criteria of selection:

- Continuity of urban fabric
- Representativeness of different typologies

Methodology

STAGE ONE_ Morphological analysis

Computation of urban variables:

Urban density

- *Density* - total built volume on a given site over site area [m^3/m^2].

Urban layout descriptors

- *Site coverage* (Coverage) – buildings' footprint area over site area, [m^2/m^2];
- *Frontal area density* (FAD) – buildings' façades area over site area, [m^2/m^2];
- *Number of built volumes* (NoB) – attached buildings considered as one volume;
- *Mean outdoor distance* (mDistance) – mean distance between built volumes on the ground level, [m];
- *Standard deviation of outdoor distance* (sDistance), [m];
- *Standard deviation of building height* (sHeight) – weighted by footprint area, [m];
- *Standard deviation of built volumes* (sVolume), [m^3].

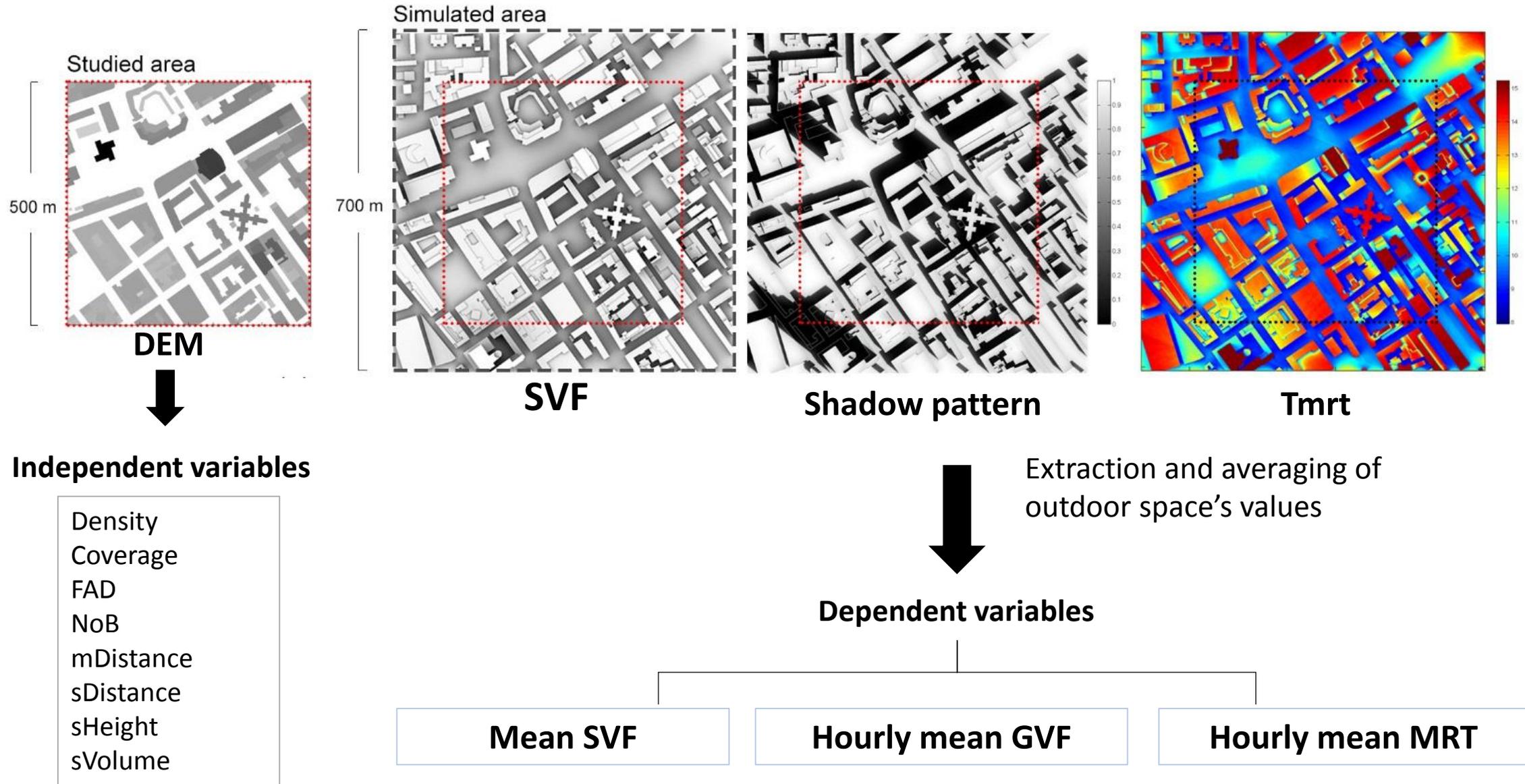
Methodology

STAGE TWO_ Radiation simulations

- The **solar and longwave environmental irradiance geometry** (SOLWEIG) model simulates spatial variations of 3D radiation fluxes (incoming to / outgoing from the ground, direct and reflected) and mean radiant temperature (T_{mrt}) as well as shadow patterns in complex urban settings.
- SOLWEIG's inputs: **3-D urban geometry** (in DEM format), a **24-hour weather file** (temperature, humidity, global, direct and diffuse solar radiation) and **geographical information** (latitude, longitude and elevation of London: 51.5°, -0.17° and 0m).
- Simulations were run for **8 days** (21 Jun, 26 Jul, 23 Apr, 20 Mar, 19 Oct, 23 Nov, 19 Jan and 29 Dec). **Sunny and cloudy days** have been considered, **evenly distributed in the year** in order for the effect of solar elevations to be examined.

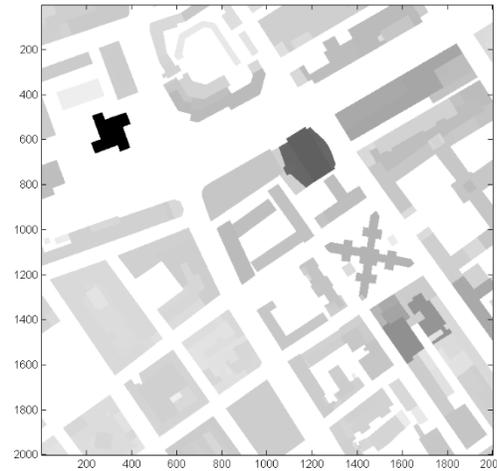
Methodology

STAGE THREE _ Statistical analysis



Results

A. Urban variables – mean SVF



?



Results

A. Urban variables – mean SVF

Pearson Correlation and Linear Regression tests revealed a strong correlation between urban variables and mean SVF values.

- **Density – Mean SVF:**

strong negative correlation >>> $r = -0.94$

linear relationship >>> $R^2 = 0.88$ ($mSVF = 0.685 - 0.017 * \text{density}$)

best curve fit >>> logarithmic, $R^2 = 0.91$

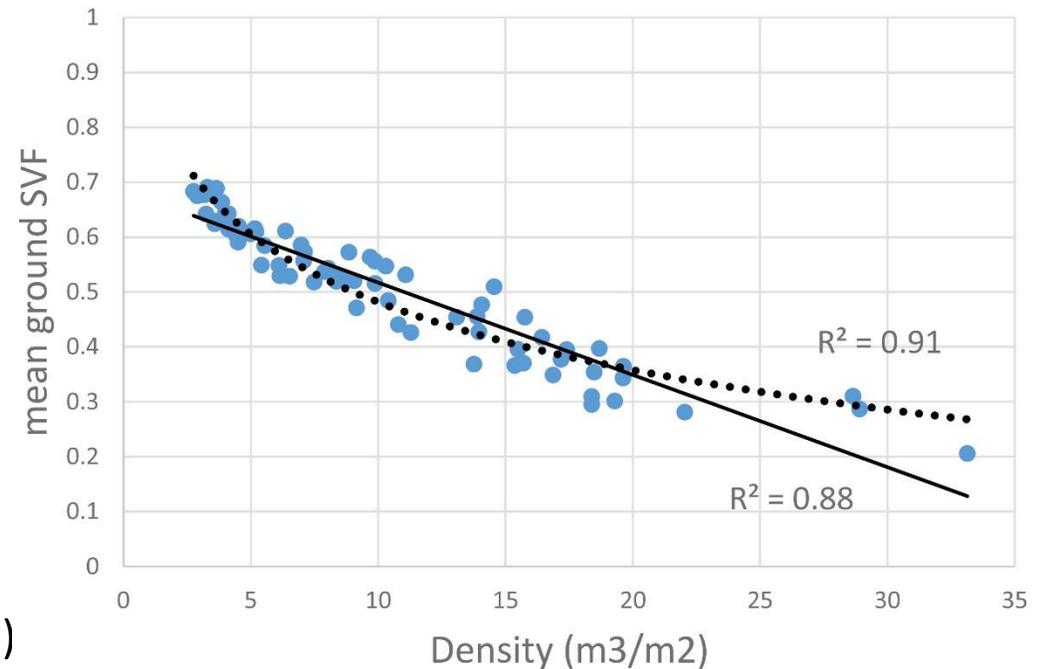
- **7 Urban layout descriptors – Mean SVF:**

strongest variables (partial correlation for density) >>>

Coverage ($r = -0.70$), **mDistance** ($r = 0.53$), **FAD** ($r = -0.47$),

sVolume ($r = 0.37$), **sHeight** ($r = 0.36$), **sDistance** ($r = 0.32$) & **NoB** ($r = 0.29$)

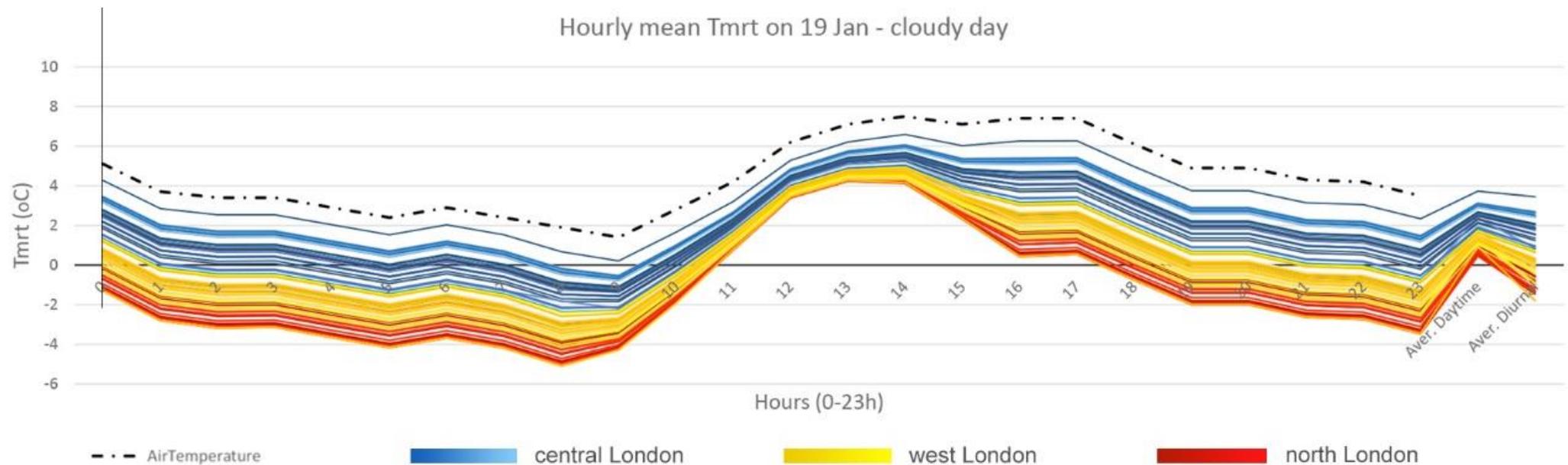
Linear model (Coverage, mDistance, FAD) >>> $R^2 = 0.98$



Results

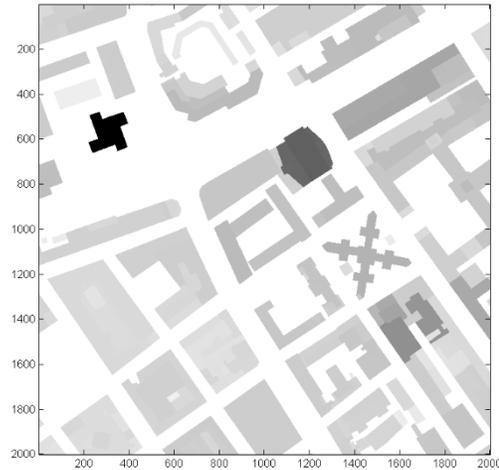
Mean Radiant Temperature (Tmrt) in the absence of direct solar radiation

- The strong relationship between urban variables and mean SVF explains the **high correlation ($R^2 > 0.95$) between urban variables and average Tmrt at night and under cloudy conditions.**
- The outdoor spaces of central London's squares (of higher density) are warmer than those of west and north London, due to greater longwave radiation emitted and reflected by building volumes.



Results

B. Urban variables and mean SVF – Hourly mean GVF



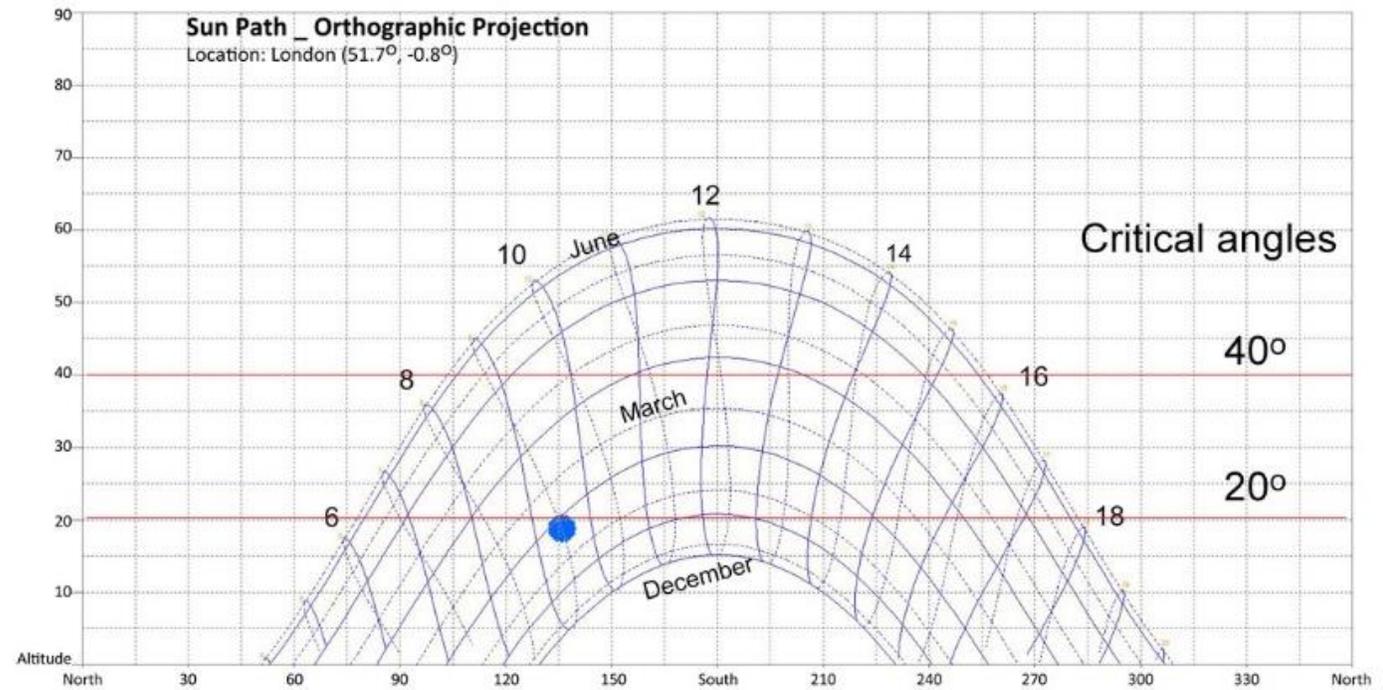
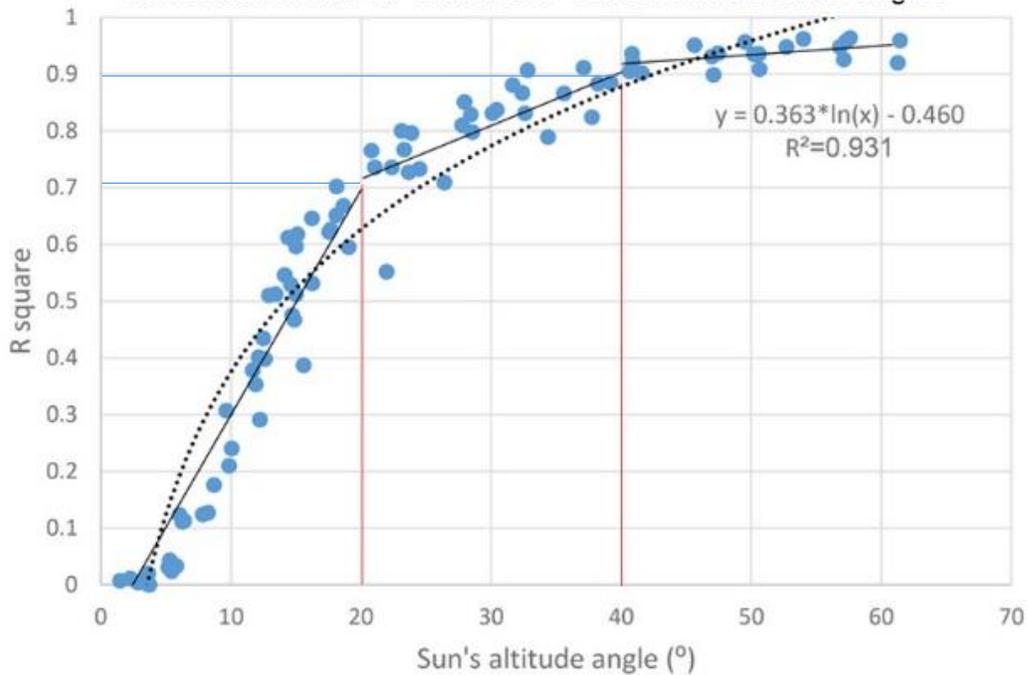
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Results

Mean SVF – Hourly mean GVF

Their correlation (R^2) versus sun's altitude angles



Results

Which urban layout descriptors affect hourly shadow fractions the most?

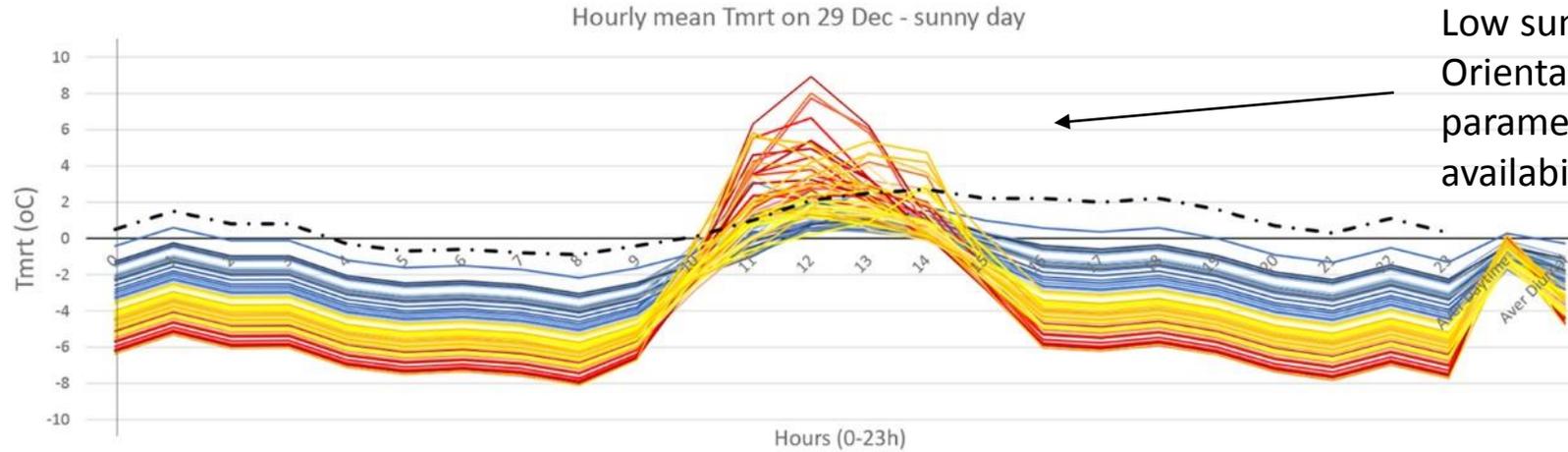
Partial Correlation analysis was performed for the daytime hours of 21st June (summer solstice) controlling density variable.

- 7 am to 9 am & 3pm to 6 pm (morning and late afternoon hours) >>> **Site Coverage** (r= -0.47 to -0.77)
- 10 am to 2 pm (hours close to midday) >>> **mean Distance** or **Frontal Area Density**
- **sDistance**, **sVolume** and **sHeight**, all referring to the homogeneity of the urban form, were also found to be significant in the different hours of the 21st of June

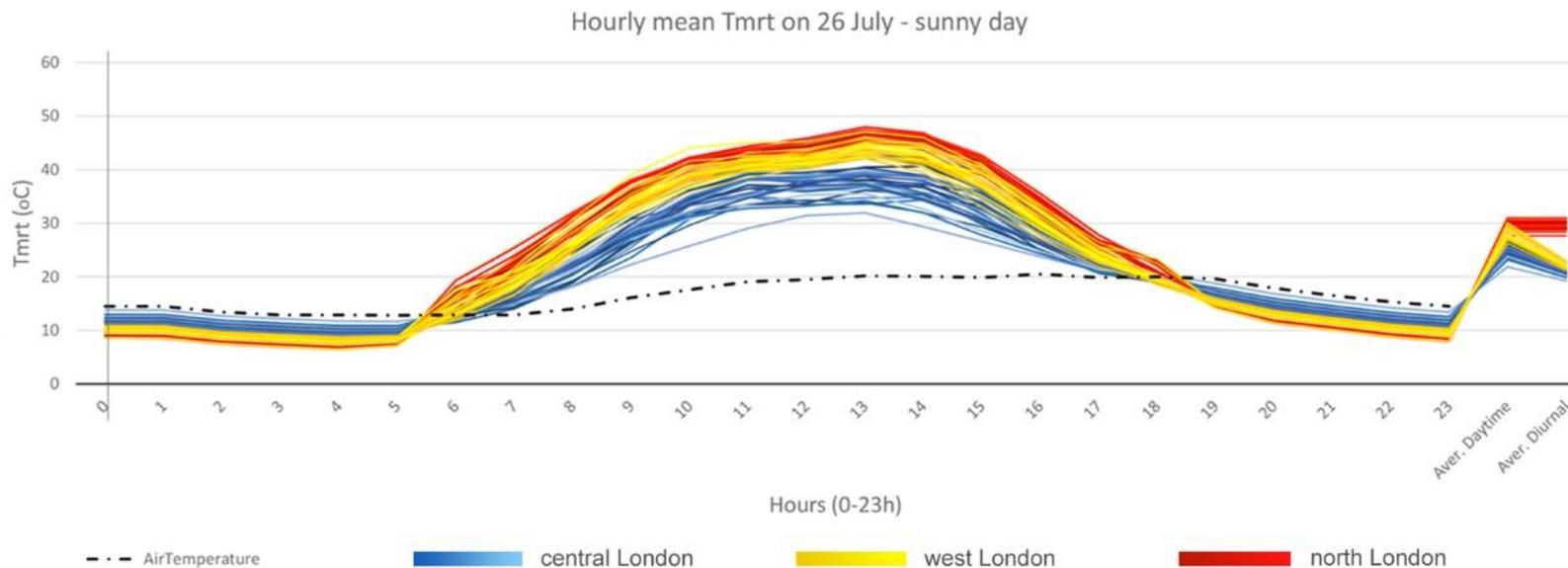
Results

Mean Radiant Temperature (Tmrt) in the presence of direct solar radiation

Sunny winter day
(29 December)



Sunny summer day
(26 July)



Results

Sensitivity of the results to the spatial scale?

Test:

Division of 72 squares of 500 x 500 m into 288 squares of 250 x 250 m size.
 Repetition of statistical tests (for mean SVF).

Result:

The sensitivity of the results to the spatial scale was found to be low.

	500 X 500m	250 x 250 m	Change
Pearson Correlation: Density – mean SVF	r= -0.94	r= - 0.89	- 5.3 %
Partial Correlation: strongest urban variables for mean SVF	Coverage (r= -0.70)	Coverage (r= -0.66)	-5.7%
	mDistance (r= 0.53)	mDistance (r= 0.66)	
	FAD (r= -0.47)	FAD (r= -0.52)	

Conclusions

To what extent simple urban geometrical variables can predict the radiant environment?

Overall, the radiant environment at the district level can be predicted to a great degree by urban variables and thus, is amenable to being modified through urban planning.

- **Absence of solar radiation:**

The correlation with mean outdoor SVF value is significantly strong. As a consequence, the urban variables can predict the average outdoor T_{mrt} in the night time and under cloudy conditions.

- **Presence of solar radiation:**

The correlation with the hourly mean GVF (hourly shadow fractions) and therefore, with average T_{mrt} in sunny conditions, varies with sun's altitude angle, and can be significantly strong for a specific range of angles.

The critical angles identified are essentially independent of location; thus, they could be advised for other parts of the world.

Conclusions

Which urban variables affect the radiation availability in outdoor spaces the most?

- **Site coverage** was found to be the strongest variable followed by **mean distance between building volumes** and **frontal area density**. As regards site coverage, in agreement with Lindberg and Grimmond (2011).
- Urban layout descriptors capturing **the randomness of urban layout** also present a significant positive correlation. In general agreement with those studies (e.g. Cheng et al., 2006) arguing that increasing urban randomness enhances the solar availability in the urban fabric.

References

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The End

Thank you for your attention!

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A black silhouette of a city skyline is positioned at the bottom of the slide, featuring various building shapes and heights.

Appendix

SOLWEIG 2013a

File Help

STEP 1 Load DSMs => STEP 2 Specify Output => STEP 3 Load/Create SVF => **STEP 4 Load Met. data** => STEP 5 Execute

Digital Surface Model Sky View Factor

Geographical location

Sweden, Göteborg

Longitude (decimal degree): 11.57

Latitude (decimal degree): 57.7

Altitude (masl): 3

UTC (h): 1

Add new location

Personal parameters

Absorption (shortwave) 0.70

Absorption (longwave) 0.95

Posture Standing

Urban parameters

Albedo 0.20

Emissivity (walls) 0.90

Emissivity (ground) 0.95

Transmissivity (vegetation) 0.02

Daily shading

Set date 13/02/2014

Time interval: 30 minutes

Calculate Daily Shading

Optional settings

Do not use vegetation

Show hourly images of Tmrt during execution

Set point of Interest

Appendix

		Correlations								
		Density	Coverage	sHeight	FAD	NoB	mDistance	sDistance	sFootprint	sVolume
Density	Pearson Correlation	1								
	Sig. (2-tailed)									
Coverage	Pearson Correlation	.901**	1							
	Sig. (2-tailed)	.000								
sHeight	Pearson Correlation	.805**	.582**	1						
	Sig. (2-tailed)	.000	.000							
FAD	Pearson Correlation	.944**	.830**	.794**	1					
	Sig. (2-tailed)	.000	.000	.000						
NoB	Pearson Correlation	-.757**	-.829**	-.531**	-.691**	1				
	Sig. (2-tailed)	.000	.000	.000	.000					
mDistance	Pearson Correlation	-.306**	-.389**	-.208	-.364**	-.019	1			
	Sig. (2-tailed)	.009	.001	.079	.002	.876				
sDistance	Pearson Correlation	-.070	-.063	-.135	-.144	-.283*	.905**	1		
	Sig. (2-tailed)	.559	.600	.257	.227	.076	.000			
sFootprint	Pearson Correlation	.790**	.839**	.504**	.639**	-.762**	-.155	.079	1	
	Sig. (2-tailed)	.000	.000	.000	.000	.000	.193	.507		
sVolume	Pearson Correlation	.715**	.633**	.584**	.524**	-.569**	-.087	.038	.865**	1
	Sig. (2-tailed)	.000	.000	.000	.000	.000	.466	.752	.000	

** . Correlation is significant at the 0.01 level (2-tailed).

* . Correlation is significant at the 0.05 level (2-tailed).