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2	Development of an Improved Urban Emissivity model based on Sky View Factor
3	for retrieving effective emissivity and surface temperature over urban areas
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22 Abstract

This study aims to evaluate the effects of urban geometry on retrieval of emissivity 23 and surface temperature in urban areas. An improved urban emissivity model based on 24 sky view factor (IUEM-SVF) was further enhanced to consider all radiance 25 contributions leaving the urban canopy, including (i) emission by all facets within an 26 instantaneous field of view (IFOV); (ii) reflection by all facets of emission from 27 surrounding facets; and (iii) propagation of emitted and reflected radiation with 28 multiple reflections (scattering) within a complex 3D array of urban objects. The 29 30 effective emissivity derived from IUEM-SVF was evaluated with a microscale radiative transfer and energy balance model: Temperatures of Urban Facets in 3-D (TUF-3D). 31 IUEM-SVF performs well when urban facets have uniform emissivity and temperature; 32 33 e.g., root mean square deviations (RMSD) are less than 0.005 when material emissivity is larger than 0.80 ($\varepsilon \ge 0.80$). However, when material emissivities are variable within 34 the observed target, differences of effective emissivity between IUEM-SVF and TUF-35 36 3D become larger, e.g., RMSD of 0.010. When the effect of geometry is not considered and a mixed pixel emissivity is defined, the difference is even much larger (i.e. 0.02) 37 and this difference increases with the decrease of sky view factor. Thus, the geometry 38 effect must be considered in the determination of effective emissivity. Effective 39 emissivity derived from IUEM-SVF is used to retrieve urban surface temperature from 40 a nighttime ASTER thermal infrared image. Promising results are achieved in 41 comparison with standard LST products retrieved with the Temperature and Emissivity 42 Separation (TES) algorithm. IUEM-SVF shows promise as a means to improve the 43

44	accuracy of urban surface temperature retrieval. The effect of thermal heterogeneity on
45	the effective emissivity was also evaluated by TUF-3D, and results show that the
46	thermal heterogeneity cannot be neglected since the RMSD between the effective
47	emissivity based on TUF-3D and IUEM-SVF is relatively large.
48	

49 Keywords: emissivity; temperature, geometry; thermal image

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51	1.	Introduction

Land Surface Temperature (LST) obtained by remotely sensed data provides a 52 synoptic view and an effective measure of the Surface Urban Heat Island (SUHI), by 53 observing the temperature differences between urban surfaces and rural areas (Dousset 54 55 and Gourmelon, 2003; Voogt and Oke, 2003; Hu and Brunsell, 2013). For a flat and homogenous surface, remote sensors obtain surface exitance and emittance, after 56 correction for atmospheric effects and taking into account the reflection of atmospheric 57 58 emission. Surface temperature can be retrieved from the material emissivity and surface emittance through the Planck function (Sobrino et al., 2004; Sobrino et al., 2012). 59 60 However, the geometry of urban areas is always complex. This results in anisotropy of satellite-observed surface emittance, due to the complexity of urban geometry and the 61 different component temperatures within mixed pixels. The effect of different 62 component temperatures in a mixed pixel over rural areas has been well-studied 63 (Sobrino and Caselles, 1990; Menenti et al., 2001; Jia et al., 2003; Chen et al., 2004; 64 Rasmussen et al., 2011; Ren et al., 2011; Guillevic et al., 2013; Ren et al., 2014; Cao et 65

66	al., 2015). Other studies of directional anisotropy of surface temperature have shown
67	that geometric effects should be considered in the retrieval of surface temperature over
68	urban areas (Voogt and Oke, 1998; Lagouarde et al., 2004; Soux et al., 2004; Lagouarde
69	and Irvine, 2008; Voogt, 2008; Lagouarde et al., 2010; Lagouarde et al., 2012; Duffour
70	et al., 2016; Krayenhoff and Voogt, 2016). The aforementioned studies have examined
71	different component effects on the radiance measured within the IFOV, but have not
72	studied the impacts of the complex building geometry and thermal heterogeneity on the
73	retrieval of effective emissivity and land surface temperature in urban environments.
74	A few terms of thermal infrared radiometry are first defined in this study: (i)
75	emittance, (ii) exitance, and (iii) effective emissivity (Becker and Li, 1995; Norman
76	and Becker, 1995; Norman et al., 1995). The emittance is the radiance emitted by all
77	horizontal surfaces and vertical facets directly. The exitance is the sum of the emittance,
78	the radiance re-emitted after absorption and radiance reflected by the surfaces and
79	building facets. The effective emissivity is defined as the ratio of the total exitance of a
80	pixel to the emittance of a blackbody at the same temperature (Yang et al., 2015a). Due
81	to the multiple scattering and reflection caused by buildings in urban environments, the
82	exitance in built-up areas is different from surface emittance. Multiple scattering and
83	reflection also increase the radiance absorbed and emitted by the 3D-surface within a
84	satellite pixel (Yang et al., 2015a). The total exitance of built-up areas is thus higher
85	than that of a flat surface with the same material and temperature. The term "exitance"
86	is thus more appropriate than emittance and it refers to the radiance leaving the urban
87	canopy. For satellite data with a constant IFOV or footprint, the thermal sensor

measures the exitance and not only the emittance. Thus, it is not always accurate to use 88 a material emissivity to retrieve the surface temperature directly from the exitance, and 89 90 an effective emissivity should be derived. If the surface is flat, the exitance is similar to the surface emittance, which can be derived from material emissivity and the Planck 91 function, with a small contribution owing to reflection of atmospheric emittance. The 92 exitance at the top of the urban canopy includes both the radiance emitted and that 93 reflected by each facet (e.g. horizontal and vertical) of the 3D environment. Thus, the 94 exitance can be determined by the three dimensional urban geometry, component 95 96 temperature and material emissivity of surface components. The relationship between effective emissivity and geometry effect has been discussed in several studies 97 (Sutherland and Bartholic, 1977; Harman et al., 2004; Danilina et al., 2012; Yang et al., 98 99 2015a).

In order to improve the accuracy of retrieval of urban surface temperature from the 100 average exitance measured by space- or airborne-sensors over a footprint (i.e., from the 101 IFOV), Yang et al (2015a) parameterized the urban effective emissivity using the sky 102 view factor (UEM-SVF). This model considers the additional radiance caused by the 103 cavity effect and the scattered radiance from neighboring pixels and the atmosphere, 104 however it does not include radiance scattered by the surface elements within the pixel 105 directly reflected to the sky. In this study, the UEM-SVF method was further improved 106 by considering all radiance contributions leaving the urban canopy: (a) emission by all 107 facets within an IFOV; (b) reflection by all facets of radiation emitted by the 108 surroundings; (c) propagation of emitted and reflected radiation with multiple 109

reflections (scattering) within the complex 3D array of urban objects. Thus the 110 improved UEM-SVF, hereafter "IUEM-SVF", was to calculate the cavity effect on 111 emission and reflection separately, while the UEM-SVF in Yang et al. (2015) only 112 accounts for the cavity effect on emission. In addition, the IUEM-SVF was evaluated 113 with a micro-scale three-dimensional (3D) radiation exchange and urban energy 114 balance model, Temperatures of Urban Facets in 3D (TUF-3D) (Krayenhoff and Voogt, 115 2007). TUF-3D model can be applied to calculate an effective emissivity, which takes 116 all radiative interactions within a scene into account during calculation at a sub-facet 117 118 resolution specified by the user (Krayenhoff and Voogt, 2007). Subsequently, TUF-3D was used to evaluate the effect of thermal heterogeneity on urban effective emissivity 119 and surface temperature retrieval. 120

121 Additionally, the IUEM-SVF was used to retrieve urban surface radiometric temperatures from the ASTER Band 13 data on December 16th, 2015. These 122 temperatures were subsequently compared with the surface temperature product 123 124 ASTER 2B03 derived from Temperature and Emissivity Separation (TES) algorithm (Gillespie et al., 1998). The ASTER TES algorithm is based on an empirical 125 relationship between the minimum spectral emissivity (ε_{min}) and Minimum-Maximum 126 Difference (MMD), established by analyzing 86 laboratory TIR emissivity spectra data 127 (Gillespie et al., 1998). Oltra et al. (2014) analyzed the performance of the ASTER TES 128 algorithm based on simulated data and the ε_{min} -MMD relationship derived from 129 130 material emissivity over urban areas and results showed that the retrieved emissivity was higher than the material emissivity (Oltra et al., 2014). This is because the radiance 131

obtained by remote sensors is affected by multiple reflections caused by geometry effect over urban areas and the actual spectral radiance observed over urban areas does not follow the empirical relationship ε_{min} -*MMD*. The limitation of the application of TES in urban surface temperature was also analyzed in this study.

136

137 **2. Data**

Kowloon Peninsula of Hong Kong was selected as the study area for this research 138 (Figure 1). One set of airborne high resolution thermal images acquired at 12:57 (Hong 139 Kong noon time) on August 6th, 2013 was used to estimate differences in component 140 temperatures. The component temperatures were derived by sampling the image data. 141 The material of buildings was taken as concrete and cement. The mean rooftop 142 143 temperature (Tr) is 328.56 K, the mean street temperature (Ts) is 316.64 K, and mean wall temperature (Tw) is 304.94 K, respectively, in a built-up area of Kowloon 144 peninsula, at 22.85 °N, 114.08 °E. 145

The ASTER thermal radiance data on December 16th, 2015 14:36 (UTC) (Hong Kong local time is 10:36 pm) were used to retrieve urban surface temperature using the single channel method since the urban radiative transfer model can be applied in urban surface temperature retrieval using single channel method, while the ASTER 2B01 LST product was used to evaluate the derived LST from our improved effective emissivity model and single channel method. ASTER Band 13 was selected for LST retrieval since band 10 and 14 which are close to the edge of atmospheric window, are more easily affected by atmospheric effects. In addition, a declining of sensitivity of band 12 hasalso been reported (Nichol et al., 2009).

High spatial resolution airborne Lidar data (Lai et al., 2012) and building GIS data 155 covering the entire Hong Kong territories were acquired for calculating SVF. The 156 LiDAR data and 3D building data are provided by the Hong Kong Lands Department 157 and the Hong Kong Civil Engineering and Development Department. The method of 158 calculating SVF on vertical facets is based on the method presented by Kanda et al. 159 (2005), Zakšek et al. (2011) and Yang et al. (2015). Land use and land cover data e.g. 160 161 woodland, grassland and impervious surface with 6 m resolution were also used. Land use and land cover data are provided by the Hong Kong Planning Department. The 162 building GIS data were used to distinguish the impervious surface on buildings and 163 164 road pavements. More details about the land use and land cover data, building GIS data, as well as the airborne Lidar data can be found in Yang et al. (2015b). The spectral 165 Library of Impervious Urban Materials (Kotthaus et al., 2014) and spectral response 166 167 function of ASTER band 13 were used to derive the emissivities of building and impervious surface, where ASTER Spectral Library 2.0 (Baldridge et al., 2009) and 168 spectral response function of ASTER band 13 were used to derive the emissivities of 169 trees and grass. The spectral emissivity of wall facets, rooftop, road pavement, tree, and 170 grass were 0.886, 0.945, 0.948, 0.973, and 0.986 respectively in ASTER band 13. More 171 information on the spectral emissivity of ASTER band 13 can be found in Yang et al. 172 173 (2015a).

174

"Insert Figure 1 here" 175 176 3. Method 177 3.1 IUEM-SVF model 178 Assuming the radiometric data are acquired by a satellite sensor at relatively low 179 spatial resolution and at only one view angle, the component temperatures within a 180 pixel cannot be easily observed. The IUEM-SVF model assumes that the pixel is 181 isothermal, thus the effect of thermal heterogeneity on radiative transfer in mixed pixels 182 183 is not accounted for. If the geometric effect is considered and the thermal heterogeneity effect is not accounted for, the exitance of pixel i with the cavity effect on emission 184

185 can be defined as $\varepsilon(i)' B(T_s(i))$ and $\varepsilon(i)'$ can be formulated as (Yang et al., 2015a):

186

187
$$\varepsilon(i)' = \varepsilon(i)/[1 - (1 - \varepsilon(i))(1 - v_{sky}(i))]$$
 (1)

188

where $T_s(i)$ is the surface radiometric temperature of pixel *i* for the isothermal 189 pixel, B is the Planck function. $\varepsilon(i)'$ is the effective emissivity due to the cavity 190 effect within the pixel i. $\varepsilon(i)$ is the material emissivity of the pixel i and can be 191 calculated as the area-weighted average of the components in a mixed pixel using very 192 high resolution land use data and building GIS data. $v_{sky}(i)$ is the Sky View Factor 193 (SVF) of pixel *i*. This equation accounts for the increases in emission due to the cavity 194 effect in pixel *i*. The SVF is defined as the fraction of the hemisphere occupied by the 195 sky observed at a certain location (Zakšek et al., 2011). The SVF was determined from 196

a Digital Surface Model (DSM) along with a search radius, and SVF values determined
in different search directions were averaged over a single pixel.

The exitance is the fraction of these radiation components (a, b, c in Equation 2) reflected to the sky: (a) radiance emitted within the pixel but blocked by wall facets of buildings within a pixel; (b) radiance from sky; (c) radiance from adjacent pixels. $v_{sky}(i)(1 - \varepsilon(i))$ is the fraction of all radiance reflected to sky. Then the mean radiance, $Ref_1(i)$, received by pixel *i* and then reflected to sky after the first reflection, can be calculated as:

205

206
$$Ref_{1}(i) = v_{sky}(i)(1 - \varepsilon(i))\{(1 - v_{sky}(i))\varepsilon(i)B(T_{s}(i)) + v_{sky}(i)*R_{at}^{\downarrow}(i) + v_{sky}(i)*R_{adj}^{\downarrow}\}$$
207
$$v_{adj}(i)*R_{adj}\}$$
(a) (b) (2)
208 (c)

209 $R_{at}^{\downarrow}(i)$ is the atmospheric downward radiation (assumed isotropic), R_{adj} is the 210 mean radiance of the adjacent pixels. $v_{adj}(i)$ is the view factor from the pixels 211 surrounding pixel *i* (Yang et al., 2015b).

After a second reflection within the pixel *i*, the radiance reflected to sky can be written as:

214
$$Ref_{2}(i) = v_{sky}(i)(1 - \varepsilon(i))(1 - v_{sky}(i))(1 - \varepsilon(i)) * \{(1 - v_{sky}(i))\varepsilon(i)B(T_{s}(i)) + v_{sky}(i) * R_{at}^{\downarrow}(i) + v_{adj}(i) * R_{adj}\} = (1 - \varepsilon(i))(1 - v_{sky}(i))Ref_{1}(i)$$
216
$$v_{sky}(i)Ref_{1}(i)$$
(3)

217

After n^{th} reflection within the pixel *i*, the radiance reflected to sky can be

219 formulated as:

221
$$Ref_n(i) = (1 - v_{sky}(i))(1 - \varepsilon(i))Ref_{n-1}(i)$$
 (4)

224
$$\operatorname{Ref}(i) = \operatorname{Ref}_1(i) + a * \operatorname{Ref}_1(i) + \dots + a^{n-1} * \operatorname{Ref}_1(i) = \frac{\operatorname{Ref}_1(i)\{1 - a^{m-1}\}}{1 - a}$$
 (5)

226 where
$$a = (1 - v_{sky}(i))(1 - \varepsilon(i))$$

228 When m tends to infinity, a^{m-1} tends to 0, thus Ref(*i*) can be rewritten as:

230
$$\operatorname{Ref}(i) = \frac{\operatorname{Ref}_1(i)}{1-a}$$
 (6)

And combining equation 2 with equation 6, the equation is:

234
$$\operatorname{Ref}(i) = \frac{v_{sky}(i)(1-\varepsilon(i))}{1-a} \{ (1-v_{sky}(i))\varepsilon(i)B(T_s(i)) + v_{sky}(i) * R_{at}^{\downarrow}(i) + v_{adj}(i) * R_{adj}^{\downarrow} \}$$
236
$$(7)$$

where $v_{sky}(i)(1-\varepsilon(i))/(1-a)$ on the right side of Equation 7 is equal to $(1-\varepsilon(i))$. Then the total exitance on the urban canopy can be defined as:

241
$$E(i) = \varepsilon(i)'B(T_s(i)) + \operatorname{Ref}(i) = \varepsilon(i)'B(T_s(i)) + (1 - \varepsilon(i)')\{(1 - v_{sky}(i))\varepsilon(i)B(T_s(i)) + v_{sky}(i) * R_{at}^{\downarrow}(i) + v_{adj}(i) * R_{adj}$$
(8)

243

This is the exitance, i.e. the radiance leaving the urban canopy, which is observed by remote sensors. $(1 - \varepsilon(i)')\{(1 - v_{sky}(i))\varepsilon(i)B(T_s(i)) + v_{sky}(i) * R_{at}^{\downarrow}(i) + v_{adj}(i) * R_{adj}\}$ in Equation (8) is the fraction of atmospheric emittance and radiance, reflected by adjacent pixel and the sub-surface within the pixel *i*. $\varepsilon(i)'$ is the effective emissivity of the surface within a pixel and $(1 - \varepsilon(i)')$ is the effective reflectance of the pixel when the transmittance is 0.

In order to estimate $\varepsilon(i)'$ using TUF-3D, the atmopheric emittance was defined as 0 and the adjacent pixel effect was also neglected. This allows the ratio of the total exitance to the radiance of the blackbody at the same temperature to be written as:

253

254
$$\epsilon(i)'' = \epsilon(i)' + (1 - \epsilon(i)')(1 - v_{sky}(i))\epsilon(i)$$
 (9)

The first term in Equation (9), $\varepsilon(i)'$, is effective emissivity of the pixel *i*, which 255 considers the additional emitted radiance because of the re-absorption caused by 256 blockage of sub-surface within pixel i (Yang et al., 2015a). The term (1 -257 $v_{sky}(i)$ (i) is the fraction of blocked radiance by the urban structures within pixel i 258 and $(1 - \varepsilon(i)')$ is the effective reflectance of pixel i to the sky, thus the term in 259 Equation (9), $(1 - \varepsilon(i)')(1 - v_{sky}(i))\varepsilon(i)$, is the fraction of the radiance emitted by 260 facets within pixel i onto other facets within pixel i and from there reflected back to 261 the sky. This is the IUEM-SVF model of urban effective emissivity under the condition 262

that the atmospheric downward radiance and adjacent pixel effect are removed from 263 the total exitance. When SVF tends to 0 under the theoretical condition, the reflection 264 tends to 0 and all radiances are absorbed, i.e. $\varepsilon(i)'$ tends to 1, similar to an "effective" 265 blackbody. When the SVF tends to 1, the surface emits to the sky and no radiance is 266 intercepted and re-absorbed, thus $\varepsilon(i)'$ is the material emissivity and the reflection 267 among facets within a pixel tends to 0. If material emissivity of sub-surface within the 268 pixel is the same, effective emissivity increases with decreasing SVF. For the real urban 269 areas of Hong Kong, the values of SVF between 0.22 and 0.91 located between the 2.5% 270 271 and 97.5 % under the normal distribution curve.

272

273 *3.2 TUF-3D model*

274 The TUF-3D model (Krayenhoff and Voogt, 2007) was used to evaluate the results from the IUEM-SVF model and to analyze the relationship between exitance and 275 surface temperature. The radiation model of the TUF-3D uses the radiosity approach 276 277 with analytical view factors and accounts for all (effectively infinite) radiation reflections in a scene (Krayenhoff and Voogt, 2007). TUF-3D calculates the upward 278 radiation (radiation to sky in the scene) and the radiation balance of each facet (e.g. 279 radiation balances of wall, street and rooftop) considering the effect of geometry. More 280 281 detail about TUF-3D can be found in Krayenhoff and Voogt (2007).

282

283 *3.3 Comparison between IUEM-SVF and TUF-3D*

284 The TUF-3D model is based on the Stefan-Boltzmann law and broadband

285	emissivity. However, remote sensing applications require narrowband spectral
286	emissivity to retrieve the land surface temperature and the geometry effect should be
287	considered. The radiative transfers of narrowband and broadband longwave radiation
288	in urban areas are the same, thus the impact of urban geometry on effective emissivity
289	can be evaluated using TUF-3D by specifying the narrowband material emissivities as
290	the inputs. Additionally, the optimized version of TUF-3D used here only represents
291	relatively simple urban covers with urban geometry characterized by a constant ratio of
292	building height to building spacing. For each experiment, the ratio of building height to
293	spacing was varied between 0.5 and 4, at an interval of 0.5, i.e. eight urban geometry
294	cases were considered. The SVF values ranged from 0.59 to 0.86 across the cases.
295	Experiments to evaluate IUEM-SVF effective emissivity with TUF-3D are

designed as follows and the summary of the settings for TUF-3D can be referred Table1:

a) In order to evaluate the geometry effect with material homogeneity and isothermal
sub-surfaces, uniform surface temperature and material emissivities for all facets
are defined in TUF-3D. The surface temperature was set as 303.15 K. Four values
of emissivity were used, e.g. 0.8, 0.85, 0.90, 0.95 for uniform material emissivity
for each numerical experiments (Case a1, a2, a3 and a4 in Table 1). Geometry is the
only variable parameter.

b) Effective emissivity is usually calculated as the area-weighted average of the
 component emissivity without considering the effect of geometry or spatial thermal
 heterogeneity in the application over urban areas (Case b in Table 1). These

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308

estimates were then evaluated with TUF-3D, assuming the same temperature for all components.

c) The IUEM-SVF, which accounts for geometry and the area-weighted average of 309 component emissivities, was used to estimate the effective emissivity in surface 310 temperature retrieval from satellite remote sensing data. The configuration applied 311 in TUF-3D is different to previous approaches, i.e. the material emissivity is 312 specified for each facet, but all facets are assumed to have the same temperature to 313 evaluate the effect of the material heterogeneity (Case c). The comparison between 314 315 IUEM-SVF and TUF-3D was designed to evaluate whether it is necessary to consider the actual 3D distribution of surface emissivity. IUEM-SVF does not 316 account for thermal heterogeneity of mixed pixels. The area-weighted average of 317 318 the component emissivities in a mixed pixel was applied. The SVF is applied to account for the effects of geometry. The differences in effective emissivity between 319 IUEM-SVF and TUF-3D are, therefore, related to the 3D distribution of emissivity 320 321 of facets.

d) In some cases, urban spaces are highly complex in terms of both geometry and
heterogeneity of emissivity. Under these conditions, the spatial heterogeneity of
surface temperature may become large. This aspect cannot be evaluated using
IUEM-SVF, however TUF-3D can be used. To determine realistic within-pixel
thermal heterogeneity, airborne thermal image data at very high spatial resolution
is used. Component temperatures based on the airborne images were input into
TUF-3D to analyze the effect of thermal heterogeneity on radiative transfer in urban

spaces (Case d1 and d2 in Table 1). The effective emissivity based on TUF-3D and
effective emissivity based on IUEM-SVF under the isothermal cases c) was
compared to evaluate the thermal heterogeneity effect on the effective emissivity
and surface temperature retrieval.

TUF-3D calculates upwelling radiance at the top of the urban canopy and the full 333 radiative balance of each facet (and sub-facet components). Since emittance, 334 reflectance and exitance are calculated separately, the effects of the geometry of built-335 up space can be evaluated in detail. TUF-3D calculates the radiation flux density 336 337 reaching a horizontal plane in the sky, i.e. it is the signal captured by a nadir-oriented radiometer with ~180° field of view. The radiation flux density calculated using TUF-338 3D, accounts for the actual 3D distribution of material emissivity and thermal 339 340 heterogeneity in case d1 and d2, respectively.

341

To evaluate the impact of using a nadir – viewing imaging radiometer to retrieve 342 the radiometric surface temperature of a complex 3D urban surface, two effective 343 emissivities were defined and calculated. A first effective emissivity was obtained by 344 combining the upwelling radiation calculated by TUF-3D (and measured by a nadir -345 viewing imaging radiometer) with the complete surface temperature, i.e. the area-346 weighted average of all facet temperatures. Then a second effective emissivity (mean 347 effective emissivity of the 3D surface) was calculated by taking the mean exitance over 348 all facets, i.e. including the vertical walls not observable by a narrow field of view 349 nadir-oriented radiometer. To compute this second effective emissivity, the complete 350

surface temperatures in the case c, d1 and d2 were used. The rooftop temperature (Tr) 351 is 328.56 K, street temperature (road) (Ts) is 316.64 K, and wall temperature (Tw) is 352 353 304.94 K respectively in a built-up area of Kowloon peninsula from high resolution thermal image. These component temperatures were used as the input in the TUF-3D 354 model (case d1). In order to investigate the effect of heterogeneous thermal distribution 355 for the radiative transfer, the rooftop temperature then was defined as 304.94 K and the 356 wall temperature was defined as 328.56 K to simulate the upwelling radiation and the 357 complete exitance (case d2). and the respective setups for the experiments or sub-358 experiments for comparison between IUEM-SVF and TUF-3D are listed in the Table 1. 359

360

361 *3.4 Application of IUEM-SVF*

362 According to Lagouarde et al. (2012), the nighttime thermal infra-red anisotropy remains lower than 1 K for zenith viewing angles up to 50°, and nighttime anisotropy 363 is insensitive to the azimuth viewing direction. This is because the component 364 temperatures are rather similar, and thus such targets can be taken as isothermal. Our 365 numerical experiments (a1) through (d2) explore a broader range of cases than the one 366 applying to the observations by Lagouarde et al. (2012). The winter nighttime ASTER 367 data acquired on December 16th 2015, 14:36 (UTC) was used to retrieve the land surface 368 temperature. The effective emissivity was estimated for each pixel by applying IUEM-369 SVF and then it was applied to retrieve the urban surface radiometric temperature from 370 the band 13 of ASTER 2B01 product. Equation 8 was used to retrieve the radiometric 371 surface temperature from band 13 ASTER 2B01 at-surface radiance product. The 372

atmospheric effect was neglected since the ASTER radiance product 2B01 is the derived ground radiance with minimizing the atmospheric effect. The $v_{sky}(i)$ and $v_{adj}(i)$ were calculated using the DSM Lidar data and the building GIS data, as described by Yang et al. (2015b) and Zakšek et al. (2011). Two sets of SVF for only calculating within a pixel and considering the effects from adjacent pixels were generated as described by Yang et al. (2015b).

The radiometric surface temperature retrieved was then compared with the LST product ASTER 2B03 based on the Temperature and Emissivity Separation method (TES). This algorithm is based on a relationship derived from spectral libraries between the minimum value of spectral emissivity and its spectral variance to estimate the emissivity in each ASTER band. While the geometry effect changes the spectral emissivity in urban areas, the limitation of the TES application in urban surface temperature retrieval was analyzed based on the comparison between these LSTs.

Additionally, considering the effective emissivity calculated as the area-weighted average of the component emissivity without considering geometry effect in case b) of section 3.3, the retrieved urban surface temperature under this condition was also evaluated. An overview of the application of IUEM-SVF for urban surface temperature retrieval is shown in Figure 2.

391

392

"Insert Figure 2 here"

393

394 **4. Results**

4.1 Comparison between the IUEM-SVF and TUF-3D

396	a)	Figure 3 compares the effective emissivities derived from TUF-3D with those
397		derived from IUEM-SVF, when the components (e.g. rooftop, street and wall)
398		in the domain have same temperature and material emissivity. The effective
399		emissivity derived from IUEM-SVF is similar to the effective emissivity
400		derived from the TUF-3D model, with a correlation coefficient r^2 of 0.99, and
401		RMSDs of 0.004, 0.003, 0.002 and 0.002 when material emissivity is 0.80,
402		0.85, 0.90 and 0.95 respectively (Cases a1 to a4 in Table 1). This indicates that
403		the IUEM-SVF model can provide an accurate estimation of the effective
404		emissivity when all pixel elements have the same temperature and material
405		emissivity.
406		
407 408 409		"Insert Figure 3 here"
410	b)	The plan area index was set as 0.25 in the simplified configuration of TUF-3D.
411		In this condition, the area-weighted average of the component emissivity
412		without considering geometry does not change with increasing building height.
413		On the contrary, the effective emissivity based on the TUF-3D radiative fluxes
414		increases with increasing building height (decrease of SVF). In the experiment
415		with uniform surface temperature and material emissivities for all facets (Case
416		a1 in Table 1), the difference between these two sets of emissivities can reach
416 417		a1 in Table 1), the difference between these two sets of emissivities can reach0.10 when the material emissivity is assumed as 0.8 and the difference is about

emissivity was set as 0.945 and street emissivity was set as 0.948 (e.g. same 419 with the ASTER band 13) (Case b in Table 1), the area-weighted mean 420 emissivity of a horizontal surface was 0.947. The effective emissivity from 421 TUF-3D increased with the decrease of SVF, thus the difference between the 422 effective emissivity from TUF-3D and the area-weighted mean emissivity of 423 a horizontal surface increased. When the H/W (SVF) is 4 (0.59) in the 424 simplified configuration of TUF-3D, the difference between these two sets of 425 emissivities was about 0.02. The high-rise buildings and narrow streets of 426 427 Hong Kong yields H/W larger than 4 in many cases, thus the difference between the average of horizontal surface emissivity and the effective 428 emissivity will be larger. 429

c) In the Case c of Table 1, the upwelling radiation and the complete exitance, 430 derived from TUF-3D, were both used to calculate the effective emissivity. 431 First the material emissivities of the mixed pixels were calculated as the area-432 weighted average of the component material emissivities of horizontal 433 surfaces and vertical walls, then the effective emissivity was calculated by 434 applying IUEM-SVF with these material emissivities and SVF. The RMSD 435 between the effective emissivity based on upwelling radiation of TUF-3D and 436 IUEM-SVF is 0.010. This result shows that the heterogeneity of material 437 emissivity has small implication on the effective emissivity in ASTER band 438 13. If the effective emissivity derived from TUF-3D is calculated as the ratio 439

440		of the complete exitance of all facets to the Stefan-Boltzmann emittance at the
441		complete surface temperature, the RMSD with IUEM-SVF is 0.018 (Figure 4).
442		
443		"Insert Figure 4 here"
444		
445	d)	The upwelling radiation and the complete exitance, calculated by TUF-3D,
446		were also used to calculate the effective emissivity. The results showed that
447		the difference between the IUEM-SVF and TUF-3D effective emissivity based
448		on the upwelling radiation and the complete surface temperature becomes
449		large (RMSD = 0.063), when the surface temperature heterogeneity in a mixed
450		pixel is large, e.g. the rooftop temperature (Tr) is 328.56 K, street temperature
451		(road) (Ts) is 316.64 K, and wall temperature (Tw) is 304.94 K (Figure 5)
452		(Case d1 in Table 1). The effective emissivity determined as the ratio of the
453		mean upwelling radiance to the emittance at the complete surface temperature
454		which is the total direct emittance of all facets, however, is even larger than 1
455		in some cases (Figure 5). As explained in section 3.3, this is caused by the
456		upwelling radiation not including the exitance of all vertical facets. The wall
457		temperature is lower than the rooftop temperature in this case, thus the
458		complete temperature decreases when the fraction of wall increases with
459		increasing H/W (or decreasing SVF). Vertical facets contribute to directional
460		radiance travelling vertically through direct emission and multiple reflection
461		and this does not linearly depend on increasing of wall areas. The contributions

462	to exitance at a horizontal surface by the building rooftop and street fractions
463	do not change with increasing H/W, thus their contribution to the radiance
464	observed by a narrow IFOV, nadir-viewing imaging radiometer remains
465	constant when H/W changes. This is not consistent with the change of
466	complete temperature caused by the increase of the H/W. In order to evaluate
467	the thermal heterogeneity, the wall temperature is defined as higher than
468	rooftop temperature (e.g. the rooftop temperature (Tr) is 304.94 K, street
469	temperature (road) (Ts) is 316.64 K, and wall temperature (Tw) is 328.56 K)
470	(Case d2 in Table 1). In this condition, the complete temperature increases with
471	increasing H/W, but the contribution to the upwelling radiation due to multiple
472	reflection does not agree well with the increases of the complete temperature.
473	The effective emissivity based on the upwelling radiation decreases with
474	increasing H/W (Figure 6). The effective emissivity derived from the ratio of
475	the complete exitance of all facets to the emittance at the complete surface
476	temperature is smaller than 1, and not sensitive to the thermal heterogeneity
477	(Figure 4, 5 and 6). The RMSD between the effective emissivity derived from
478	IUEM-SVF and TUF-3D is 0.018.
479	
480	"Insert Figure 5 and Figure 6 here"
481	

482 Results show that urban geometry should not be neglected and estimating483 emissivity using an area-weighted approach without geometry effect would give a large

error in a dense built-up area. When the pixel is thermally heterogeneous, the deviation 484 between the effective emissivity determined as the ratio of observed (line of sight) 485 486 exitance to the emittance at the complete surface temperature and the effective emissivity determined by IUEM-SVF is large (e.g. 0.063). When the pixel is isothermal, 487 the deviation is small (e.g. 0.010) due to the factor of material heterogeneity only. The 488 complete exitance is not the radiance observed by the remote sensor. A nadir-viewing 489 radiometer captures only total upward exitance, but cannot observe all facets with 490 consideration of thermal heterogeneity of the target. An imaging radiometer with multi-491 492 angular capability can thus be used to improve the sampling of the thermal heterogeneity of the urban surface and retrieval of the radiometric surface temperature 493 in urban areas. 494

495

496 *4.2 Application of IUEM-SVF and retrieval of the surface radiometric temperature*

In order to evaluate the importance of considering geometry effect in estimating 497 effective emissivity, two cases were tested: a) weighted average of component 498 emissivity without considering geometry effect and b) using the IUEM-SVF method. 499 In case (b), the SVF was determined in two ways: (b1) where the SVF is calculated and 500 limited within the coverage of an ASTER pixel (90 m x 90 m); and (b2) where adjacent 501 pixels are taken into account during SVF calculation. The SVF based on b1 was used 502 to input into the calculation of effective emissivity based on IUEM-SVF and separate 503 the effect of neighboring pixels. The map of SVF (b1) (Figure 7a) shows higher values 504 than the map of SVF (b2) (Figure 7b). Figure 7c is the difference of two SVF maps and 505

the difference of these two SVF maps over built-up areas is higher than that over flat
areas. The difference between two sets of SVFs range from 0.0 to 0.20 with mean value
of 0.074 (Figure 7d).

First, emissivity was estimated using the area-weighted average approach with component emissivities and without considering geometry effect (Figure 8a). This emissivity was calculated for ASTER band 13 using land use and land cover data, building GIS data, the ASTER spectral emissivity library and applied to a flat surface. Figure 8b shows the material emissivity for ASTER band 13 of the complete facets and Figure 8c shows the effective emissivity for ASTER band 13 calculated with Equation 9.

- 516
- 517

"Insert Figures 7 and 8 here"

519	Second, the surface radiometric temperature was retrieved by applying the single
520	channel method to ASTER band 13 data and using different sets of effective emissivity
521	data. The difference in LST between the retrieval based on IUEM-SVF (Figure 9b) and
522	the ASTER 2B03 LST product (Figure 9a) is much smaller than the difference in LST
523	between the retrieval without geometry effect (Figure 9c) and the ASTER 2B03 LST
524	product (Figure 9a). The histogram of the differences (Figure 9d) illustrates the impact
525	of the effective emissivity on retrieved LST. The LST without considering geometry
526	effect is much higher than the ASTER 2B03 LST product (mean difference about 1.72
527	K) and the LST with geometry effect (mean difference about 1.7 K). Thus the most

values of the differences between the LST without geometry effect and the other two 528 sets of LSTs are positive and these differences can reach 2 to 3 K (Figure 9d). The 529 spatial pattern of the LST without considering geometry is significantly different from 530 the ASTER 2B03 LST product, while the spatial pattern of the LST with geometry 531 effect is similar to the ASTER 2B03 LST product. The difference between the LST 532 with geometry effect and ASTER 2B03 LST product ranges from -2 K to 2 K and the 533 RMSD between these two kind of LST is about 0.91 K. The temperature in built-up 534 areas is higher than the flat impervious areas (Figure 9a to c). 535

536 The difference between the ASTER data product and the LST with geometry effect is not very large because the retrieval of emissivity from the spectral radiance 537 (TES) accounts for at least part of the geometry effects, although the material emissivity 538 is used to construct the relation between the minimum spectral emissivity and the 539 maximum minimum difference (ε_{min} -MMD). This difference is due to the estimation 540 of the material emissivity on the basis of land cover. The empirical relationship between 541 the minimum spectral emissivity and the maximum minimum difference (ε_{min} -MMD) 542 is an error source since this relationship was established by analyzing 86 laboratory TIR 543 emissivity spectra data (Gillespie et al., 1998), and they do not take the geometry factor 544 into account. Moreover, these materials are different from building construction 545 materials in urban areas (e.g. glass, metal). Payan and Royer (2004) demonstrated that 546 urban metals do not follow these empirical relationships, and nowadays more and more 547 548 new materials and metals are used in the construction industry. Additionally, the geometry effect changes the spectral contrast, e.g. due to the geometry effect, the 549

550	effective spectral emissivity in the high density urban space tends to 1 and the spectral
551	contrast tends to 0. Thus the empirical relationship between the minimum emissivity
552	and spectral contrast does not necessarily apply to high density urban areas.
553	
554	"Insert Figure 9 here"
555	
556	5. Discussion
557	The complex geometry of urban surfaces has significant implications in the
558	retrieval of temperature over urban areas (Voogt, 1995; Voogt, 2008; Lagouarde et al.,
559	2010). This study focused on the multiple reflection due to the geometry effect over the
560	built-up space, and analyzed the relationship between upward radiance and the urban
561	surface temperature. The performance of IUEM-SVF was evaluated with the TUF-3D
562	model. When the component material and surface temperature are uniform, the IUEM-
563	SVF model estimates the effective emissivity accurately. This implies that the upward
564	radiance of the urban canopy significantly relates to the complete surface temperature
565	in the IUEM-SVF, under the assumption that the pixel is homogenous. When the pixel
566	is thermally heterogeneous, the difference between the effective emissivity derived
567	from the upward radiance of TUF-3D and complete surface temperature and the
568	effective emissivity derived from IUEM-SVF becomes larger and this difference
569	changes with the thermal distribution of the facets. The upward radiance observed by
570	nadir remote sensor is just a part of the total exitance from all facets and it is the result

571 of the complex radiative interactions of different facets within a pixel, while the

572 complete surface temperature is just the simple linear area-weighted average of 573 component temperatures. Both models evaluated in this study have limitations: TUF-574 3D only works with simple urban geometry without considering spatial variety, and 575 IEUM-SVF does not consider the differences in component temperatures within a grid 576 cell (pixel). Thus the IEUM-SVF still requires further evaluation in the future study.

Lagouarde et al. (2010) and Lagouarde et al. (2012) investigated anisotropy during 577 both nighttime and daytime over Marseille and Toulouse, France and their experimental 578 results showed that the differences in surface brightness temperature vary between -5 579 580 and 7 K between nadir and off-nadir measurements in daytime. However, the nighttime thermal infra-red anisotropy remains lower than 1 K when zenith viewing angle up to 581 50°, and the nighttime anisotropy is insensitive to the azimuth viewing direction. This 582 583 is because the nighttime difference between the component temperatures is much smaller than in daytime, and the primary forcing for directional anisotropy is zenith 584 viewing angle, and not azimuth viewing angle (in contrast to daytime when the sun 585 586 generates strong azimuthal temperature variation). Compared with the material emissivity based on classification method or simply linear mixing of the component 587 emissivities in mixed pixel, the IUEM-SVF model improves the effective emissivity 588 estimation and reduces the errors in retrieval of urban surface temperature when the 589 pixel is isothermal or close to isothermal. Accordingly, retrieval of urban surface 590 radiometric temperature is much easier with radiometric data observed in nighttime and 591 IUEM-SVF can be used to derive effective emissivity and further retrieve urban surface 592 radiometric temperature with promising performance. 593

When the pixel is heterogeneous, the thermal heterogeneity within a mixed pixel 594 in urban areas has a significant impact on the exitance measured by a remote sensor, 595 thus the errors of retrieved surface temperature for summer daytime data would be 596 larger and the retrieved surface temperature from remote sense requires better physical 597 interpretation. One solution is to consider the radiative transfer among the components 598 in the heterogeneous mixed pixel, however, this is challenging due to the low spatial 599 resolution of satellite TIR data, which do not capture the temperature distribution in a 600 mixed pixel. The direction of observation can also be taken into account to retrieve the 601 602 surface radiometric temperature, and data obtained from off-nadir sensor provide information of the vertical facets over urban areas, which may provide supporting 603 information to retrieve urban complete surface radiometric temperature for the urban 604 605 energy heat flux estimation. When the IUEM-SVF is used to retrieve the urban surface temperature under the condition of thermal heterogeneity, the geometry effect on 606 effective emissivity can be considered. It can reduce the bias of the retrieved urban 607 608 surface temperature, derived from the emissivity based on classification and assuming horizontal surfaces. The surface temperature retrieved from IUEM-SVF considers all 609 facets within a pixel and this is not just a simple linear mixture of different components. 610 It is also different from the complete surface temperature. In addition, considering the 611 complexity of urban surface temperature retrieval, validation of the output surface 612 temperature from satellite data is still a challenge and this will be studied in near future. 613 For the application of the TES algorithm for ASTER in urban areas, the urban 614 geometry effect makes the spectral contrast of the urban canopy different from the 615

material spectra and a flat surface. One solution is to establish a new relationship for
different urban geometry characteristics and surface materials using geometrical
parameters such as SVF and/or H/W.

619

620 **6.** Conclusion

This study improved UEM-SVF model (now the IUEM-SVF model) by 621 considering the effect of 3-D urban geometry, not only in terms of the cavity effect on 622 emission, but also on reflection by the facets within a pixel. The IUEM-SVF was further 623 624 evaluated with the micro-climate radiative transfer TUF-3D model. The IUEM-SVF shows promising results when the pixel is thermally homogeneous: in this case the 625 RMSDs between the effective emissivity derived from IUEM-SVF and TUF-3D are 626 627 0.004, 0.003, 0.002 and 0.002 when the material emissivity is 0.80, 0.85, 0.90 and 0.95 respectively. This implies that the observed exitance of the urban canopy can be related 628 to the surface radiometric temperature under the assumption that the pixel is uniform in 629 630 material composition and thermal status. When the pixel is heterogeneous, the difference between the effective emissivity derived from IUEM-SVF and the effective 631 emissivity derived from TUF-3D is larger. When the IUEM-SVF is used to retrieve 632 urban surface temperature under the thermally heterogeneous condition, e.g. summer 633 daytime, the bias can be reduced compared with using the emissivity based on 634 classification. The retrieved urban surface temperature is a summarized value including 635 the interaction among horizontal surfaces and vertical facets. Although IUEM-SVF can 636 be applied to retrieve the urban surface radiometric temperature, and is outperformed 637

on nighttime image, it still requires some further evaluation in the near future. The Temperature and Emissivity Separation (TES) algorithm of ASTER based on the ε_{min} -*MMD* relationship of the material emissivity may cause bias for retrieval of urban surface temperature, and this bias depends on the geometry characteristics of the builtup space. The ε_{min} -*MMD* relationship can be reconstructed by considering urban geometry effect and can further be applied to the ASTER TES algorithm in order to retrieve accurate urban surface radiometric temperature.

645

646

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