# The near-bed flow structure and bed shear stresses within emergent vegetation canopies

Mario Conde-Frias<sup>1</sup>, Marco Ghisalberti<sup>1</sup>, Ryan Joseph Lowe<sup>1</sup>, Maryam Abdolahpour<sup>1</sup>, and Vahid Etminan Farooji<sup>1</sup>

<sup>1</sup>University of Western Australia

November 23, 2022

#### Abstract

The structure of the bottom boundary layer (BBL) in aquatic flows influences a range of biophysical processes, including sediment transport, hyporheic exchange, and biofilm formation. While the structure of BBL above bare sediment beds has been well studied, little is known about the complex near-bed flow structure within canopies of aquatic vegetation. In this study, we used high-resolution laboratory measurements and numerical Large Eddy Simulations to investigate the near-bed mean and turbulent flow properties within staggered-ordered emergent canopies under a wide range of flow and canopy conditions. There is strong horizontal variability of key near-bed flow characteristics on the scale of the vegetation elements. Measurement locations that provide single-point flow characteristics closest to the spatially-averaged values were identified. The spatiallyaveraged BBL thickness is influenced strongly by canopy density. This impact of canopy density is engendered through its direct control of near-bed turbulent kinetic energy (TKE), which in turn is negatively correlated with BBL thickness, both locally in a given flow and across the range of flow conditions studied here. A model based on the near-bed TKE is developed to predict the BBL thickness and, ultimately, the bed shear stress. The strong agreement between model predictions and experimental data may explain why both TKE and bed shear stress may be seen as drivers of sediment transport processes in vegetated flows. These findings provide new insights into the physical links between near-bed flow variables and therefore contribute to the understanding of some of the complex biophysical processes present in vegetated flows

1	The near-bed flow structure and bed shear stresses within emergent vegetation
2	canopies
3	M. Conde-Frias <sup>1,2,3</sup> , M. Ghisalberti <sup>1,2</sup> , R. Lowe <sup>1,2,3</sup> , M. Abdolahpour <sup>1,2,3</sup> , and V. Etminan <sup>1</sup>
4 5	<sup>1</sup> Oceans Graduate School, University of Western Australia, Perth, Australia. <sup>2</sup> UWA Oceans Institute, University of Western Australia, Perth, Australia.
6 7	<sup>3</sup> ARC Centre of Excellence for Coral Reef Studies, University of Western Australia, Perth, Australia.
8	Corresponding author: Mario Conde-Frias (mario.conde-frias@research.uwa.edu.au)
9	Key Points
10 11	• Vegetated flows are characterized by strong spatial variability of key near-bed flow characteristics on the scale of the vegetation elements.
12 13	• The thickness of the bottom boundary layer (locally and in the mean) is strongly controlled by the near-bed turbulent kinetic energy (TKE).
14 15	• Spatial changes of near-bed TKE induce changes in the BBL thickness, which in turn generates strong spatial variability of bed shear stress
16	(The above elements should be on a title page)

# 18 Abstract

The structure of the bottom boundary layer (BBL) in aquatic flows influences a range of 19 20 biophysical processes, including sediment transport, hyporheic exchange, and biofilm formation. 21 While the structure of BBL above bare sediment beds has been well studied, little is known about the complex near-bed flow structure within canopies of aquatic vegetation. In this study, we used 22 high-resolution laboratory measurements and numerical Large Eddy Simulations to investigate the 23 near-bed mean and turbulent flow properties within staggered-ordered emergent canopies under a 24 25 wide range of flow and canopy conditions. There is strong horizontal variability of key near-bed 26 flow characteristics on the scale of the vegetation elements. Measurement locations that provide single-point flow characteristics closest to the spatially-averaged values were identified. The 27 spatially-averaged BBL thickness is influenced strongly by canopy density. This impact of canopy 28 29 density is engendered through its direct control of near-bed turbulent kinetic energy (TKE), which 30 in turn is negatively correlated with BBL thickness, both locally in a given flow and across the 31 range of flow conditions studied here. A model based on the near-bed TKE is developed to predict 32 the BBL thickness and, ultimately, the bed shear stress. The strong agreement between model predictions and experimental data may explain why both TKE and bed shear stress may be seen as 33 drivers of sediment transport processes in vegetated flows. These findings provide new insights 34 into the physical links between near-bed flow variables and therefore contribute to the 35 understanding of some of the complex biophysical processes present in vegetated flows. 36

#### 37 **1 Introduction**

38 Canopies formed by aquatic vegetation provide numerous ecosystem services in riverine 39 and coastal environments. For instance, these canopies can help to stabilize mobile sediment beds 40 by transforming an erosional bed into a depositional one (Shields et al., 1995). The enhanced drag forces induced by the canopy can also modify circulation and sediment transport pathways, which 41 42 over time can shape the morphological evolution of riverine and coastal systems (Ward et al., 43 2000; Hooke, 2007; van Katwijk et al., 2010; Vargas-Luna et al., 2015). Moreover, canopies can mitigate the impact of coastal hazards (i.e. coastal erosion, flooding, etc.) by dissipating wave 44 energy (Massel et al., 1999; Türker et al., 2006; Lowe et al., 2007). Finally, aquatic canopies help 45 shape a wide range of environmental and ecological processes, such as improving water quality 46

(Cundy et al., 2005) and promoting biodiversity by creating sheltered habitats (Hemminga & 47 Duarte, 2000). 48

49 The interaction between hydrodynamics and sediment processes in aquatic systems is 50 controlled by the layer of water adjacent to the bed, the bottom boundary layer (BBL) (Nowell & 51 Jumars, 1984; Wüest & Lorke, 2003; Trowbridge & Lentz, 2018). The physical structure of this 52 layer governs several critical physical, chemical, and biological processes, such as sediment transport (Yalin, 1977; van Rijn, 2007a, 2007b), hyporheic exchange (Grant et al., 2018; Roche et 53 54 al., 2018; Voermans et al., 2018) and biofilm formation (Beer & Kühl, 2001). Due to its importance, the structure of the BBL above bare beds has been widely investigated, with the 55 56 vertical structure of mean velocities, turbulence levels, and shear stresses well-known. However, 57 the influence of vegetation on BBL characteristics is poorly understood due to the fine-scale 58 variability of the flow within aquatic vegetation and the difficulties associated with obtaining highresolution measurements within this layer. Thus, studies that directly resolve near-bed velocities, 59 60 stresses, and turbulence intensities in vegetated flows are of paramount importance in understanding the variation of near-bed hydrodynamic parameters (including those relevant to 61 62 sediment transport) with canopy characteristics.

This study aims to characterize the near-bed mean and turbulent flow structure within the 63 BBL in emergent canopies. By combining laboratory measurements and numerical simulations for 64 65 a wide range of canopy densities and flow conditions, we determine BBL thickness (O(mm)) and its variation with flow and canopy characteristics. In addition, we examine the vertical and 66 67 horizontal structure of the mean velocity, near-bed stresses, and turbulent kinetic energy within 68 the BBL. Finally, we elucidate the link between bed shear stress and turbulent kinetic energy, which explains why both have been used as predictors of sediment transport in vegetated systems. 69

70

#### 1.1 Flow and vegetation interaction

71 Given the vast diversity of plant morphology in natural environments, canopies are often modelled simply as arrays of rigid dowels (Nepf & Ghisalberti, 2008; Abdolahpour et al., 2018) 72 73 arranged in different configurations (e.g. squared, staggered or randomly). The key geometric characteristics of such a model canopy are the diameter (d) and the height (h) of the vegetation 74 75 elements, as well as the number of elements per bed area (n). A nondimensional measure of the

- canopy density is the solid volume fraction occupied by the canopy elements  $\phi$ , equal to  $(\pi/4)ad$
- for cylindrical dowels, with a (= nd) being the frontal area per canopy volume. The variation of
- these canopy properties can greatly affect the mean and turbulent velocity structure within
- canopies by controlling the drag exerted on the flow (Nepf & Vivoni, 2000; Nepf, 2012a).
- 80

81



Figure 1. Typical structure of the mean streamwise velocity  $(\bar{u})$  field in an emergent canopy normalized by the local maximum velocity  $(\bar{u}_{max})$ , obtained here from numerical Large Eddy Simulation. Element wakes, and the tortuous nature of the flow within the canopy create high spatial variability at the element scale.

85 Due to the obstructions presented by canopy elements and element wakes, the flow is forced to move around each element, such that the velocity field is spatially heterogeneous at the 86 scale of the element diameter (Figure 1). To account for this element-scale spatial heterogeneity, 87 the double-averaging procedure was employed, whereby instantaneous velocity statistics ( $\zeta$ ) are 88 89 first decomposed into a temporal average (indicated by an overbar) and deviations from it (indicated by a prime), such that  $\zeta = \overline{\zeta} + \zeta'$ . Time-averaged quantities are further decomposed into 90 a spatial mean and deviation, with angular brackets denoting a horizontal average at height z over 91 several canopy elements and double primes indicating deviations from the horizontal mean ( $\bar{\zeta}$  = 92  $\langle \bar{\zeta} \rangle + \bar{\zeta}''$ ) (Raupach & Shaw, 1982; Nikora et al., 2001). The double-averaged Navier-Stokes 93

equation in the streamwise direction for steady and uniform flow through aquatic vegetationbecomes:

96

97
$$0 = gsin\theta - \frac{1}{\rho} \frac{\partial \langle \bar{p} \rangle}{\partial x} - \frac{\partial}{\partial z} \left[ \langle \overline{u'w'} \rangle - \langle \bar{u}'' \overline{w}'' \rangle + v \frac{\partial \langle \bar{u} \rangle}{\partial z} \right] - \frac{1}{2} \frac{C_D a}{(1 - \phi)} \langle \bar{u} \rangle^2,$$
(1)
(1)
(1)

98

(Raupach and Shaw 1982) where u and w are the velocity components in the directions of x99 (streamwise direction) and z (vertical direction), respectively,  $\rho$  is the water density,  $\nu$  is the 100 101 kinematic viscosity,  $\theta$  is the bed slope,  $C_D$  is the canopy drag coefficient, g is gravitational acceleration, and p is pressure. Term (i) is the gravity force, term (ii) is the pressure gradient, term 102 (iii) represents the spatially-averaged Reynolds stress, term (iv) the momentum flux associated 103 104 with spatial correlations of time-averaged velocity fields (termed the 'dispersive' stress) and term (v) the spatially-averaged viscous stress. The sum of these three terms defines the total fluid shear 105 stress,  $\tau$ . Term (vi) represents the spatially averaged drag associated with the canopy elements 106 (Nepf, 2012a). 107

The relative importance of each term in Eq. (1) depends on the properties of the flow, 108 109 vegetation, and height above the bed. In emergent canopies, the momentum equation can generally 110 be simplified into a balance between pressure gradient and the canopy drag over most of the water 111 column (Nepf, 2012a). However, within the strong shear of the BBL, other terms in Eq. (1) may become significant. As little is known of the complex near-bed flow structure within canopies, we 112 113 aim here, using a combination of experimental and numerical approaches, to directly resolve the mean velocity structure and stress components (terms (iii), (iv) and (v) in Eq. 1) in the region 114 adjacent to the bed across a wide range of canopy densities. 115

# 116 **1.2 Interaction between hydrodynamics and sediment transport**

117 Sediment transport in unvegetated channels is typically estimated from empirical 118 formulations based on the shear stress at the bed,  $\tau_b$  (often expressed in terms of friction velocity, 119  $u_* = \sqrt{\tau_b/\rho}$ ) (van Rijn, 1987; James et al., 2002; Jordanova & James, 2003)- In sediment transport 120 models developed for bare beds, the role of turbulence in suspending and transporting sediment

may be considered implicitly, given the inherent proportionality between the mean bed stress and turbulence intensities in the bare bed boundary layer (Stapleton & Huntley, 1995; Nepf, 1999). However, this proportionality breaks down in canopy flows, where the turbulence can be predominantly generated in plant wakes rather than at the bed (Nepf, 1999, 2012b). As a consequence, predictive models developed for bare beds are unlikely to be quantitatively applicable to vegetated environments.

Despite the fundamental differences between unvegetated and vegetated flows, some 127 128 studies (Jordanova & James, 2003; Kothyari et al., 2009; Larsen et al., 2009) have suggested that bed shear stress may control sediment transport within vegetation. However, it is still not well 129 established how canopies modify bed shear stresses. To bridge this gap, several methods for 130 estimating bed shear stress within canopies have been proposed (e.g. Jordanova & James, 2003; 131 Kothyari et al., 2009; Yang et al., 2015; Etminan et al., 2018). In general, these methods are based 132 133 on one of the following approaches: 1) estimation of the bed shear stress by the subtraction of the element drag (term (vi) in Eq. 1) from the driving forces (terms (i) and (ii) in Eq. 1) and 2) 134 135 estimation of bed shear from velocity measurements in the near-bed region. The main limitation of the first approach is that both variables (drag and the driving force) are often orders of magnitude 136 larger than the bed shear stress, resulting in potentially large errors when computing small 137 differences. Similarly, methods based on the second approach require measurements of the 138 velocity gradient within the viscous-stress dominated sublayer  $(H_{\rm w})$  in vegetated flows (region 139 immediately adjacent to the bed where turbulence is negligible, and the viscous stress dominates 140 141 the total stress). These methods are experimentally tricky, given that the thickness of this layer is typically  $\leq O(\text{mm})$ . Despite these difficulties, Yang et al. (2015) developed a model to estimate 142 the bed shear stress in aquatic vegetation based on the linear variation of the stress within the 143 144 viscous-stress dominated sublayer  $(H_{y})$ . In this model, the temporally- and spatially-averaged bed shear stress ( $\langle \bar{\tau}_b \rangle$ ) is estimated from the streamwise velocity above the BBL  $\langle \bar{u} \rangle$  and  $H_v$  (taken in 145 their study as min( $\frac{d}{2}$ , (22 ± 3) $\nu/u_*$ )). Although this model represents a step forward in the 146 prediction of bed shear stress in emergent aquatic vegetation, it is still limited since it does not 147 148 embed the significant impact that seems to have the canopy density in  $H_v$  and ultimately in the bed shear stress. To address some of the limitations found in the previous model, an alternative estimate 149 of the thickness of  $H_v$  can be inferred by assuming a balance between the production of the 150

turbulent kinetic energy in the cylinder wakes and the viscous dissipation of TKE near the bed(Etminan et al. 2018).

Several studies of sediment dynamics in vegetated systems (Tinoco & Coco, 2014, 2016, 153 2018; Yang et al., 2016; Yang & Nepf, 2018, 2019) have concluded that sediment transport 154 processes are correlated with levels of near-bed turbulence (a combination of the turbulence 155 produced in element wakes and the BBL). For instance, the threshold for sediment motion in 156 emergent vegetation was driven by a critical value of the turbulent kinetic energy, which can be 157 158 deduced from the critical velocity required for sediment motion in bare beds (Yang et al., 2016). 159 In a subsequent study, a reinterpretation of the bed-shear-stress-based Einstein-Brown bed-load transport model in terms of the turbulent kinetic energy was proposed to improve predictions of 160 bed-load transport within aquatic vegetation (Yang & Nepf 2018). Finally, improvements in the 161 prediction of suspended sediment concentrations within aquatic vegetation were found when the 162 163 effect of wake-generated turbulence was incorporated into traditional sediment transport models developed for bare beds (based on the Shields parameter) (Tinoco & Coco, 2016). Despite the 164 165 evidence of the strong role of the TKE on sediment transport within vegetated regions, from a 166 mechanistic perspective, turbulent kinetic energy does not represent a force on sediment grain, 167 whereas bed shear stresses do (as by definition represent the force exerted on the bed).

Previous studies have suggested that sediment transport models based on the bed shear stress fail within vegetated regions because they do not account for the vegetation-generated turbulence. In other words, they suggest that the 'external' turbulence from the plants and the bed shear stress are decoupled in these flows (Tinoco & Coco, 2018; Yang & Nepf, 2018). However, we believe that the complete unlink between the TKE and bed shear stress within vegetated flows is not necessarily valid.

Here, we examine the links between bed shear stresses and near-bed TKE levels in emergent canopies and hypothesize that the TKE generated by the vegetation increases vertical momentum transport, which in turn creates a thinner boundary layer and ultimately an increase in the bed shear stress. Accordingly, one of the priorities of this integrated experimental and numerical study is to look at the extent to which the near-bed turbulent kinetic energy controls the BBL thickness and, ultimately, the bed shear stresses within aquatic vegetation.

# 180 2 Methodology

# 181 **2.1 Experimental configuration**

Experiments were carried out in a 20-m-long, 0.6-m-wide, and 0.45-m-deep recirculating flume in the Coastal and Offshore Engineering Laboratory at the University of Western Australia (Figure 2a).

185



186

Figure 2. (a) Side view of the experimental setup. The flow depth was fixed at H = 26.5 cm. (b) Top view of the staggered canopy array, with PIV measurements taken at the four transects indicated by the coloured lines. *S* is the distance between the elements.

The flow, from 6.4-27.5 L/s, was generated by a recirculating pump. A honeycomb flow 190 straightener was installed on the upstream side of the flume to promote uniform parallel flow. 191 Experiments were designed to elucidate the impact of canopy density and flow speed on the near-192 193 bed flow structure in emergent aquatic vegetation. Given that most emergent aquatic vegetation is characterized by rounded shoots of high stiffness (Nepf, 2012a), simplified rigid elements 194 195 (cylindrical wooden dowels with a diameter d = 0.65 cm and height of 30 cm) were used as an 196 idealized proxy for emergent canopies. The dowels were fixed to a perforated PVC baseboard in a staggered pattern (Figure 2b) across the channel width. To ensure fully-developed canopy flow, 197

the total length of the canopy model  $(L_c)$  was at least five times the drag length scale  $L_d$ , defined as  $L_d = (C_D a)^{-1}$  (Lowe et al., 2005; Morse et al., 2002). To fully cover typical field ranges, four canopy densities ( $\phi$ ) ranging from 0.016 to 0.098 were studied. For each canopy density, five flow conditions ( $U_b$ ) were tested. Here  $U_b$  is the bulk channel velocity, defined as Q/(WH), with W the flume width and H the water depth (which was kept constant at 26.5 cm for all cases). This experimental program is summarized in Table 1.

204

Density,	Frontal area,	Bulk channel velocity,
${oldsymbol{\phi}}$	<i>a</i> [m <sup>-1</sup> ]	<i>U<sub>b</sub></i> [m/s]
0.016	3.1	
0.025	5.2	0.04.0.06.0.00.0.12.0.19
0.044	9.3	0.04, 0.06, 0.09, 0.13, 0.18
0.098	20.8	

 Table 1. Experimental conditions

# 205 **2.2 Particle Image Velocimetry**

206 Velocity measurements were obtained using particle image velocimetry (PIV) at four locations within the array (Figure 2b). The PIV system consisted of a 300 mW continuous-wave 207 (CW) 532 nm DPSS laser with a 60° fan angle and a light sheet thickness of 1 mm. Images were 208 recorded for 7 min with a 12-bit, 1024 x 1024, CMOS camera (Photron FASTCAM SA3), 209 equipped with a Nikon AF Nikkor 50mm f/1.4d lens. The frame rate of the camera was adjusted 210 to ensure that maximum particle displacement was always less than a quarter of the interrogation 211 window (Raffel et al., 2012) and ranged from 125-500 fps. Polyamide particles with a nominal 212 diameter of  $43 - 55 \,\mu\text{m}$  and a specific gravity of 1.04 were used as tracer particles. 213

214 The analysis of recorded images was carried out using the open software PIVlab (Thielicke & Stamhuis, 2014). Before the PIV analysis, all images were pre-processed to enhance the 215 216 contrast. Velocity fields were determined indirectly by finding the displacement of particles between two consecutive frames through a multi-pass window deformation scheme with 217 decreasing interrogation window sizes. Depending on the experiment, two or three successive 50% 218 size passes with 50% overlapped interrogation windows ( $64 \times 64$  to  $16 \times 16$  pixels or  $32 \times 32$  to 219 220  $16 \times 16$  pixels) along with a Gaussian sub-pixel interpolation scheme (Nobach & Honkanen, 2005; Raffel et al., 2012) were used to obtain high-resolution velocity maps near the bed. However, due 221

to the finite number of particles, fewer particles in each window are cross-correlated as the 222 interrogation area gets smaller, creating more spurious velocity vectors (Raffel et al., 2012). To 223 224 overcome this, an ensemble correlation approach was implemented (Santiago et al., 1998; Meinhart et al., 2000;), whereby cross-correlation for several image pairs is computed, with 225 correlation maps then averaged for peak detection and calculation of velocity components. This 226 approach increases the SNR (as more image pairs are ensemble-correlated), allowing to obtain 227 high-resolution velocity fields even with low particle densities. In each experiment, the number of 228 images for the ensemble correlation was chosen to provide velocity data at a minimum frequency 229 of 25 Hz, which, for the frame rate used, were between 5-20 image pairs. Erroneous velocity 230 vectors (outliers) were identified and eliminated through a combination of a global standard 231 deviation filter and a local median filter, which evaluates the velocity fluctuation with respect to 232 233 the median in 3 x 3 neighbourhoods around a central vector (Westerweel & Scarano, 2005).

#### 234 2.3 Numerical model

Three-dimensional Large Eddy Simulations (OpenFOAM v2.3.0) of flow through 235 emergent vegetation using the numerical methodology of Etminan et al. (2018) were employed in 236 this study. A detailed description can be found in Etminan et al. (2017, 2018), and therefore, only 237 a summary is provided herein. A set of four cylinders (d = 0.01 m) in a staggered arrangement 238 239 was used to represent emergent vegetation. To simulate an infinite array of cylinders, cyclic boundary conditions were imposed in the streamwise and spanwise directions of the computational 240 241 domain, formed by four sets of an O-grid block and a Cartesian H-grid block (refer to Figure 2 in Etminan et al. (2018)). In addition, a no-slip condition was imposed around each cylinder and at 242 243 the bed. The data was only collected after 15 flow-through periods to ensure a fully-developed flow condition. 244

The ratio between the diameter of, and distance between, the elements was varied to achieve canopy densities ranging from  $\phi = 0.016 - 0.25$  (Table 2). For each density, three flow conditions (based on the pore velocity,  $U_p$ ) were tested; this velocity is defined as that averaged over the fluid space (Tanino & Nepf, 2008b) and can be calculated as  $Q/WH(1 - \phi)$ , equivalent to  $U_p = U_b/(1 - \phi)$ . Time-averaged flow parameters were evaluated across a minimum period of 45 flow-through periods.

Density,	Frontal area,	Pore velocity,
$\phi$	<i>a</i> [m <sup>-1</sup> ]	<i>U<sub>p</sub></i> [m/s]
0.016	2.0	
0.04	5.1	
0.08	10.2	0.05 0.10 0.12
0.12	15.3	0.05, 0.10, 0.15
0.20	25.5	
0.25	32.0	

**Table 2. Numerical simulation conditions** 

253 Previous studies have extensively validated the numerical model employed here for 254 emergent canopy flow (Etminan et al., 2017, 2018).

A comparison of the time-averaged streamwise velocity and turbulence intensity profiles 255 against experimental data reported by Liu et al. (2008) at five locations within the array showed 256 257 that the model accurately reproduces the experimental profiles of streamwise velocity and turbulence intensity at all five locations (as shown in Figure 3 of Etminan et al. (2017)). In addition, 258 to specifically investigate the capacity of the model to reproduce the near-bed flow structure, 259 numerical profiles of both viscous and Reynolds stresses in the near-bed region were compared 260 against experimental data reported in Yang et al. (2015). This comparison again yielded strong 261 agreement between numerical and experimental data ( $R^2 = 0.95$  and 0.81 for viscous and 262 Reynolds stresses, respectively, Figure 3 in Etminan et al. (2018)). 263

# 264 **2.5 Parameter definitions and data analysis**

# 265 **2.5.1. Experimental data**

Traditionally, the boundary layer thickness is defined as the height above the boundary at which the velocity reaches 99% of the free stream velocity (Nowell & Jumars, 1984). This definition, however, cannot be easily applied in vegetated flows due to the generation of a secondary flow pattern (velocity overshoot) at the base of the canopy elements (Figure 3a). In the case of emergent vegetation, the velocity reaches a depth-uniform value some distance from the bed (inset figure 3a). The local thickness of the BBL ( $\tilde{\delta}$ ) is defined here as the height at which the temporally-averaged velocity ( $\bar{u}$ ) first reaches this value (Figure 3a). For the experimental data presented in this paper, all spatially-averaged flow statistics are averages over transects 2, 3 and 4 (Figure 2b). In particular, the temporally- and spatially-averaged streamwise velocity ( $\langle \bar{u} \rangle$ ) in the region z/d > 4 was always within 5% of the estimated pore velocity  $U_p$  (Figure 3b) across all runs. Additionally,  $\langle \bar{u} \rangle$  at the top of the spatially-averaged BBL layer thickness ( $z = \langle \tilde{\delta} \rangle$ ) can be approximated to  $U_p$ .



278

Figure 3. (a) Time-averaged streamwise velocity profiles normalized by the pore velocity  $(U_p)$  at a typical location on each of the measured transects (Figure 2b) from experimental data with  $\phi = 0.025$  and  $U_p = 0.065$  cm/s. The arrow indicates the BBL thickness  $(\tilde{\delta})$  and the black dashed line represents the depth-uniform velocity. The inset shows an enlarged view. (b) The temporally- and spatially-averaged streamwise velocity  $(\langle \bar{u} \rangle)$  profile normalized by the pore velocity  $(U_p)$ . The blue dashed line indicates the spatially-averaged BBL layer thickness  $(\langle \tilde{\delta} \rangle)$  and the red dashed line represents the depth-uniform velocity. The temporally- and spatially-averaged streamwise velocity at the top of  $\langle \tilde{\delta} \rangle$  is within 5% the estimated pore velocity (arrow).

In all flows, the temporally- and spatially-averaged total stress (
$$\langle \bar{\tau}(z) \rangle$$
):

287 
$$\langle \bar{\tau}(z) \rangle = \langle \rho v \frac{\partial \bar{u}}{\partial z} \rangle - \rho \langle \overline{u'w'} \rangle - \rho \langle \bar{u}'' \overline{w}'' \rangle, \qquad (2)$$

At the bed (z = 0), for smooth and impermeable beds, the no-slip condition requires the second and third terms in (2) to be zero, such that  $\langle \bar{\tau} \rangle |_{z=0} = \rho v \frac{\partial \langle \bar{u} \rangle}{\partial z} \Big|_{z=0}$ ; therefore, the spatially-averaged bed shear stress is simply:

291 
$$\langle \bar{\tau}_b \rangle = \rho v \frac{\partial \langle \bar{u} \rangle}{\partial z} \Big|_{z=0}$$
 (3)

292 The temporally- and spatially-averaged turbulent kinetic energy is defined as:

293 
$$\langle \overline{k_t} \rangle = 0.5 (\langle \overline{u'^2} \rangle + \langle \overline{v'^2} \rangle + \langle \overline{w'^2} \rangle), \qquad (4)$$

where u', v', and w' are the turbulent velocity fluctuations in streamwise, spanwise, and vertical directions, respectively. Given that the PIV measurements do not record the spanwise component of the velocity, we are forced to assume that  $\overline{u'^2} \approx \overline{v'^2}$  (as in Tanino & Nepf (2008a)), such that the turbulent kinetic energy for emergent canopies can be approximated as:

298 
$$\langle \overline{k_t} \rangle \approx 0.5 (2 \langle u'^2 \rangle + \langle w'^2 \rangle).$$
 (5)

In this study, we will show the existence of a clear relationship between the BBL thickness and the near-bed TKE (section 3.2). Moreover, we will develop a TKE-based model for the BBL thickness and, ultimately, the bed shear stress (Section 4.1). Therefore, a predictive formulation for TKE is needed. Here, we employ the model of Tanino and Nepf (2008) for the turbulence intensity within an array of emergent cylinders:

304 
$$\bar{k}_{t} = \gamma^{2} \left( C_{D} \frac{l_{t}}{d} \frac{2\phi}{(1-\phi)\pi} \right)^{2/3} U_{p}^{2}, \tag{6}$$

In (6),  $\gamma$  is an empirical constant,  $C_D$  is the element drag coefficient, and it was estimated as  $C_D = 1 + Re_c^{2/3}$ , where  $Re_c = (U_p d/\nu)((1 - \phi)/(1 - \sqrt{2\phi/\pi}))$  (Etminan et al., 2017). The eddy length-scale  $l_t$  is indicative of the scale associated with the mixing due to turbulent eddies (Tennekes & Lumley, 1972; Tanino & Nepf, 2008a).

#### 309 2.5.2. Numerical data

To provide a statistical measure of the spatial variability of the velocities, total stress, and turbulent kinetic energy within the BBL for different canopy densities and flow conditions, the ratio between the RMS value ( $\bar{\zeta}_{RMS}$ ) and the temporally- and spatially-averaged value ( $\langle \bar{\zeta} \rangle$ ) was calculated for all numerical runs. Given the full 3D spatial coverage of the LES simulations, robust

estimates of spatially-averaged quantities can be obtained from the numerical model. Furthermore,

all three components of the TKE could be measured directly, as per Eq. (4). For a given flow

variable ( $\zeta$ ), the horizontally-averaged root-mean-squared (RMS) value ( $\zeta_{RMS}$ ) is calculated as

317 
$$\zeta_{RMS}(z) = \sqrt{\frac{1}{n} \sum \left[ \bar{\zeta}(x, y, z) - \langle \bar{\zeta} \rangle(z) \right]^2}, \qquad (7)$$

where  $\bar{\zeta}$  is the temporally-averaged value,  $\langle \bar{\zeta} \rangle$  is the temporally- and spatially-averaged value, and *n* is the number of elements used in the sum. Flow statistics recorded in the region immediately adjacent to the cylinders (see figure 4 for further detail) were excluded from calculations of spatial averages as they are not representative of values at the canopy scale. The diameter of this excluded area ( $d_{exc}$ ) was defined in Etminan et al. (2018) and is reported in Table 3.

Density,	Diameter of the excluded area,
φ	$d_{exc}/d$
0.016	2.5
0.04	2.5
0.08	2.0
0.12	1.8
0.20	1.6
0.25	1.6

 Table 3. Excluded area diameter of the circle around the elements.

Finally, the mean absolute percentage deviation (*M*) was calculated to identify single-point locations that most closely match the spatial average value of the flow characteristics for a staggered-ordered emergent array of canopies:

327 
$$M(x,y) = \frac{100}{m} \sum_{1}^{m} \left| \frac{\overline{\zeta}(x,y,z) - \langle \overline{\zeta} \rangle(z)}{\langle \overline{\zeta} \rangle(z)} \right|, \tag{8}$$

328 where m is the number of vertical values used in the sum.

# 329 **3 Results**

# 330 **3.1 Horizontal variability of the flow characteristics within the BBL**

Direct local estimates of near-bed flow characteristics from numerical simulations show 331 variations of the same order as the temporally and spatially averaged values, revealing the high 332 spatial variability at the element scale (Figure 4). The highest values of BBL thickness ( $\delta(x, y)$ ) 333 334 and temporally averaged streamwise velocity  $(\bar{u}(x, y))$  are mostly confined to the area between element rows, whereas the smallest values are found in the element wakes (Figure 4a,b). Such 335 336 regions of diminished and elevated velocities are a direct consequence of the sheltering and 337 channelling effect produced by the canopy elements. Conversely, the highest values of local turbulent kinetic energy ( $\bar{k}_t(x, y)$ ) are found in cylinder wakes with the smallest values in the 338 channel between rows (Figure 4c). This variability indicates the major influence of element wakes 339 on local values of  $\bar{k}_t$  and, ultimately, on the temporally- and spatially-averaged value  $(\langle \bar{k}_t \rangle)$ . 340





Figure 4. Horizontal variability of the dimensionless (a) BBL thickness, along with (b) streamwise velocity; (c) turbulent kinetic energy and (d) total stress, all vertically-averaged over the local BBL thickness  $\delta(x, y)$  for the numerical case with  $Re_{p,d} = 500$  and density  $\phi = 0.04$ . The flow is from left to right. The dashed line represents the area excluded from horizontal average calculations. Over a scale of an element diameter, variations of more than twice the temporally- and spatially- averaged value can be observed in near-bed flow characteristics, indicating the significant horizontal variability of flow through emergent vegetation. As negative values of streamwise velocity and total stress are possible, the lower limit of the colour bar differs between panels.

349

The horizontal variability of the total stress reveals areas of extreme stress  $(\bar{\tau}(x, y)/\langle \bar{\tau} \rangle \gg$ 350 2) on the sides of the cylinders (Figure 4d). These areas of enhanced stress are mainly generated 351 by the compression of streamlines around canopy elements. Moreover, local values of negative 352 stresses can be observed in the areas directly upstream and downstream of the cylinders. Regions 353 354 of negative stress have been previously observed around individual cylinders, and they are linked to the development of horseshoe vortices (upstream region) and flow recirculation in the wake 355 356 region (downstream region) (Schanderl et al., 2017). Further evidence of flow recirculation in the wake is seen in negative values of streamwise immediately downstream of the cylinders (Figure 357 358 4b).

Additionally, far from the region adjacent to the cylinders (outside the dashed line circle), 359 the local stress values in the middle channel are close to the spatially-averaged value, and areas of 360 361 reduced stress are observed in the cylinder wakes (white and blue colour in Figure 4d). The maps of near-bed flow characteristics in Figure 4 allow clear spatial correlations to be identified. For 362 instance, locations of BBL thickness and streamwise velocity are positively correlated, whereas 363 locations of BBL thickness and TKE are negatively correlated. The observed interplay between 364 365 these variables indicates the possible impact of the streamwise velocity and near-bed TKE in 366 controlling the BBL thickness, which we address in the next section.

The influence of canopy density and Reynolds number (defined as  $Re_{p,d} = U_p d/\nu$ ) in the 367 horizontal variability of near-bed flow characteristics is presented in Figure 5. Monotonical 368 increases in RMS values of bottom boundary layer thickness, streamwise velocity and total stress 369 370 (normalized by the temporally- and spatially-averaged values) are observed with canopy density 371 (Figure 5a,b,d). This increase in horizontal variability with density is generated by stronger channelling and sheltering effects in denser vegetation, confirming that the element configuration 372 largely controls spatial variability of the near-bed flow characteristics. In contrast, the horizontal 373 374 variability of near-bed TKE decreases with canopy density (Figure 5c). This decrease in dense canopies results from an overall increment of the wake turbulence produced by the larger number 375 376 of elements per unit area.

Finally, the magnitude of normalised RMS values of near-bed flow characteristics (visualised in Figure 4, quantified in Figure 5) highlights the difficulties associated with understanding near-bed hydrodynamic processes from single-point observations. Numerical models that can resolve this variability, or at the very least provide an indication of the measurement locations required to capture it, are helpful tools in understanding spatial variation in the near-bed flow and have been employed here.



Figure 5. The ratio between the horizontally-averaged RMS values and the temporally- and spatiallyaveraged values of (a) boundary layer thickness, (b) streamwise velocity, (c) turbulent kinetic energy, and d) total stress. The canopy density largely controls the horizontal variability of the near-bed flow characteristics.

387

7 **3.2** The bottom boundary layer thickness

In this section, experimental and numerical data are integrated to quantify the thickness of the bottom boundary layer across different canopy densities and highlight how turbulence levels regulate this thickness. Bottom boundary layer thickness with different canopy and flow properties

showed that this layer is primarily controlled by the canopy density (decreasing on denser 391 canopies) and the element diameter (Figure 6a). This result is consistent with previous 392 393 observations of the linear-stress region (a sublayer within the BBL) decreasing with canopy density (Yang et al., 2015; Etminan et al., 2018). The Reynolds number has a secondary influence on the 394 BBL thickness (BBL thickness decrease with Reynolds number), but it is generally minor 395 compared to the impact of the canopy density. It can be observed from Figure 6a that as the canopy 396 density increases, the boundary layer thickness becomes less sensitive to changes in density. 397 Conversely, the dimensionless spatially-averaged TKE  $(\langle \bar{k}_t \rangle / \langle \bar{u} \rangle^2)$  monotonically increases with 398 canopy density (Figure 6b), as a result of the increase in wake turbulence produced by a larger 399 400 number of cylinders per unit area.



401

Figure 6. (a) The strong correlation between the dimensionless spatially-averaged BBL thickness  $(\langle \tilde{\delta} \rangle / d)$ and the canopy density ( $\phi$ ). There is strong agreement between numerical and experimental estimates of BBL thickness. The solid line represents the line of best fit,  $\langle \tilde{\delta} \rangle / d = 0.06/\phi^{0.5}$ . (b) variation of dimensionless spatiallyaveraged turbulence intensity  $\langle \bar{k}_t \rangle / \langle \bar{u} \rangle^2$  with the canopy density for the experimental and numerical data. The solid black line represents the curve given by Eq. 6.

407 As suggested in Figure 6, there is a strong negative correlation between the dimensionless 408 spatially-averaged thickness of the BBL and the dimensionless temporally- and spatially-averaged 409 turbulent kinetic energy at the top of the BBL ( $z = \langle \tilde{\delta} \rangle$ ) (Figure 7). This relationship takes the 410 form

411 
$$\frac{\langle \tilde{\delta} \rangle}{d} = 0.022 \frac{\langle \bar{u} \rangle^2}{\langle \bar{k}_t \rangle} \bigg|_{z = \langle \tilde{\delta} \rangle}.$$
 (9)

412



413

Figure 7. The inversely proportional relationship between the dimensionless spatially-averaged BBL thickness  $(\langle \tilde{\delta} \rangle / d)$  and dimensionless TKE  $(\langle \bar{k}_t \rangle / \langle \bar{u} \rangle^2)$  at the top of the bottom boundary layer. The blue markers represent numerical data and the red markers experimental data. The solid black line represents the best fit curve given by Eq. 9. The inversely proportional relationship supports the notion of a direct influence of the TKE on the thickness of the BBL for the spatially-averaged values.

In addition to the correlation between spatially-averaged values, there is a strong spatial correlation between BBL thickness and near-bed TKE within a given canopy (Figure 8). Notably, even though there is some scatter, most of the locally-measured data (red patches) follow the trends of the spatially-averaged value (the line in Figure 8, shown in Eq. 9).

The inversely proportional relationship of boundary layer thickness and turbulence intensity with canopy density suggests the possibility of a causal link between these two variables, where the increase in the vertical moment transport generated by the 'external' turbulence from the plants creates a thinner boundary layer. It is noteworthy that within each patch, where the density is not changing, the local values of BBL thickness and the near-bed TKE have an inverse

relationship within a given flow, i.e., local values of high TKE generates a thinner local BBL. This
observed local agreement further indicates a causal link between near-bed TKE and BBL
thickness.

431



#### 432

Figure 8. Local values of dimensionless bottom boundary layer thickness as a function of the dimensionless turbulent kinetic energy (evaluated at the top of the BBL) for three canopy densities = 0.016,  $\phi$  = 0.08,  $\phi$  = 0.25 and  $Re_{p,d}$  = 500. The solid black line represents the best fit curve given by Eq. 9. The colour bar indicates the data density normalized by the total data in the domain.

#### 437 **3.3 Vertical structure of stresses within BBL**

438 The significant spatial variability of the near-bed flow characteristics (Figure 4) indicates that dispersive stresses may contribute significantly to momentum transport in the near-bed region. 439 440 The vertical structure of the stress components (terms *i*, *ii* and *iii* in Eq. 1) and the total shear stress within the BBL are presented in Figure 9 for a flow through a canopy with a density of 0.04. This 441 442 vertical structure defines three important sublayers within the BBL. In the sublayer closest to the bed,  $z \leq \langle \tilde{\delta} \rangle / 4$  (ranging from 0.25-1.5 mm in this study), the viscous stress is the dominant 443 component of the total shear stress, and the Reynolds and dispersive stresses are both negligible. 444 The second sublayer, in which all three stress components are significant, lies between  $\langle \tilde{\delta} \rangle/4$  and 445

446  $3\langle \tilde{\delta} \rangle/4$ . Finally, within the third sublayer  $3\langle \tilde{\delta} \rangle/4 \lesssim z \lesssim \langle \tilde{\delta} \rangle$ , the dispersive and Reynolds stresses 447 dominate. Note that this vertical structure of the normalized stresses within the bottom boundary 448 layer was qualitatively similar in shape across all densities and flows. Of particular significance is 449 that the BBL in emergent canopies is a region of roughly constant total shear stress related to the 450 bed shear stress.



▲ Dispersive Stress ▲ Reynolds Stress ▲ Viscous stress ▲ Total stress

451

Figure 9. Vertical structure of spatially averaged stress components within the BBL for the numerical case with density  $\phi = 0.04$  and  $Re_{p,d} = 500$  (for which  $\langle \tilde{\delta} \rangle = 0.34$  cm). Stress components have been normalized by  $\langle \bar{\tau}_b \rangle$ , the value at the bed (z = 0). Roman numerals define three important sublayers across the BBL. Although the contributions of the components vary with height above the bed, the BBL in emergent canopies is seen to be a region of approximately constant shear stress.

457 The vertical variation of the stress components emphasises the fact that direct measurements of bed shear stress require either velocity gradient measurements within the region 458 459 adjacent to the bed (Region I) or measurements across the horizontal plane to accurately capture 460 the dispersive stresses. Either requirement is likely to prove impractical in experimental studies. Therefore, we propose a simple scaling approach to estimate the spatially-averaged bed shear 461 stress in emergent canopies. Assuming self-similarity of the velocity profile in the BBL, the 462 velocity gradient at the bed scales upon  $\bar{u}|_{z=\delta}/\delta$  and therefore, it is possible to approximate the 463 temporally- and spatially- averaged bed shear stress (Eq. 3) as: 464

465 
$$\langle \bar{\tau}_b \rangle = \mu \frac{\partial \langle \bar{u} \rangle}{\partial z} \Big|_{z=0} \sim \mu \frac{\langle \bar{u} \rangle|_{z=\langle \tilde{\delta} \rangle}}{\langle \tilde{\delta} \rangle}.$$
(10)

There is a strong agreement ( $R^2 = 0.89$ ) between measured bed shear stresses and those 466 467 predicted by the scaling relationship in Eq. 10 (Figure 10). This strong agreement supports the self-similarity assumption of the mean velocity profile in the BBL as well as the validity of Eq. 10 468 in predicting bed shear stress in aquatic vegetation. Despite the strong agreement, the model 469 underestimates the bed shear stress at low canopy densities ( $\phi < 0.04$ ) and high Reynolds 470 numbers ( $Re_{p,d}$ >1000). Ultimately, the validity of Eq. 10 depends upon a self-similarity of mean 471 velocity profiles in the BBL of vegetated flows, a self-similarity which appears to break down at 472 low canopy density and high Reynolds numbers. Thus, this underestimation can be explained by, 473 at low density and high Reynolds number, there is a reduced impact of the vegetation on the mean 474 475 velocity structure; in the limit of zero density, the vertical velocity gradient will tend towards that observed in flow over a bare bed. Indeed, in low-density canopies, the thicknesses of the viscous-476 stress-dominated sublayer (Layer I in Figure 9) in vegetated  $(H_v)$  and unvegetated  $(H_{v,unveg})$  beds 477 are equivalent (Yang et al., 2015). The scaling relationship in Eq. 10 was modified to reflect this 478 change in velocity structure at low canopy density: 479

480 
$$\langle \bar{\tau}_b \rangle = \mu \frac{\partial \langle \bar{u} \rangle}{\partial z} \Big|_{z=0} \sim \mu \frac{\langle \bar{u} \rangle|_{z=\tilde{\delta}_e}}{\tilde{\delta}_e},$$
 (11)

where  $\tilde{\delta}_e$  is the effective BBL thickness given by  $\tilde{\delta}_e = \min(\langle \tilde{\delta} \rangle, H_{v,unveg})$  with  $H_{v,unveg}$  taken as 25 $\nu/\langle \bar{u}_* \rangle$ , as for bare channel flows,  $H_{v,unveg}\langle \bar{u}_* \rangle \rangle \nu = 25$  (see Figure 2 in Yang et al., 2015). When considering the effective BBL thickness as the length scale that governs the velocity gradient at the bed (i.e. Eq. 11), the prediction of bed shear improves significantly ( $R^2 = 0.96$ , Figure 10b).



485

Figure 10. Agreement between measured bed shear stresses and those predicted by the scaling relationship in (a) Eq.10, the solid line represents the linear fit  $\langle \bar{\tau}_b \rangle = 2.0 \mu \langle \bar{u} \rangle |_{z=\langle \bar{\delta} \rangle} / \langle \bar{\delta} \rangle$ , and (b) Eq. 11, the solid line represents the linear fit  $\langle \bar{\tau}_b \rangle = 2.0 \mu \langle \bar{u} \rangle |_{z=\tilde{\delta}_e} / \tilde{\delta}_e$ . The blue markers represent the numerical data, and the red markers represent experimental data. The blue and red colour bar indicates the  $Re_{p,d}$  for numerical and experimental data, respectively.

#### 490 **4. Discussion**

## 491 **4.1. Developing a predictive model of bed shear stress**

Given the high correlation between  $\langle \bar{\tau}_b \rangle$  and  $\mu \langle \bar{u} \rangle |_{z=\tilde{\delta}_e} / \tilde{\delta}_e$  (Figure 10b), a model to predict the bed shear stress can be obtained from Eq.11. Defining the friction velocity  $\langle \bar{u}_* \rangle$  through  $\langle \bar{\tau}_b \rangle = \rho \langle \bar{u}_* \rangle^2$ , and substituting  $\langle \bar{u} \rangle |_{z=\tilde{\delta}_e} = U_p$  and  $\tilde{\delta}_e = \min(\langle \tilde{\delta} \rangle, H_{v,unveg})$  into Eq. 11, a model for friction velocity ( $\langle \bar{u}_* \rangle_{mod}$ ) can be written as  $\langle \bar{u}_* \rangle_{mod} = \max\left(\sqrt{2\nu U_p}/\langle \tilde{\delta} \rangle, \sqrt{2\nu U_p}/H_{v,unveg}\right)$ . Using the TKE-dependent relationship for BBL thickness in Eq. 9,  $\sqrt{2\nu U_p}/\langle \tilde{\delta} \rangle$  can be rewritten as:

498 
$$C_{\sqrt{\frac{\langle \bar{k}_t \rangle |_{z=\langle \tilde{\delta} \rangle}}{Re_{p,d}}}},$$
 (12)

499 with C = 8. Furthermore, by characterising  $H_{v,unveg}$  in terms of a bed drag coefficient, 500  $C_f$  (defined as  $\langle \bar{u}_* \rangle^2 / U_p^2$ ), such as  $H_{v,unveg} = 2\nu/C_f U_p$ , we can express  $\sqrt{2\nu U_p/H_{v,unveg}}$  as:

501 
$$\sqrt{C_f} U_p.$$
 (13)

502 Therefore, the model for the friction velocity becomes

503 
$$\langle \bar{u}_* \rangle_{mod} = \max\left(C\sqrt{\frac{\langle \bar{k}_t \rangle |_{z=\langle \tilde{\delta} \rangle}}{Re_{p,d}}}, \sqrt{C_f}U_p\right).$$
 (14)

This model (Eq. 14) requires *a priori* knowledge of the temporally- and spatially-averaged TKE at the top of the BBL  $(\langle \bar{k}_t \rangle |_{z = \langle \bar{\delta} \rangle})$  and the bed drag coefficient ( $C_f$ ). The model for predicting TKE in Eq. 6 can be employed here for estimating  $\langle \bar{k}_t \rangle |_{z = \langle \bar{\delta} \rangle}$ ; if we assumed Eq 6 to be valid at every vertical position (as in Xu & Nepf (2020)), then

508 
$$\langle \bar{k}_t \rangle \Big|_{z=\langle \tilde{\delta} \rangle} = \gamma^2 \left( C_d \frac{\langle l_t \rangle}{d} \frac{2\phi}{(1-\phi)\pi} \right)^{2/3} \langle \bar{u} \rangle \Big|_{z=\langle \tilde{\delta} \rangle}^2.$$
 (15)

Given the wide range of densities in this study, the characteristic eddy length scale  $(l_t)$  depends on the ratio of element diameter (d) to element spacing  $(s_n)$ . Following Tanino & Nepf (2008), we used a length scale of  $\langle l_t \rangle = d$  for  $d/s_n < 0.53$  and  $\langle l_t \rangle = s_n$  for  $d/s_n \ge 0.53$ . The velocity at the top of the BBL  $\langle \bar{u} \rangle |_{z=\langle \tilde{\delta} \rangle}$  was replaced by the cross-sectional average fluid velocity,  $U_p$  (Figure 3b) The empirical coefficient  $\gamma^2$  in Eq. 15 was estimated from the least-squares fitting between  $\langle \bar{k}_t \rangle |_{z=\langle \tilde{\delta} \rangle}$  and  $\gamma^2 \left( C_d \frac{\langle l_t \rangle}{d} \frac{2\phi}{(1-\phi)\pi} \right)^{2/3} \langle \bar{u} \rangle |_{z=\langle \tilde{\delta} \rangle}^2$  using both experimental and numerical data (Figure 11).





517 Figure 11. Least-squares fitting between the measured value of TKE at the top of BBL vs that predicted by 518 Eqn. 13. The solid black line represents the line of best fit given by  $y = \gamma^2 x$  with  $\gamma^2 = 0.70$ . Markers are as in Figure 519 7.

520 The bed drag coefficient required in Eq. 14 was found using the relation for  $H_{v,unveg}$ 521  $(H_{v,unveg}\langle \bar{u}_* \rangle \setminus v = 25)$  found by Yang et al. (2015), leading to

522 
$$C_f = \frac{2\nu}{H_{v,unveg}U_p} = 0.08 \frac{\langle \bar{u}_* \rangle}{U_p}.$$
 (16)

Since by definition  $C_f = \langle \bar{u}_* \rangle^2 / U_p^2$ , it must therefore take a value of approximately 0.0064. Substituting  $\langle \bar{k}_t \rangle |_{z=\langle \tilde{\delta} \rangle}$ ,  $C_f$ ,  $Re_{p,d}$  and  $U_p$  in Eq.14, the agreement between the measured  $\langle \bar{u}_* \rangle$  and modelled friction velocities ( $\langle \bar{u}_* \rangle_{mod}$ ) is shown in figure 12. The strong agreement ( $R^2 = 0.90$ ) between measured and modelled (Eq. 14) values of friction velocity (Figure 12) demonstrates the robust predictive capacity of the model, which only requires bulk properties of the system (i.e., the pore velocity, the canopy density and the bed drag coefficient) as inputs.





531 Figure 12. Comparison between measured spatially-averaged values of friction velocity with those predicted 532 from Eq. 14 ( $\langle \bar{u}_* \rangle_{mod}$ ). The solid line represents perfect agreement. Markers are as in Figure 7.

533

# **4.2.** Implications for interpreting experimental measurements of flow properties

Given the significant spatial variability of near-bed flow characteristics in vegetation canopies, it is paramount to identify the locations that can be targeted in experimental studies to provide representative values of the spatially-averaged values. The horizontal locations where the mean absolute percentage deviation (M, Eqn. (8)) for streamwise velocity ( $\bar{u}$ ), turbulent kinetic energy ( $\bar{k}_t$ ) and total stress ( $\bar{\tau}$ ) is less than 5% within the BBL are shown in Figure 13.

The horizontal distributions of  $\bar{u}$ ,  $\bar{k_t}$  and  $\bar{\tau}$  show that the regions with significant deviations 540 from spatial means (white regions in Figure 13) are concentrated around the wakes of the elements 541 542 and are found to vary significantly with canopy density. As the density increases, these regions become narrower (relative to S/d), mainly due to the intensification of the flow channelling 543 544 created by a staggered-ordered canopy model, along with the more complex wake interactions produced in dense canopies. Furthermore, in the near-bed region, the average deviation from 545 spatially-averaged values (M > 5%) across the total plan area for the bed shear stress is enhanced 546 due to the strong interaction of the flow with the bed. As in this region, the local changes of near-547

# bed TKE induce changes in the BBL thickness (Figure 8), which in turn generates strong spatial variability of bed shear stress.



550

Figure 13. Horizontal distribution of locations with  $M \le 5\%$  for the streamwise velocity,  $\bar{u}$  (red markers), the turbulent kinetic energy,  $\bar{k}_t$  (blue markers) and the total stress,  $\bar{\tau}$  (yellow markers) within the BBL ( $z \le \langle \delta \rangle$ ) for three canopy densities. The dashed circle around the cylinders represents the excluded area for all the statistics. Spatially representative statistics can be obtained in the region encompassed by a measurement volume that extends from one side of an upstream cylinder to the same side of a downstream cylinder in the same row.

556 While the previous maps of *M* suggest the existence of local preferred measurement regions 557 (low values of *M*), we caution that their generalization to more heterogeneous and complex 558 canopies requires further research to understand the spatial variability of the flow characteristic. 559

#### 560 **4.3 Implications for predicting sediment transport**

561 Current models can use either stress or TKE as an assumed driver of sediment motion & 562 transport. On the one hand, models based on TKE as the main driver of sediment transport (e.g. 563 Tinoco & Coco, 2014, 2016, 2018; Yang et al., 2016; Yang & Nepf, 2018, 2019) suggest that bed 564 shear stress-based sediment transport models (developed for bare beds) fail within vegetated 565 regions because they do not account for the vegetation generated turbulence. In other words, the

'external' turbulence from the plants and the bed shear stress are decoupled in these flows. 566 However, our results (Figures 7, 8 and 12) indicate the opposite, i.e., higher values of vegetated 567 568 generated TKE create a thinner BBL thickness, which in turn increases the bed shear stress. The strong link between the bed shear stress and the near-bed TKE may explain why predictive models 569 using TKE are seen to excel in predicting sediment transport in vegetation canopies. On the other 570 571 hand, previous approaches based on predicting the thickness of the viscous-stress dominated sublayer within the BBL (e.g. Nepf, 2012b; Yang et al., 2015) have taken it to be proportional to 572 the element diameter, d, such that  $\overline{\tau} = \mu U/d$ , (with U the depth-averaged velocity). As a result, 573 these models are unable to capture a dependence of boundary layer thickness on canopy density. 574 The results of this study suggest, however, that for constant element diameter, both the viscous 575 sublayer and the total BBL thickness vary significantly with canopy density (Figure 6), which may 576 577 be one reason why robust sediment transport predictions for vegetation canopies using the bed shear stress have proven elusive. 578

In section 4.1 (Eq.14), a new method (validated in Figure 12) is proposed to estimate bed 579 580 shear stresses in emergent canopies based on the thickness of the boundary layer, the bed drag 581 coefficient and the temporally- and spatially-averaged velocity. This model incorporates the effect 582 of the near-bed TKE on the bed shear stress. Notably, the model presented in this study (Eq.14) 583 represents an alternative method for bed shear stress prediction, without requiring high-resolution flow measurements near the bed, by directly linking the bed shear stress with the near-bed turbulent 584 kinetic energy. Finally, further research is needed to precisely assess the capacity of this model to 585 586 drive predictions of the onset of sediment motion and transport rates in vegetated environments. 587

#### 588 **5 Conclusions**

Flows through emergent canopies are characterised by significant horizontal variability in the thickness of the bottom boundary layer, velocities, stresses, and turbulent kinetic energy within the layer. It is shown here that canopy density and element diameter control the boundary layer thickness through its direct control of near-bed turbulent kinetic energy (TKE). Locally and in a given flow, the near-bed turbulent kinetic energy (TKE) was negatively correlated with BBL thickness across the studied flow conditions. Accordingly, a model for prediction of bed shear stress is presented here that is based upon evaluation of this BBL thickness, and requires only

596 canopy density, bed drag coefficient and bulk velocity as further inputs. There is excellent 597 agreement between model predictions and direct bed shear stress measurements in laboratory and 598 numerical experiments. The link between 'external' turbulence from the plants, the BBL thickness, 599 and ultimately the bed shear stress may explain why stress and TKE may be seen as drivers of 500 sediment transport processes in vegetated flows.

# 601 6 Acknowledgments

This project forms part of a PhD study by M.C. at the University of Western Australia supported by the Commonwealth Government through an Australian Government Research Training Program Scholarship. Additional support was also provided by the Australian Research Council (Discovery Project DP200101545 and DP170100802).

# 606 7 References

- Abdolahpour, M., Ghisalberti, M., McMahon, K., & Lavery, P. S. (2018). The impact of flexibility on
  flow, turbulence, and vertical mixing in coastal canopies. *Limnology and Oceanography*, *63*(6),
  2777–2792. https://doi.org/10.1002/lno.11008
- Beer, D. De, & Kühl, M. (2001). Interfacial microbial mats and biofilms. *The Benthic Boundary Layer*.
  https://doi.org/10.1128/AEM.70.11.6551
- 612 Cundy, A. B., Hopkinson, L., Lafite, R., Spencer, K., Taylor, J. A., Ouddane, B., et al. (2005). Heavy
- 613 metal distribution and accumulation in two Spartina sp.-dominated macrotidal salt marshes from the
- 614 Seine estuary (France) and the Medway estuary (U.K.). *Applied Geochemistry*, 20(6), 1195–1208.
- 615 https://doi.org/10.1016/j.apgeochem.2005.01.010
- 616 Etminan, V., Lowe, R., & Ghisalberti, M. (2017). A new model for predicting the drag exerted by

617 vegetation canopies. *Water Resources Research*, *53*(4), 3179–3196.

- 618 https://doi.org/10.1002/2016WR020090
- Etminan, V., Ghisalberti, M., & Lowe, R. J. (2018). Predicting Bed Shear Stresses in Vegetated Channels. *Water Resources Research*, 54(11), 9187–9206. https://doi.org/10.1029/2018WR022811
- 621 Grant, S. B., Gomez-Velez, J. D., & Ghisalberti, M. (2018). Modeling the Effects of Turbulence on
- 622 Hyporheic Exchange and Local-to-Global Nutrient Processing in Streams. *Water Resources*
- 623 *Research*. https://doi.org/10.1029/2018WR023078
- Hemminga, M. A., & Duarte, C. M. (2000). Fauna associated with seagrass systems. In Seagrass Ecology

- 625 (pp. 199–247). Cambridge University Press. https://doi.org/10.1017/CBO9780511525551.007
- Hooke, J. M. (2007). Monitoring morphological and vegetation changes and flow events in dryland river
- 627 channels. *Environmental Monitoring and Assessment*, 127(1–3), 445–457.
- 628 https://doi.org/10.1007/S10661-006-9294-6
- James, C. S., Jordanova, A. A., & Nicolson, C. R. (2002). Flume experiments and modelling of flowsediment-vegetation interactions. *IAHS-AISH Publication*.
- Jordanova, A. a., & James, C. S. (2003). Experimental Study of Bed Load Transport through Emergent
  Vegetation. *Journal of Hydraulic Engineering*, *129*(6), 474–478.
  https://doi.org/10.1061/(ASCE)0733-9429(2003)129:6(474)
- van Katwijk, M. M., Bos, A. R., Hermus, D. C. R., & Suykerbuyk, W. (2010). Sediment modification by
- 635 seagrass beds: Muddification and sandification induced by plant cover and environmental
- 636 conditions. *Estuarine, Coastal and Shelf Science*, 89(2), 175–181.
- 637 https://doi.org/10.1016/J.ECSS.2010.06.008
- Kothyari, U. C., Hashimoto, H., & Hayashi, K. (2009). Effect of tall vegetation on sediment transport by
  channel flows. *Journal of Hydraulic Research*, 47(6), 700–710.
  https://doi.org/10.3826/jhr.2009.3317
- Larsen, L. G., Harvey, J. W., & Crimaldi, J. P. (2009). Predicting bed shear stress and its role in sediment
  dynamics and restoration potential of the Everglades and other vegetated flow systems. *Ecological Engineering*, *35*(12), 1773–1785. https://doi.org/10.1016/j.ecoleng.2009.09.002
- Liu, D., Diplas, P., Fairbanks, J. D., & Hodges, C. C. (2008). An experimental study of flow through rigid
  vegetation. *Journal of Geophysical Research: Earth Surface*, *113*(4), 1–16.
  https://doi.org/10.1029/2008JF001042
- Lowe, R., Koseff, J. R., & Monismith, S. G. (2005). Oscillatory flow through submerged canopies: 1.
  Velocity structure. *Journal of Geophysical Research C: Oceans*, *110*(10), 1–17.
- 649 https://doi.org/10.1029/2004JC002788
- Lowe, R., Falter, J. L., Koseff, J. R., Monismith, S. G., & Atkinson, M. J. (2007). Spectral wave flow
  attenuation within submerged canopies: Implications for wave energy dissipation. *Journal of*
- 652 *Geophysical Research*, *112*(C5), C05018. https://doi.org/10.1029/2006JC003605
- Massel, S. R., Furukawa, K., & Brinkman, R. M. (1999). Surface wave propagation in mangrove forests. *Fluid Dynamics Research*, 24(4), 219–249. https://doi.org/10.1016/S0169-5983(98)00024-0

- 655 Meinhart, C. D., Wereley, S. T., & Santiago, J. G. (2000). A piv algorithm for estimating time-averaged
- 656 velocity fields. *Journal of Fluids Engineering, Transactions of the ASME.*
- 657 https://doi.org/10.1115/1.483256
- Morse, A. P., Gardiner, B. A., & Marshall, B. J. (2002). Mechanisms controlling turbulence development
  across a forest edge. *Boundary-Layer Meteorology*. https://doi.org/10.1023/A:1014507727784
- Nepf, H. (1999). Drag, turbulence, and diffusion in flow through emergent vegetation. *Water Resources Research*, 35(2), 479–489. https://doi.org/10.1029/1998WR900069
- Nepf, H. (2012a). Flow and Transport in Regions with Aquatic Vegetation. *Annual Review of Fluid Mechanics*, 44(1), 123–142. https://doi.org/10.1146/annurev-fluid-120710-101048
- Nepf, H. (2012b). Hydrodynamics of vegetated channels. *Journal of Hydraulic Research*, 50(3), 262–279.
   https://doi.org/10.1080/00221686.2012.696559
- Nepf, H., & Ghisalberti, M. (2008). Flow and transport in channels with submerged vegetation. *Acta Geophysica*, 56(3), 753–777. https://doi.org/10.2478/s11600-008-0017-y
- Nepf, H., & Vivoni, E. (2000). Flow structure in depth-limited, vegetated flow. *Journal of Geophysical Research: Oceans*, *105*(C12), 28547–28557. https://doi.org/10.1029/2000JC900145
- 670 Nikora, V., Goring, D., McEwan, I., & Griffiths, G. (2001). Spatially Averaged Open-Channel Flow over
- 671 Rough Bed. *Journal of Hydraulic Engineering*, *127*(2), 123–133.
- 672 https://doi.org/10.1061/(ASCE)0733-9429(2001)127:2(123)
- Nobach, H., & Honkanen, M. (2005). Two-dimensional Gaussian regression for sub-pixel displacement
  estimation in particle image velocimetry or particle position estimation in particle tracking
  velocimetry. *Experiments in Fluids*. https://doi.org/10.1007/s00348-005-0942-3
- Nowell, A. R. M. M., & Jumars, P. A. (1984). Flow Environments of Aquatic Benthos. *Annual Review of Ecology and Systematics*, 15(1), 303–328. https://doi.org/10.1146/annurev.es.15.110184.001511
- Raffel, M., Willert, C. E., Wereley, S., & Kompenhans, J. (2012). *Particle Image Velocity A Practical Guide. Journal of visualized experiments : JoVE.* https://doi.org/10.3791/4265
- 680 Raupach, M. R., & Shaw, R. H. (1982). Averaging procedures for flow within vegetation canopies.
- 681 *Boundary-Layer Meteorology*, 22(1), 79–90. https://doi.org/10.1007/BF00128057
- van Rijn, L. C. (1987). Closure to "Sediment Transport, Part I: Bed Load Transport" by Leo C. van Rijn
  (October, 1984, Vol. 110, No. 10). *Journal of Hydraulic Engineering*.

- 684 https://doi.org/10.1061/(asce)0733-9429(1987)113:9(1189)
- van Rijn, L. C. (2007a). Unified view of sediment transport by currents and waves. I: Initiation of motion,
  bed roughness, and bed-load transport. *Journal of Hydraulic Engineering*.
- 687 https://doi.org/10.1061/(ASCE)0733-9429(2007)133:6(649)
- van Rijn, L. C. (2007b). Unified view of sediment transport by currents and waves. II: Suspended
  transport. *Journal of Hydraulic Engineering*. https://doi.org/10.1061/(ASCE)0733-
- **690** 9429(2007)133:6(668)
- Roche, K. R., Blois, G., Best, J. L., Christensen, K. T., Aubeneau, A. F., & Packman, A. I. (2018).
  Turbulence Links Momentum and Solute Exchange in Coarse-Grained Streambeds. *Water Resources Research*, 54(5), 3225–3242. https://doi.org/10.1029/2017WR021992
- Santiago, J. G., Wereley, S. T., Meinhart, C. D., Beebe, D. J., & Adrian, R. J. (1998). A particle image
  velocimetry system for microfluidics. *Experiments in Fluids*.
- 696 https://doi.org/10.1007/s003480050235
- Schanderl, W., Jenssen, U., & Manhart, M. (2017). Near-Wall Stress Balance in Front of a Wall-Mounted
  Cylinder. *Flow, Turbulence and Combustion*, 99(3–4), 665–684. https://doi.org/10.1007/s10494017-9865-3
- Shields, F. D., Bowie, A. J., & Cooper, C. M. (1995). Control of streambank erosion due to bed
   degradation with vegetation and structure. *JAWRA Journal of the American Water Resources Association*. https://doi.org/10.1111/j.1752-1688.1995.tb04035.x
- Stapleton, K. R., & Huntley, D. A. (1995). Seabed stress determinations using the inertial dissipation
   method and the turbulent kinetic energy method. *Earth Surface Processes and Landforms*, 20(9),
   807–815. https://doi.org/10.1002/ESP.3290200906
- Tanino, Y., & Nepf, H. (2008a). Lateral dispersion in random cylinder arrays at high Reynolds number.
   *Journal of Fluid Mechanics*, 600, 339–371. https://doi.org/10.1017/S0022112008000505
- 708 Tanino, Y., & Nepf, H. M. (2008b). Laboratory Investigation of Mean Drag in a Random Array of Rigid,
- Emergent Cylinders. *Journal of Hydraulic Engineering*, *134*(1), 34–41.
- 710 https://doi.org/10.1061/(ASCE)0733-9429(2008)134:1(34)
- 711 Tennekes, H., & Lumley, J. L. (1972, March 15). A First Course in Turbulence. The MIT Press.
- 712 https://doi.org/10.7551/mitpress/3014.001.0001

- 713 Thielicke, W., & Stamhuis, E. J. (2014). PIVlab Towards User-friendly, Affordable and Accurate
- Digital Particle Image Velocimetry in MATLAB. *Journal of Open Research Software*, 2.
  https://doi.org/10.5334/jors.bl
- 716 Tinoco, R., & Coco, G. (2014). Observations of the effect of emergent vegetation on sediment
- resuspension under unidirectional currents and waves. *Earth Surface Dynamics*, 2(1), 83–96.
- 718 https://doi.org/10.5194/esurf-2-83-2014
- Tinoco, R., & Coco, G. (2016). A laboratory study on sediment resuspension within arrays of rigid
  cylinders. *Advances in Water Resources*, *92*, 1–9. https://doi.org/10.1016/j.advwatres.2016.04.003
- 721 Tinoco, R., & Coco, G. (2018). Turbulence as the Main Driver of Resuspension in Oscillatory Flow
- Through Vegetation. *Journal of Geophysical Research: Earth Surface*, *123*(5), 891–904.
- 723 https://doi.org/10.1002/2017JF004504
- Trowbridge, J. H., & Lentz, S. J. (2018). The bottom boundary layer. *Annual Review of Marine Science*.
   https://doi.org/10.1146/annurev-marine-121916-063351
- Türker, U., Yagci, O., & Kabdaşli, M. S. (2006). Analysis of coastal damage of a beach profile under the
  protection of emergent vegetation. *Ocean Engineering*, *33*(5–6), 810–828.
  https://doi.org/10.1016/j.oceaneng.2005.04.019
- Vargas-Luna, A., Crosato, A., & Uijttewaal, W. S. J. (2015). Effects of vegetation on flow and sediment
   transport: Comparative analyses and validation of predicting models. *Earth Surface Processes and Landforms*, 40(2), 157–176. https://doi.org/10.1002/esp.3633
- Voermans, J. J., Ghisalberti, M., & Ivey, G. N. (2018). A Model for Mass Transport Across the SedimentWater Interface. *Water Resources Research*, 54(4), 2799–2812.
  https://doi.org/10.1002/2017WR022418
- Ward, P. D., Montgomery, D. R., & Smith, R. (2000). Altered river morphology in south africa related to
- the permian-triassic extinction. *Science (New York, N.Y.)*, 289(5485), 1740–1743.
- 737 https://doi.org/10.1126/SCIENCE.289.5485.1740
- Westerweel, J., & Scarano, F. (2005). Universal outlier detection for PIV data. *Experiments in Fluids*.
  https://doi.org/10.1007/s00348-005-0016-6
- 740 Wüest, A., & Lorke, A. (2003). Small-scale hydrodynamics in lakes. Annual Review of Fluid Mechanics,
- 741 *35*(Section 3), 373–412. https://doi.org/10.1146/annurev.fluid.35.101101.161220

- 742 Xu, Y., & Nepf, H. (2020). Measured and Predicted Turbulent Kinetic Energy in Flow Through Emergent
- 743 Vegetation With Real Plant Morphology. *Water Resources Research*, *56*(12), 1–20.
- 744 https://doi.org/10.1029/2020WR027892
- 745 Yalin, M. S. (1977). Mechanics of sediment transport.
- Yang, J., & Nepf, H. (2018). A Turbulence-Based Bed-Load Transport Model for Bare and Vegetated
  Channels. *Geophysical Research Letters*, 45(19), 10,428-10,436.
- 748 https://doi.org/10.1029/2018GL079319
- Yang, J., & Nepf, H. (2019). Impact of Vegetation on Bed Load Transport Rate and Bedform
  Characteristics. *Water Resources Research*, 55(7), 6109–6124.
- 751 https://doi.org/10.1029/2018WR024404
- Yang, J., Kerger, F., & Nepf, H. (2015). Estimation of the bed shear stress in vegetated and bare channels
  with smooth beds. *Water Resources Research*, *51*(5), 3647–3663.
- 754 https://doi.org/10.1002/2014WR016042
- Yang, J., Chung, H., & Nepf, H. (2016). The onset of sediment transport in vegetated channels predicted
  by turbulent kinetic energy. *Geophysical Research Letters*, 43(21).
- 757 https://doi.org/10.1002/2016GL071092