MODELLING OF PARTICULATE MATTER FATE ON URBAN HIGHWAY STORMWATER CONTROL SYSTEMS

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ABSTRACT

Anthropogenic activities, especially vehicular traffic, produce load of pollutants that accumulate on impervious surfaces. In highways, exhaust, automobile parts wear and lubricating parts along with heavy metals and other pollutants accumulated on the pavement surface during dry periods are the sources of pollution. Rainfall-runoff process promotes surface wash-off, contributing to stormwater pollutants' load. The cumulative load in runoff is normally expressed as an exponential model with a peak concentration at the beginning of the rainfall event known as the first flush, where most part of the load is washed off at the beginning of the event reaching an early peak. This consideration motivates often the use of a first flush storage tank (FFT) to treat stormwater discharge from highways that is then discharged into water courses without any other treatment. It is considered that the FFT would retain the most polluted part of the runoff; however, a weak first flush is observed in some rainfall events, especially for low flow rates. Also, the vehicular traffic occurring during the rainfall event serves as a continuous source. Therefore, the objective of this research is to compare the efficiency on particle removal of an FFT with other methods of stormwater treatment, in this case an infiltration-exfiltration system (IES) consisting of a gravel swale with porous asphalt surface, through hydrological modelling of six rainfall events measured on a highway in Cincinnati (USA). The results showed a compatible removal rate for both the FFT and IES for the six analysed events, consisting of three mass-limit events and three flow-limit events. Particle transport modelling could represent well the behaviour of the events and can be used as a tool to choose between systems, where after setting the particle removal efficiency, other factors can be considered, like cost and system area consumption. This research can be followed up with continuous rainfall simulations and using computational fluid dynamics (CFD) to model IES particle removal.

Keywords: first flush tank, infiltration–exfiltration system, permeable pavement, stormwater treatment, SUDs.

1 INTRODUCTION

Stormwater is an important source of pollutants into watercourses. Rainfall-runoff process promotes surface wash-off of the accumulated sediments during dry periods. In highways, exhaust, automobile parts wear and lubricating parts along with heavy metals and other pollutants are the sources of pollution. The chemical analysis of stormwater in comparison with wastewater showed a higher presence of heavy metals and comparable or higher values for BOD and chemical oxygen demand (COD). A study held in Hamilton County (Ohio, USA) compared the stormwater runoff with wastewater flow and found an approximately equivalent annual load of total suspended solids (TSS) and COD. Particulate matter (PM) plays the role of a vector for the transport of pollutants such as heavy metals, organics and nutrients, and they end up being washed out by stormwater, eventually reaching streams [1–3]. An optimal solution for stormwater management is treatment of its source, reducing the peak flow and volume within treatment plant limits and including systems that also treat stormwater [4–6]. Best management practices (BMPs) of stormwater and sustainable urban drainage systems (SUDs) include systems, components and practices that favour stormwater

management at its source, dealing with runoff peak flow, volume and treatment [7, 8]. This research compares the efficiency of particle removal using two different stormwater control systems, a first flush tank (FFT) and an infiltration–exfiltration system (IES) with a porous surface layer, using data measured from six single rainfall-runoff events in the highway Interstate 75 (I-75) in Cincinnati (Ohio, USA).

2 METHODOLOGY

2.1 Experimental site and rainfall events data

Rainfall-runoff samples were collected from an experimental site in Cincinnati that receives runoff contribution from a 15×20 m area in I-75. Hydrological data is presented in Table 1, which shows three high-intensity events (18/06/1996, 07/07/1996 and 08/08/1996) and three low-intensity events (17/10/1996, 25/11/1996 and 16/12/1996). The highest rainfall intensity and runoff volume was observed on the 07/07/1996 event (110.17 mm/h and 9643 L, respectively) and the lowest on the 25/11/1996 event (3.05 mm/h and 150 L, respectively). More details on the hydrological data can be found elsewhere [9, 10].

2.2 Hydrological modelling

Stormwater management model (SWMM) was used to simulate the hydrological process of rainfall runoff using kinematic wave [11–13]. Rainfall-runoff models consider hydrological losses by infiltration and depression storage, neglecting evapotranspiration and interception, from vegetation or vehicular traffic [14]. To take account of these hydrological abstractions not considered on the model, the percent of imperviousness and the depression storage were used to reach the best accordance between the model and the measured results.

2.2.1 PM transporting modelling

PM mass accumulates during dry periods in a process known as build-up, which is due to wash-off from surfaces by stormwater during rainfall-runoff events. PM mass accumulation is a function of previous dry period expressed in days (PDD) or hours (PDH) [15–17]. The

Event measured	Rainfall duration (min)	Runoff volume (L)	Rain depth (mm)	i _{max} a (mm/h)	i_avr (mm/h)	Event classification
18/06/1996	63	2779	11.3	55.08	15.1	Mass-limited,
07/07/1996	50	9643	40.4	110.17	45.96	high-runoff
08/08/1996	51	3877	14.1	91.44	18.95	volume events
17/10/1996	616	3693	29.1	18.36	3.04	
25/11/1996	150	216	3.1	3.05	1.13	Flow-limited,
16/12/1996	340	269	3.4	2.04	0.6	low-runoff volume events

Table 1: Hydrologic data from I-75 site [9].

^aMaximum rainfall intensity.

^bAverage rainfall intensity.

EPA SWMM code proposes an empirical exponential model for build-up accumulation according to eqns (1) and (2) [17]:

$$M_{a}(t) = \frac{Accu}{Disp} \cdot \left(1 - e^{-Disp \cdot t_{se}}\right)$$
(1)

$$t_{se} = t_{sr} + \frac{1}{\text{Disp}} \cdot \ln\left(\frac{\text{Accu}}{\text{Accu} - \text{Disp} \cdot M_{ar}}\right)$$
(2)

where $M_a(t)$ is the accumulated mass function of time (kg/ha), Accu is the accumulation coefficient rate expressed in kg/ha d, Disp is the dispersion coefficient expressed in d⁻¹ and t_{se} is the equivalent dry time expressed in days. The value of Accu normally associated with land use ranges from 5 to 35 kg/ha d [16]. This rate may also vary with the characteristics of the surroundings; in [18], similar rates for residential and commercial zones were observed. In literature, the dispersion parameter is estimated to range from 0.08 to 0.4 d⁻¹ and is normally used as calibration with experimental data [16]. The equivalent dry time (eqn (2)) considers a residual mass that remains on the watershed after a rainfall-runoff event and is calculated according to eqn (2), where t_{sr} is the real PDD and M_{ar} is the mass remaining (kg).

Rainfall-runoff process promotes surface wash-off of the accumulated sediments during dry periods. Previous research on paved urban watersheds demonstrated that the response of solid fraction transport is influenced by the type of event: mass-limited flow limit according to runoff intensity, duration and traffic measured during runoff [9, 12, 19, 20].

Mass-limited events can be described by an exponential model (eqn (3)), whereas flow-limited events are described by a linear model (eqn (4)) [20].

$$\Delta M_t = M_0 (1 - e^{-k_1 V_T}) \tag{3}$$

$$\Delta M_t = k_0 V_T \tag{4}$$

where ΔM_t is the cumulative pollutant mass delivered, k is the wash-off coefficient and V_T is the cumulative volume. The cumulative load on runoff for mass-limited events presents a peak concentration at the beginning of the rainfall event, known as the first flush, where most part of the load would be washed off at the beginning of the event, reaching an early peak [9]. Flow-limited events may present a weak first flush. The first flush behaviour can be identified by plotting the normalized cumulative flow volume and mass against the normalized elapsed time, where the first flush occurs when the mass curve is above the flow volume curve. The index event mean concentration (EMC) is often used to characterize concentration; however, it does not indicate temporal variations during events and cannot represent the first flush concept [9]. Typical EMC for TSS in urban areas ranges from 180 to 484.

Build-up and wash-off were modelled for the six registered events using eqns (1), (3) and (4) and SWMM using eqn (1) for build-up and EMC measured for each event for wash-up. The equivalent dry time (t_{se}) was not considered. The EMC measured for each single event can be found in [9, 10].

2.3 Stormwater control

2.3.1 First flush separation tank (FFT)

The first flush concept motivated the development of first flush storage tanks that partition pollutants on a runoff volumetric base when they reach an established accumulated

volume equal to the first flush storage capacity and would remove most PM mass. FFT is normally designed to retain the first 5 mm depth of a rainfall event or 50 m^3/ha_{imp} (volume for each hectare impermeable of contribution area). The system consists of a diverter that conveys runoff towards the tank until it reaches the full storage capacity. Then the runoff is conveyed directly to the drainage or sewer system. The FFT storage capacity has the most relevant role in pollutant removal on an annual basis (TSS basis), when compared with the number of tanks (one single tank downstream or multiple tanks per watershed) or their kind (transit or capture) [21]. The number of rainfall events with strong first flush will influence pollutants' removal from an FFT, as observed in a study held in Pavia (Italy) which concluded that this type of solution can be cost saving in stormwater management [22]. Partitioning and directing to the sewer system only the stormwater fraction with the highest contaminant concentration improves treatment plant performance [23]. Storage retention time can also promote load removal by sedimentation [24].

In this study, the TS removal efficiency for an FFT with a capacity of 1500 L was analysed using the hyetograph, accumulated volume and pollutograph for each registered event and modelled data with eqns (3) and (4) and SWMM.

2.3.2 Infiltration-exfiltration system

The IES consists of a porous surface and an aggregate base and is placed on road gutters. It functions as a filter for removing the particles from runoff [25]. The porous structure functions to retain particles present in the runoff, reducing the pollutant load [7, 26–28]. The high void content results in less strength; for this reason, permeable pavements are normally applied in areas with low volume traffic and limited heavy vehicle loading [29]. The IES uses permeable pavement technology (pervious concrete or porous asphalt) limited to gutters area with reduced vehicular traffic, therefore allowing application on highways. The filter capacity is defined by the pore media average diameter (d_m) and the particle diameter (d_{p}) that govern the particle transport within the porous media. Three main mechanisms of transport can be distinguished: surface (cake/schmutzdecke), straining filtration and physical-chemical filtration, depending on the average pore diameter and the particle diameter ratio [2, 30]. Laboratory measurement of a porous asphalt surface with 50 mm thickness submitted to various loadings, rainfall intensities and durations showed a mass-based particle removal ranging from 88% to 97% through cake and straining mechanisms [28]. Data obtained from an IES with 90 mm pervious concrete and 600 mm oxide-coated media layers showed a mass-based particle removal from 83% to 99%. The IES efficiency in particle removal was compared with an FFT, considering the six rainfall events presented in Table 1.

2.4 Description of goodness-of-fit tests

The following parameters were chosen to verify if SWMM was effective in capturing the significant components of the storm: peak flow, runoff duration, total volume of flow, runoff coefficient (c) and runoff flow. The criterion used to compare peak flow, time to peak and total volume of flow was percent error, while Nash–Sutcliffe efficiency (NSE) was used for accumulated volume. The measured build-up and wash-off and modelling were analysed using r-squared and percentual error.

3 RESULTS

3.1 Hydrological modelling

Rainfall-runoff transformation was modelled using SWMM and goodness-of-fit analysis and is presented in Tables 2 and 3. Considering cumulative volume, high-intensity events presented a compressive better fit when the watershed imperviousness was 100%, whereas in low-intensity events, a better fit was found when the imperviousness was 50% (Table 2). This could be because the vehicular traffic has more effect on rainfall-runoff process during low-intensity events.

A comparison of measured and modelled parameters considering 100% imperviousness for high-intensity events, 50% for low-intensity events and 75% for all events is presented in Table 3. Overall, high-intensity events showed a better fit than low-intensity events, where the effect of abstractions could be more accentuated. If holding a rainfall continuous simulation, and so using the same modelling parameters for all event types would result in overestimating the runoff volume for low-intensity events and underestimating for high-intensity events, reaching an overall percent error of 3%. Also, considering 75% imperviousness, peak flow is underestimated for all rainfall types with a compressive percentual error of 7%.

3.2 PM transporting modelling

TS and TSS build-up measured and modelled using eqn (1) is shown in Fig. 1, considering for TS, Accu of 7 kg/ha d and Disp of $0.27 d^{-1}$ and for TSS, Accu of 5 kg/ha d, Disp of $0.25 d^{-1}$. Goodness of fit for wash-off and build-up was also modelled using SWMM, considering the parameters already mentioned for build-up and the single-event EMC for wash-off, obtaining a overestimation for mass from 10% to 19%.

Table 4. PM mass was obtained through runoff; for this reason, to model build-up, only high-intensity events were considered based on the hypothesis that the entire accumulated surface mass was wash-off with runoff.

Rainfall event	Modelled100 ^a	Modelled75 ^b	Modelled50 ^c			
18/06/1996	0.86	0.91	-			
07/07/1996	0.72	0.97	-			
08/08/1996	0.81	0.17	-			
17/10/1996	-	-	0.97			
25/11/1996	-	-	0.88			
16/12/1996	-	-	0.86			
^a Modelled 100, 100% imperviousness. ^b Modelled 75, 75% imperviousness. ^c Modelled 50, 50% imperviousness. NSE, Nash–Sutcliffe efficiency.						

Fable 2: NSE for total vo	olume.
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Rainfall event	Parameter	Meas- ured	Modelled ^a	Percent error	Mod- elled ^b	Percent error
18/06/1996	с	0.82	0.929	13%	0.697	15%
18/06/1996	Peak flow (L/min)	244	268.2	10%	204	16%
18/06/1996	Runoff duration, min	76	57	25%	51	33%
18/06/1996	Runoff volume (L)	2779	3154	13%	2367	15%
07/07/1996	с	0.79	0.983	24%	0.737	7%
07/07/1996	Peak flow (L/min)	322	492	53%	376.2	17%
07/07/1996	Runoff duration (min)	60	67	12%	61	2%
07/07/1996	Runoff volume (L)	9643	11928	24%	8951	7%
08/08/1996	с	0.91	0.932	2%	0.7	23%
08/08/1996	Peak flow (L/min)	391	400.2	2%	311.14	20%
08/08/1996	Runoff duration (min)	52	45	13%	63	21%
08/08/1996	Runoff volume (L)	3877	3266.64	16%	2451	37%
17/10/1996	с	0.42	0.483	15%	0.725	73%
17/10/1996	Peak flow (L/min)	44.3	29.4	34%	42.6	4%
17/10/1996	Runoff duration (min)	609	614	1%	629	3%
17/10/1996	Runoff volume (L)	3693	3499.8	5%	5248.62	42%
25/11/1996	с	0.23	0.339	47%	0.509	121%
25/11/1996	Peak flow (L/min)	9	4.8	47%	6.6	27%
25/11/1996	Runoff duration (min)	146	109	25%	117	20%
25/11/1996	Runoff volume (L)	216	259.26	20%	388.92	80%
16/12/1996	с	0.26	0.352	35%	0.528	103%
16/12/1996	Peak flow (L/min)	4	3.6	10%	4.8	20%
16/12/1996	Runoff duration (min)	324	341	5%	355	10%
16/12/1996	Runoff volume (L)	269	209.01	22%	436.5	62%

Table 3: Rainfall-runoff modelling goodness of fit.

^aConsidering 100% imperviousness for high-intensity events (18/06, 7/7 and 8/8) and 50% imperviousness for low-intensity events (17/10, 25/11, 16/12). ^bConsidering 75% imperviousness.



Figure 1: Build-up measured and modelled for TS and TSS, considering the high-intensity rainfall events.



Figure 2: Measured and modelled particle mass using (a) eqn (3) for TS to the 18/06/1996 rainfall event (mass limited) and (b) eqn (4) for TS to the 16/12/1996 rainfall event (flow limited).

	TS						
D - ' 6- 11	Build-up		Wash-off				
event			Reg	gression ^a	SWMM ^b		
	R-squared	Percentual error ^a	R-squared	Percentual error ^c	Percentual error ^c		
18/06/1996	0.75	21%	0.98	5%	14%		
07/07/1996		26%	0.9	3%	23%		
08/08/1996		11%	0.96	9%	16%		
17/10/1996	-	-	0.96	1%	19%		
25/11/1996	-	-	1	3%	16%		
16/12/1996	-	-	0.96	7%	10%		

Table 4: Goodness of fit for build-up and wash-off modelling.

Dainfall	Build-up		Wash-off			
event			Reg	SWMM ^b		
	R-squared	Percentual error ^a	R-squared	Percentual error ^c	Percentual error ^c	
18/06/1996	0.62	25%	0.97	6%	14%	
07/07/1996		54%	0.96	3%	26%	
08/08/1996		15%	0.94	9%	16%	
17/10/1996	-	-	0.95	1%	7%	
25/11/1996	-	-	1	0%	16%	
16/12/1996	-	-	0.87	14%	10%	

^aRegression from measured data considering eqns (3) and (4) according to event classification.

^bSWMM using single-event EMC for quality modelling.

^cPercentual error for total event mass in mg.

Figure 2 illustrates wash-off modelling for a mass-limited event using eqn (3) and for a flow-limit event using eqn (4). Modelling underestimated total wash-off mass from 1% to 9%. Goodness-of-fit analysis for both build-up and wash-off is shown. Wash-off and build-up were also modelled using SWMM, considering the parameters already mentioned for build-up and the single-event EMC for wash-off, in which an overestimation of mass from 10% to 19% was obtained.

Wash-off and build-up were also modelled using SWMM, considering the parameters already mentioned for build-up and the single-event EMC for wash-off, which gave an over-estimation of mass from 10% to 19%.



Figure 3: First flush plots: (a) 18/06/1996 and (b) 16/12/1996 rainfall event.

3.2.1 First flush analysis

Normalized mass and flow against normalized time were used to analyse first flush, where a mass curve above flow denoted a first flush behaviour. The 18/06/1996 rainfall event presented a clear first flush typical of a mass-limited event, while for the 16/12/1996 rainfall event, a weak first flush behaviour coherent with flow-limited events was observed (Fig. 3). The same trend was observed for the other events.

3.3 Stormwater control system

Considering the six registered single events, an FFT with 1500 L capacity (5 mm rainfall depth) would remove 65% of TS mass. Two events of low intensity and long duration did not reach a total depth of 5 mm. The established depth of first flush considering for the FFT, and in both cases, the entire event is stored resulting in 100% removal, even though these events presented a weak first flush, as seen in Fig. 3, for the 16/12/1996 event.

For the 18/06/1996 event, with the highest removal, the FFT trapped the first two mass peaks retaining 75% of TS mass (Fig. 4). All the results obtained are gathered on Table 5.



Figure 4: Runoff and first flush tank (FFT) incoming flow and mass (TS) and cumulative volume and cumulative mass. FFT reaches full storage for 1500 L. Rainfall events: (a) 18/06/1996 (b) 08/08/1996.

Table 5: FFT particle removal (TS) from measured data, regression model and SWMM.

	Mea	sured ^a	Regression ^b		SWMM ^c	
Rainfall event	Time full capacity	TS removal (%)	TS removal (%)	Percent error ^d (%)	TS removal (%)	Percent error ^d (%)
18/06/1996	41	74%	84%	8%	48%	27%
07/07/1996	12	58%	42%	29%	13%	73%
08/08/1996	18	42%	67%	46%	46%	7%
17/10/1996	147	79%	86%	9%	50%	49%
25/11/1996	-	100%	100%	3%	100%	16%
16/12/1996	-	100%	100%	7%	100%	10%
TOTAL RE	MOVAL	65%	74%	9%	44%	35%

^aTS removal estimated from measured data.

^bRegression from measured data considering eqns (3) and (4) according to event classification.

^cSWMM using single-event EMC for quality modelling.

^dPercent error between measured and modelled data.

The lowest removal was observed for the 08/08/1996 event. Although the event presented a clear first flush, the FFT storage capacity was reached before the peak mass (Fig. 4). An FFT with 2000 L capacity, corresponding to a 7 mm depth, would remove 8% of TS mass.

SWMM modelling underestimated the TS mass removal since the EMC index did not consider a first flush. Considering a TS/TSS ratio from 1.2 to 2.5, a TSS mass removal from 80% to 95% by an IES could represents a 50%–60% removal of TS. Regression modelling overestimates removal, while SWMM underestimates, not taking into account the first flush effect.

4 CONCLUSION

Stormwater from urban paved areas must undergo treatment to comply with discharge regulations and avoid stream and rivers pollution. Source treatment should be encouraged to limit flow that reaches treatment plants. This study compared the particle removal efficiency for two stormwater control systems, an FFT, which relies on the first flush phenomenon, and an IES with porous surface that promotes particle removal through filtration mechanisms. Considering the characteristics and limitations of both systems, the choice between them relies on analysing the local rainfall regime and dominant event behaviour (mass limit or flow limit) to evaluate the feasibility of an FFT or a system based on filtration for the whole event, such as the IES. Modelling particle transport can be an important tool to choose between systems. Using exponential or linear models can represent better mass- or flow-limit events, while EMC does not represent a first flush behaviour, although it is often used. Further steps in this study would consider applying computational fluid dynamics (CFD) to model IES particle removal and to use continuous simulation to observe particle removal in the long term.

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