

Article

Assessing the Performance of Different Time of Concentration Equations in Urban Ungauged Watersheds: Case Study of Cartagena de Indias, Colombia

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Abstract: In ungauged watersheds, the estimation of the time of concentration (T_c) is always a challenging task due to the intrinsic uncertainty involved when making assumptions. Given that T_c is one of the main inputs in a hydrological analysis for the design of hydraulic structures for stormwater management, ten equations (including one proposed in several local studies) and two T_c methodologies (overland flow time plus channel flow time) were used to compute the T_c in fifteen urban ungauged watersheds, located in Cartagena de Indias (Colombia), with different area sizes and slopes to statistically assess their performance against the value obtained via the Natural Resources Conservation Service (NRCS) velocity method (assumed to be the true value). According to the Nash–Sutcliffe efficiency index, none of the equations proved to be reliable in all watersheds as only four equations predicted the T_c value in 53% of the cases. In addition, based on the percent bias, all equations tended to significantly over- or underestimate the T_c , which affects the quantification of the runoff volume necessary for, among others, the implementation of best management practices for watershed management (e.g., conventional and/or sustainable drainage system design), flood-prone area delineation and flood risk analyses, urban planning, and stream restoration.

Keywords: time of concentration; non-stationarity; ungauged urban watershed; stormwater management; Cartagena de Indias

1. Introduction

Hydrological analysis in ungauged watersheds might be challenging as several assumptions have to be made that could introduce uncertainties into the estimation of the design runoff via rainfall-runoff models, in which rainfall is one of the main inputs. This is particularly true in ungauged watersheds where, typically, rainfall data are only available from nearby gauges. In the absence of data from pluviographs that help with the understanding of rainfall duration and distribution over time, the T_c becomes a variable of paramount relevance in determining the design rainfall. The time of concentration has been defined differently [1–10] and multiple empirical equations have been derived for its estimation that can be found throughout the available literature on hydrology,

in different types of stormwater management and design of drainage structure manuals/guidelines [3, 11–14], embedded in software programs for hydrologic-hydraulic modeling such as Win TR-20 [15], Win TR-55 [16], HydroCAD® [17], Autodesk® [18], Storm Water Management Model (SWMM) [19], Hydrologic Engineering Center-Hydrologic Modeling System HEC-HMS [20], SewerCAD® [21], and StormCAD® [22], in freely available T_c online calculators such as Vlab [23], or as an input in rainfall-runoff models such as the rational method. These equations have considered different variables that range from rainfall and surface direct runoff to watershed morphological parameters (e.g., area, slope, length of flow path, and outlet elevation), land use/land cover (LULC) conditions, and channel characteristics.

The development of an equation based on local or regional data is always preferable. Manoj and Fang [24], Yoo et al. [25], De Almeida et al. [26], Sandoval-Erazo et al. [27], Michailidi et al. [9], and Perdikaris et al. [10] are examples of studies conducted to derive equations based on local available information. Unfortunately, in ungauged areas, there is no other option but to use existing empirical T_c equations. On this basis, the existence of numerous equations might become a hindrance when it comes to deciding which is best for a given area, especially when the person using a particular equation is not familiar with how it was derived and/or its advantages and intrinsic limitations. Various studies [1,2,5–8,28–36] performed in different locations around the world have resulted in large differences among the T_c values estimated by means of multiple equations, which contributes to (a) highlighting the need for assessing the performance of existing T_c equations prior to their use, (b) realizing that the equations cannot be randomly used, and (c) understanding and quantifying the impact on the subsequent computation of the design runoff.

Given the large amount of ungauged areas around the world, water-related professionals are accustomed to the use of empirical T_c equations for hydrological analyses. For instance, in Ireland [37] and Colombia [38–40], some estimate T_c as the average of T_c values from selected equations, while others prefer to eliminate the outliers before calculating the average. In the case of Colombia, this seems to be the reason that Invías [41] (pp. 22–28) and MinVivienda [42,43] recommend, respectively, fifteen (Kirpich, Témez, Williams, Johnstone–Cross, Giandotti, California culvert practice, Ventura–Heras, V.T. Chow, Kerby–Hathaway, Izzard, U.S. Federal Aviation Administration (FAA), kinematic wave, NRCS lag equation, George Rivero, and NRCS velocity method for shallow concentrated flow) and three empirical equations (Kerby–Hathaway, FAA, and NRCS velocity method for shallow concentrated flow). Whereas this practice might give the user a range of T_c values to choose from, it also raises doubts as to whether the selection of such an array of equations was indeed the result of an assessment of their performance. Furthermore, based on the authors' experience and the scarce literature on T_c equation assessment in Colombia, the reasons behind such practice when computing the T_c are unclear. The bottom line is that averaging, per se, does not suffice when it comes to assessing the performance of such equations. Indeed, few studies [44,45] have either addressed this matter (though both studies used averaging as well) or developed local/regional T_c equations based on observed data [46,47].

The inherent degree of uncertainty involved in the computation of T_c in ungauged areas should be minimized by performing an assessment of the equation(s) to be utilized [48] (pp. 9.16). Therefore, this study intends to help with this task by assessing the performance of ten empirical equations (and two T_c methodologies that combine two different equations) in fifteen small urban ungauged watersheds located in Cartagena de Indias (Colombia) by comparing the T_c values with the value obtained via NRCS velocity method [3,49], which was assumed to be the “true” value owing to (a) the absence of hyetographs and hydrographs from which T_c could have been derived, (b) the conceptual robustness behind the method [6,10,11] that permits it, to a certain extent, to make up for the lack of data, and (c) in urban areas, it is recommendable to estimate T_c as the result of length over velocity [48]. The assessment consisted of measuring accuracy, bias, the model's prediction power, and true simulated values' linear correlation through, respectively, pre-established performance categories [50–54] for the statistical indicators of ratio of root mean square error and standard deviation (RSR), percent bias ($PBIAS$), Nash–Sutcliffe efficiency (NSE), and coefficient of determination (R^2). The assessment also

took into account the performance of the T_c equations when (a) applying the empirical formulas to the average slope and total length (original approach), (b) applying the empirical formulas to each of the identified flow types and then adding them up, and (c) utilizing both the 2-year, 24 h rainfall (P_2) under stationary (SC) and non-stationary (NSC) conditions for the NRCS velocity method sheet flow formula owing to the findings of González-Álvarez et al. [55], who concluded that a non-stationary frequency analysis better fitted the 24 h rainfall time series at Aeropuerto (Apto.) Rafael Núñez rain gauge.

2. Study Area and Data

Cartagena de Indias, a city located within the Caribbean Colombian region, has a total area of approximately 623 km², with an average total annual rainfall ranging between 1000 and 1500 mm [56], a rainy season during the months of April–June and August–November [57], and reported values of daily maximum rainfall of up to 222 mm (reported in 2 October 2016 at the Rafael Núñez Airport rain gauge). Despite the fact that 87.8% of its territory is rural, more than 90% of its population (around 925,860 people) lives in the urban area [58]. This exacerbates the flooding problems that the city experiences every year during the rainy season, chiefly caused by, among others, increasing rainfall and obsolete and/or deficiently designed hydraulic structures for stormwater management [55].

The selected urban watersheds' areas range from 29.2 to 72.8 ha and discharge into either Las Quintas Swamp or Bazaruto and Juan Angola Bayous (Figure 1 and Table 1), which are the water bodies that have undergone the largest anthropogenic pressure among the six water bodies that compose the city's internal estuarine bayous and lagoons system [59].

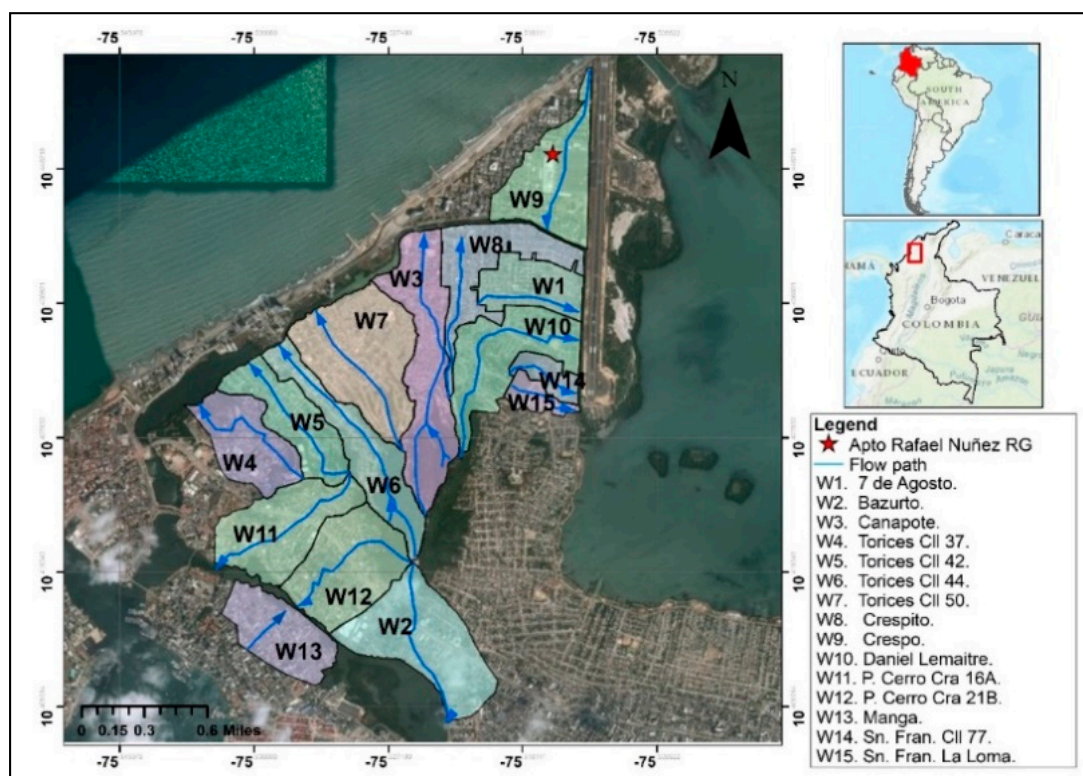


Figure 1. Study area selected watersheds. Source: adapted from Establecimiento Público Ambiental de Cartagena (EPA-Cartagena) and Universidad de Cartagena (UDC) [59].

Table 1. Selected urban watersheds.

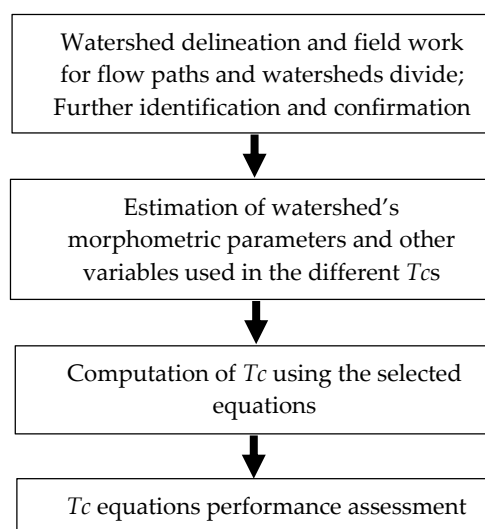
No.	Watershed Name	Area (ha)	Receiving Water Body
W1	7 de Agosto	29.20	Juan Angola Bayou (in the vicinity of Canal Paralelo)
W2	Bazurto	66.05	Las Quintas Swamp
W3	Canapote	67.87	Juan Angola Bayou (in the vicinity of Crespo)
W4	Torices, Calle 37	39.52	Juan Angola Bayou (in the vicinity of El Cabrero Lagoon)
W5	Torices, Calle 42	38.12	Juan Angola Bayou (in the vicinity of El Cabrero Lagoon)
W6	Torices, Calle 44	47.76	Juan Angola Bayou (in the vicinity of El Cabrero Lagoon)
W7	Torices, Calle 50	72.80	Juan Angola Bayou (in the vicinity of Marbella)
W8	Crespito	44.00	Juan Angola Bayou (in the vicinity of Crespo)
W9	Crespo	50.80	Juan Angola Bayou (in the vicinity of Crespo)
W10	Daniel Lemaitre	52.56	Juan Angola Bayou (in the vicinity of Canal Paralelo)
W11	Pie del Cerro, Cra. 16A	52.85	Bazurto Bayou
W12	Pie del Cerro, Cra. 21B	56.58	Bazurto Bayou
W13	Manga	38.47	Bazurto Bayou
W14	San Francisco, Calle 77 (La Amistad)	11.18	Juan Angola Bayou (in the vicinity of Canal Paralelo)
W15	San Francisco, sector La Loma	9.29	Juan Angola Bayou (in the vicinity of Canal Paralelo)

Source: EPA-Cartagena and UDC [59].

Some of the data used in this study include the P_2 value of the Apto. Rafael Núñez rain gauge (study area's nearest rain gauge), estimated by González-Álvarez et al. [55], who concluded that a non-stationary frequency analysis better fitted the 24 h rainfall time series. Thus, both stationary and non-stationary P_2 values (89 and 128 mm, respectively) were used as some of the inputs for the NRCS velocity method sheet flow formula to test whether it might bring any improvement in the performance of the T_c equations. The derivation of the remaining variables (length, slope, runoff coefficient, curve number, imperviousness and conveyance factors, and Manning coefficient) needed for the different T_c equations used is explained in the following sections as they (the variables) are all dependent on the watershed's characteristics.

3. Methodology

The proposed methodology (Figure 2) consisted of four basic tasks: (a) the watershed delineation, (b) selection of T_c equations and estimation of relevant watershed morphometric parameters and T_c equations related variables, (c) T_c computation, and (d) T_c equation performance assessment via statistical analysis. Each of these stages are detailed in the following subsections.

**Figure 2.** Methodology's flowchart.

3.1. Watershed Delineation, Morphometric Parameters, and Flow Types

The delineation of the fifteen watersheds, initial identification of the flow paths, and flow length and slope computation were conducted via ArcGIS (Version 10.6.1, ESRI Inc., Redlands, CA, USA [60]) using the 30-m National Aeronautics and Space Administration (NASA) Digital Elevation Model (DEM) [61]. Field work was also performed to (a) corroborate watersheds divide, (b) identify multiple flow paths and types of flows (sheet, shallow concentrated, and channel) within each watershed, and (c) obtain channel dimensions in watersheds where channel flow was identified.

Table 2 summarizes the different flow paths and flow types identified as well as their total length, average slope, and shallow concentrated flow as a percentage of the total flow. It might be observed that this type of flow corresponds to 46% to 100% of the total watershed identified flow types. Most of the empirical T_c equations have the slope as one of their main inputs, which greatly impacts the T_c estimation [10,62]. As a result, in this study, the slope estimation for each identified flow type and the entire watershed's flow path was carried out by carefully observing the elevation longitudinal profiles of each of the identified flow paths. This avoids incurring any possible error by omitting relevant elevation differences along them (the flow paths).

It is also noteworthy to point out that, in Colombia, contrary to selecting the longest travel time of all possible flow paths within a watershed, it is a common practice among hydrologists to select the shorter time of concentration (with a minimum possible value of 15 min), arguing that it helps to be on the safe side because a shorter T_c value will lead to a larger rainfall (or rainfall intensity) [41] (pp. 2–8). However, this omits the fact that peak discharge occurs when all parts of a watershed are contributing to the runoff, which is assumed to occur at the time of concentration [9,11,63]. In this study, the longest travel time is used as the watershed's time of concentration. Table 2 presents the paths with the longest travel time of all identified flow paths within each of the fifteen urban watersheds (Table S1). Figure 3 shows the flow paths identified in some of the selected watersheds of this study.

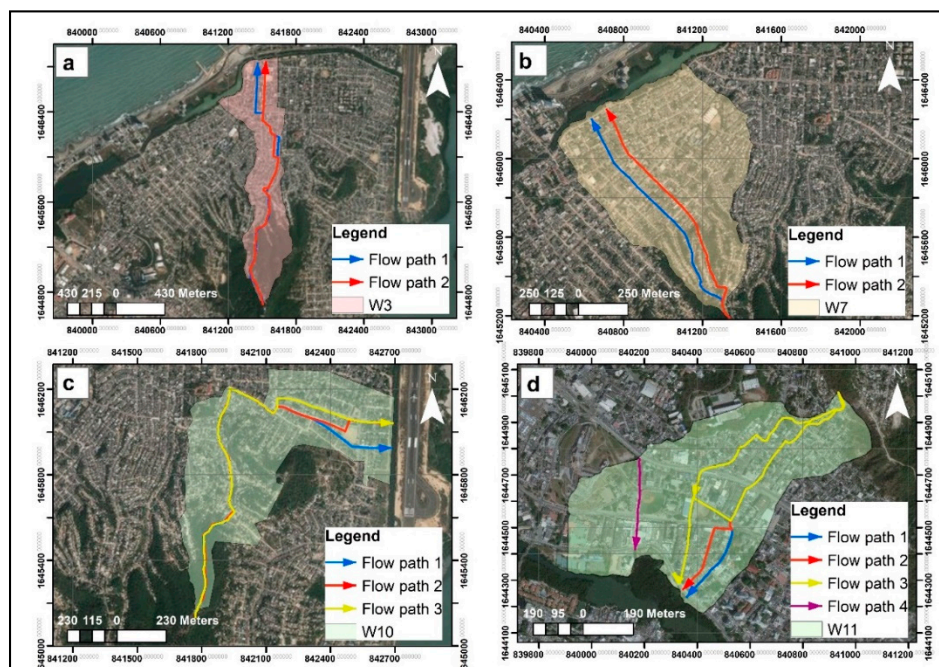


Figure 3. Watersheds' identified flow paths. (a) Canapote (W3); (b) Torices Calle 50 (W7); (c) Daniel Lemaitre (W10); (d) Pie del Cerro Cra 16A (W11).

Table 2. Identified flow types within selected watersheds.

Watershed	Flow Path	Flow Type			Total Length (m)	Watershed Avg. Slope (%)	Shallow Conc. Flow as a % of the Total Flow
		Sheet	Shallow Concentrated	Channel			
7 de Agosto	1	(100 m; 5.2%)	(748 m; 0.04%)	---	848	0.2	92%
Bazurto	7	(100 m; 39.4%)	(1264 m; 15.1%)	(170 m; 1.4%)	1534	18.8	82%
Canapote	2	(100 m; 3.3%)	(2361 m; 6.5%)	(10 m; 11.0%)	2471	6.3	96%
Torices, Calle 37	1	(60 m; 3.4%)	(782 m; 6.0%)	(398 m; 0.6%)	1240	4.3	63%
Torices, Calle 42	2	(100 m; 12.3%)	(1289 m; 6.9%)	---	1389	8.4	93%
Torices, Calle 44	1	(100 m; 10.5%)	(1836 m; 10.0%)	---	1935	9.9	95%
Torices, Calle 50	2	(100 m; 3.9%)	(1213 m; 7.3%)	(14 m; 10.1%)	1327	6.4	91%
Crespito	2	---	(664 m; 2.4%)	---	664	2.4	100%
Crespo	5	---	(498 m; 0.01%)	---	498	0.01	100%
Daniel Lemaitre	2	(100 m; 4.4%)	(1981 m; 6.1%)	---	2081	4.8	95%
Pie del Cerro, Cra 16A	3	(100 m; 8.1%)	(713 m; 12.7%)	(420 m; 0.1%)	1233	9.2	58%
Pie del Cerro, Cra 21B	4	(100 m; 14.6%)	(1291 m; 14.8%)	(27 m; 0.7%)	1418	15.2	91%
Manga	1	---	(978 m; 0.05%)	---	978	0.05	100%
San Francisco, Calle 77 (La Amistad)	2	(100 m; 7.2%)	(647 m; 7.3%)	---	747	8.3	87%
San Francisco, sector La Loma	2	(100 m; 9.1%)	(579 m; 9.5%)	---	679	9.4	85%

Note: Values in parentheses are the length and slope of the corresponding identified flow type. Blue cells indicate the selected flow path as they reported the larger time of concentration.

3.2. Selected T_c Equations

A total of ten empirical equations used for different types of flow and two additional methodologies that involve the combination of some of these ten equations were selected for the computation of the T_c : Miller [64,65], U.S. Federal Aviation Agency (FAA) [66], Espey-Winslow (E-W) [67], Kerby-Hathaway (with and without slope adjustment) [65,68], NRCS lag equation [11], Kirpich-Tennessee (with and without slope adjustment) [65,69], Bransby-Williams (B-W) [70], Johnstone-Cross (J-C) [71], Sheridan [72], NRCS velocity method for shallow concentrated flow with a local/regional coefficient for Cartagena proposed by Cartagena's Stormwater Management Master Plan (PMDPC in Spanish) [73], the Kerby-Kirpich (K-K) methodology with and without slope adjustment [14,65,74], and the T_c methodology proposed by MinVivienda [42]. A brief description of all equations and methodologies is provided in Tables S7–S10.

For performance assessment purposes, T_c values were computed by means of the NRCS velocity method [3,11], which were assumed to be the “true” (or reference) values owing to (a) the absence of observed rainfall and runoff data (ungauged watersheds) from which T_c could have been derived and (b) the methodology's robustness. The NRCS velocity method defines three types of flow, namely sheet, shallow concentrated, and channel. The time of concentration is the sum of the travel times of each identified flow type (Equation (1)). The travel time for sheet flow is estimated via Equation (2), while for shallow concentrated and channel flows, it is necessary to first calculate the velocities. Sheet flow velocity is estimated via either plot (or their corresponding equations [3,11]) if the flow path slope and soil cover type are known. Channel flow velocity is estimated through Equation (3). Finally, travel times for shallow concentrated and channel flow are estimated via Equation (4).

$$T_c = T_{sheet} + T_{Shallow Conc} + T_{Channel} \quad (1)$$

$$T_{Sheet} = \frac{0.007(n_0L)^{0.8}}{(P_2)^{0.5}(S)^{0.4}} \quad (2)$$

where T_{sheet} is sheet flow travel time (h), n_0 is the roughness coefficient, L is the flow path length (m), P_2 is the 2-year, 24 h rainfall (mm), and S is the slope (m/m).

$$V_{Channel} = \frac{R^{0.67}S^{0.5}}{n} \quad (3)$$

where $V_{Channel}$ is the channel water velocity (m/s), R is the hydraulic radius (m), S is the channel slope (m/m), and n is the Manning's coefficient.

$$T_t = \frac{L}{3600V} \quad (4)$$

where T_t (h), L (m), and V (m/s) are the travel time, flow path length, and velocity for either shallow concentrated ($T_{Shallow Conc}$) or channel ($T_{Channel}$) flows.

Additionally, in order to test how using a given equation for a type of flow for which it was not intended could introduce errors, each of the ten equations was applied to (a) each of the identified flow types within a watershed (sheet, shallow concentrated, and channel flow), and the total time of concentration was estimated as the sum of the times of concentration computed individually for each of the flow types (named each flow, E.F.), and (b) to the flow path as a whole, irrespective of the identified flow types (named total length, T.L.). Both results were compared to that of the NRCS velocity method (using P_2 under SC and NSC). For instance, if the Kirpich equation was to be used in a watershed where shallow concentrated and channel flows have been identified, then the time of concentration was estimated both ways by applying the equation to (a) each of the identified flows, and adding both results, and (b) the total length, as the original methodology that is typically used (and derived for). For the two additional T_c methodologies used in this study (Kerby-Kirpich and MinVivienda),

the sheet and shallow concentrated flows were considered as one. Additionally, these methodologies were not used for the total length scenario given the nature of the methodologies that establishes the separation of two flows to be applied.

3.3. Determination of T_c Equation Variables

Variables related to LULC, such as composite runoff coefficient (C) [75], composite curve number (CN) [3,11], and percentage of impervious area (ip) [67], were estimated by means of satellite imagery analysis (LandSat imagery processed via ENVI[®] (Version 5.1, Harris Geospatial Solutions Inc, Boulder, CO, USA [76])) (Table S2). The Espey-Winslow channelization factor (φ) and roughness (n_0) and Manning (n) coefficients were determined based on the channel soil cover observed during field visits. (Table S2). Descriptions of all variables used in the different selected T_c equations are given in Tables S7–S10.

3.4. T_c Equations Performance Assessment

The equations' performance was assessed by means of the coefficient of determination (R^2) (Equation (5)), the ratio of root mean square error and standard deviation (RSR) (Equation (6)), percent bias ($PBIAS$) (Equation (7)), and Nash–Sutcliffe efficiency (NSE) (Equation (8)) [50–52,77–79]. In Equations (5) through (8), Y_i^{obs} , Y_i^{sim} , $\overline{Y^{obs}}$, and $\overline{Y^{sim}}$ are, respectively, the values of the observed (true value obtained from NRCS velocity method), simulated (from each of the ten empirical T_c equations and two additional T_c methodologies), average observed, and average simulated values.

$$R^2 = \left[\frac{\sum_{i=1}^n (Y_i^{obs} - \overline{Y^{obs}})(Y_i^{sim} - \overline{Y^{sim}})}{\sqrt{\sum_{i=1}^n (Y_i^{obs} - \overline{Y^{obs}})^2} \sqrt{\sum_{i=1}^n (Y_i^{sim} - \overline{Y^{sim}})^2}} \right]^2 \quad (5)$$

$$RSR = \left[\frac{\sqrt{\sum_{i=1}^n (Y_i^{obs} - Y_i^{sim})^2}}{\sqrt{\sum_{i=1}^n (Y_i^{obs} - \overline{Y^{obs}})^2}} \right] \quad (6)$$


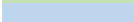
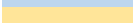


$$PBIAS = \left[\frac{\sum_{i=1}^n (Y_i^{obs} - Y_i^{sim})}{\sum_{i=1}^n (Y_i^{obs})} \right] * 100. \quad (7)$$

$$NSE = 1 - \left[\frac{\sum_{i=1}^n (Y_i^{obs} - Y_i^{sim})^2}{\sum_{i=1}^n (Y_i^{obs} - \overline{Y^{obs}})^2} \right] \quad (8)$$

The coefficient of determination indicates the association (and goodness-of-fit) between the true and the simulated values through a linear correlation and has a range of values from zero (no correlation) to one (optimal value or perfect fit). RSR is a standardized form of the root mean square ($RMSE$) that measures the error of a model, with an optimal value of zero (no error). $PBIAS$, with an optimal value of zero, shows the model's tendency to under- or overestimate when compared with the true value, where a negative $PBIAS$ indicates that the simulated value is greater than the observed one (overestimation), while positive denotes underestimation. NSE permits us to assess the model's prediction power; it ranges from $-\infty$ to one (optimal value). If negative values of NSE are obtained, then the true values' average is a better predictor than the simulated values. When NSE equals zero (or close to zero), either the average of true values or the simulated values could be used. Details of the theory and limitations/advantages of these statistical indices for model performance assessment can be widely found in the literature.

In order to better assess the equations' performance, various categories were established for each of the four statistical indices used based on their corresponding range of values, as per Moriasi et al. [51,52] (Table 3). The color bar helps with the visualization and identification of such categories.

Table 3. Statistical variables performance categories.

Performance Category	Color Bar	Statistical Variables Range of Values			
		R^2	RSR	$PBIAS$ (%)	NSE
Very good		0.85–1.00	$0.0 \leq RSR \leq 0.5$	$PBIAS < \pm 10$	0.75–1.00
Good		0.70–0.84	$0.5 < RSR \leq 0.6$	$\pm 10 \leq PBIAS < \pm 15$	0.65–0.74
Satisfactory		0.60–0.69	$0.6 < RSR \leq 0.7$	$\pm 15 \leq PBIAS < \pm 25$	0.50–0.64
Acceptable		0.40–0.59	Not applicable	Not applicable	0.40–0.49
Unsatisfactory		≤ 0.39	$RSR > 0.7$	$PBIAS \geq \pm 25$	≤ 0.39

RSR = ratio of root mean square error and standard deviation; $PBIAS$ = percent bias; NSE = Nash–Sutcliffe efficiency.

4. Results and Discussion

Tables S11–S14 show the watersheds' estimated T_c values for all equations and methodologies used for each flow and total length with P_2 values under stationary and non-stationary conditions.

Values of R^2 are presented in Table S5. The range of R^2 values under stationary conditions for each flow and total length ranged, respectively, from 0.03 (Kerby-Kirpich with slope adjusted) to 0.8 (PMDPC). For the total length under stationary conditions, the R^2 range of values was between 0.02 (Miller) and 0.82 (Kirpich). Under non-stationary conditions, the R^2 values oscillated between 0.002 (Miller) and 0.84 (PMDPC) for each flow and between 0.005 (Miller) and 0.86 (Kirpich) for total length.

The assessment of the T_c equations through the R^2 revealed, as expected, that empirical formulas perform best when applied to the total length. The coefficient of determination values increased in all cases, except for the Sheridan equation, which had a decrease of 0.01 for stationary and non-stationary conditions when the equation was applied to each flow and total length. The equations' performance improvement was also noted from the reduction in the standard deviation values in almost all equations (Tables S3 and S4). The Kirpich slope-adjusted equation had the largest R^2 increases in both stationary and non-stationary conditions (+0.18) when each flow and total length were compared (Table S5). However, the slope adjustment did not bring major improvements when compared to other equations such as FAA, unadjusted Kirpich, Johnstone-Cross, NRCS lag equation, PMDPC, or the MinVivienda methodology (with FAA and NRCS lag equation). Despite the fact that shallow concentrated flow was mostly present in all fifteen watersheds (shallow concentrated flow as a percentage of the total flow had values ranging between 58% and 100%, Table 2), channel flow equations (except for the Sheridan equation), in general, had higher R^2 values, the Kirpich equation being the one with the highest R^2 values (0.82 and 0.86 for total length under stationary and non-stationary conditions, respectively). The use of P_2 under non-stationary conditions in the NRCS velocity method sheet flow equation made R^2 values increase, which might suggest not only that NSC are more representative of Cartagena's rainfall behavior [55] but also that the NRCS velocity method sheet flow equation might work best whenever NSC are recommended after a frequency analysis has been conducted.

Regarding the assessment with respect to NSE , the equations of Kerby, NRCS lag equation, Johnstone-Cross, and PMDPC, as well as the Kerby-Kirpich slope-adjusted and MinVivienda with Kerby equation methods, showed eight watersheds (out of 15) with values within the 0.75 to 1.0 range (or very good category) when applied to total length flow and P_2 under stationary conditions (Table S15). FAA, Kirpich (adjusted and unadjusted), Kirpich-Kerby (unadjusted), MinVivienda with FAA, and MinVivienda with NRCS shallow concentrated flow followed, with seven watersheds within the very good category. The Espey-Winslow, Sheridan, and Bransby-Williams equations performed poorly, with, respectively, zero, two, and five watersheds exhibiting NSE values within the very good category. Based on these results, Espey-Winslow is not recommended. On the other hand, the Sheridan equation, despite its poor performance, demonstrated that indeed it works best when used on watersheds with low slope. The two watersheds (7 de Agosto and Manga) where the equation

had high *NSE* values had slopes of 0.13% and 0.05%. It is important to point out that the Sheridan equation did not perform well in the Crespo watershed, despite having a slope of 0.01%. A look at other watersheds' physical parameters permitted us to notice that the Crespo watershed was the least urbanized of the three (Table S2), which could partially explain its performance, with negative *NSE*, high *RSR*, and negative *PBIAS* (this denotes overestimation) (Tables S15 and S16). Overall, the poor performance of Espey–Winslow and Sheridan were in line with the findings of Gericke and Smithers [8,29].

The slope adjustment for both Kerby and Kirpich equations and the Kerby-Kirpich methodology was performed in watersheds Crespo, Pie del Cerro Cra 16A, and Manga (the only ones that complied with the adjustment conditions), aiming towards better results. Apropos the Kerby equation, however, the adjustment did not bring any improvement judging by the *NSE* values obtained under stationary conditions (Table S15) in either case (each flow and total length). As for the Kirpich equation, the adjustment made the *NSE* values decline, except for Pie del Cerro Cra 16A, where *NSE* increased from 0.62 to 0.72 when the equation was applied to each flow (Table S15). For the Kerby-Kirpich methodology, only the Pie del Cerro Cra 16A watershed presented improvement, with *NSE* values increasing from 0.68 (unadjusted) to 0.76 (slope adjusted). Based on the results and considering the watersheds' types of flow (Crespo and Manga with shallow concentrated flow and Pie del Cerro Cra 16A with all three flows), the slope adjustment seemed to work best for the channel flow (Kirpich equation).

When P_2 under NSC was used (Table S16), it was observed that the Kerby equation's performance slightly worsened when compared to that obtained with P_2 under SC. It moved from eight watersheds with *NSE* values within the very good category to seven. Similarly, the MinVivienda with Kerby equation and Kerby-Kirpich slope-adjusted method experienced the same setback: the first decreased from eight watersheds to seven and the latter from eight to six. On the other hand, the NRCS lag equation increased from eight to nine watersheds, PMDPC remained constant with eight watersheds, and MinVivienda with the NRCS equation (for shallow concentrated flow) moved from seven to nine watersheds (the *NSE* values of watersheds Bazarro and San Francisco La Loma increased from 0.7 to 1.0 and from 0.48 to 0.75, respectively).

All T_c equations applied to the total length performed better than when applied to the each length approach, corroborating the notion that their use should be limited to what they were designed for. As for the use of P_2 under SC or NSC, *NSE* results indicated that, under SC, more T_c equations ended up having up to eight watersheds (8 out of 15, or 53.3%) within the very good category. On the other hand, under NSC, more watersheds (9 out of 15, or 60%) lay within the same category despite having only two T_c equations with that result (NRCS lag equation and MinVivienda with NRCS velocity method equation for shallow concentrated flow). This could be also an indicator of the local rainfall behavior's non-stationarity, as previously mentioned. Nonetheless, further investigation is recommended.

In terms of both the error (*RSR*) and bias (*PBIAS*) performance (Tables S15 and S16), T_c equations with *NSE* values in the range of 0.75 to 1.0 had also low values of *RSR* (within the very good category, 0.0 to 0.5) in either case of P_2 under SC or NSC. T_c equations applied to total flow performed better than each flow, which was also previously noticed. Notwithstanding, most of the equations presented bias, with values greater than or equal to $\pm 25\%$ (unsatisfactory). With either P_2 under SC or NSC, there was a tendency for all equations to overestimate (negative *PBIAS*) with the each flow approach, whereas the opposite was true for the total flow approach. Kirpich was the equation with more watersheds (four) with *PBIAS* the under very good category ($PBIAS < \pm 10\%$) under stationary conditions, followed by the FAA equation and MinVivienda methodology (with Kerby and FAA equations) with three, and the Kerby, NRCS lag equation, J-C, and Sheridan, with two watersheds. Under non-stationary conditions, the number of watersheds under the very good category dropped by one unit in some of the equations: Kirpich (from 4 to 3), Kerby and J-C (from 2 to 1), J-C (from 1 to 0), and MinVivienda with Kerby equation (from 3 to 2). However, this was a slight performance diminishment as all those watersheds fell

into the following category ($\pm 10\% \leq PBIAS < \pm 15\%$, or “good” category), except for the MinVivienda with Kerby equation.

The concernment of the obtained *PBIAS* results lies chiefly in the fact that, in ungauged watersheds, where the computation of the design flow for a given return period is mostly performed via rainfall-runoff models, the *Tc* plays a key role since the duration of design rainfall depends on it. This is particularly true, for instance, when using the rational method for peak flow estimation in small urban watersheds (area < 80 ha) [14,42,43]. In this rainfall-runoff model, rainfall intensity is obtained from Intensity-Duration-Frequency (IDF) curves, whose main input is the rainfall duration, which is assumed to be equal to the *Tc*. As a result, any tendency to over- or underestimate the *Tc* will affect the design runoff and the inherent implications of this on the design of hydraulic structures, which might lead to life, infrastructure, and economic losses [80,81]. As a result, efforts to reduce possible errors when computing *Tc* should be made by minimizing any source of uncertainty, such as the estimation of the variables associated with a particular equation (rainfall, slope, area, land cover, flow path length, etc.) [5,10,82]. The more input variables a model has, there more chances there are to introduce errors, especially in hydrology, where the watershed’s morphometric parameter estimation depends on map scales and imagery interpretation. For instance, a closer look at watersheds Pie del Cerro Cra. 16A and Pie del Cerro Cra. 21B, which have similar area sizes (52.85 ha and 56.58 ha) and flow path lengths (1233 m and 1480 m) (Table 2), revealed the following:

- (a) Equations with two and three variables outperformed Espey–Winslow despite having two additional variables related to land cover (channelization factor and impervious area percentage), which did not bring any advantage, especially in the case of the channelization factor (Table 4 and Tables S7–S10). In fact, Espey–Winslow was the equation with the poorest performance of all (Tables S15 and S16).
- (b) The slope and length of each of the different flow types seems to affect the equations’ performance. Their average slope is different, with values of 9.2% and 15.2%, respectively. Additionally, Pie del Cerro Cra 21B has 91% (versus 58% of Pie del Cerro Cra 16A) of its total flow path corresponding to shallow concentrated flow, with a slope of 14.8% (Table 2). In this sense, watershed area and total urbanized area appear to also play a role in the equations’ performance given that the NRCS lag equation and Bransby–Williams performed well in the Bazurto watershed, despite having an average slope of 18.8% (larger than Pie del Cerro Cra 21B), 82% of its total flow path as shallow concentrated flow with a slope of 15.1% (similar to Pie del Cerro Cra 21B), and an area of 66.05 ha (Table 2). These two equations did not perform well in Pie del Cerro Cra 16A and Cra 21B even though the watersheds have similar area sizes and urbanized areas.

Table 4. Variables in each *Tc* Equation.

<i>Tc</i> Equation	No. of Variables	Flow Type	Variable Description
Miller	3	Sheet	Roughness coefficient, flow path length and slope
Federal Aviation Administration (FAA)	3	Sheet	Runoff coefficient, flow path length and slope
Espey–Winslow	4	Shallow Concentrated	Flow path length and slope, channelization factor, and impervious area percentage
Kerby	3	Shallow Concentrated	Roughness coefficient, flow path length and slope
Natural Resources Conservation Service (NRCS) Lag Equation	3	Shallow Concentrated	Curve number, flow path length and slope
Kirpich	2	Channel	Flow path length and slope
Bransby–Williams	3	Channel	Flow path length and slope, and watershed area
Johnstone–Cross	2	Channel	Flow path length and slope
Sheridan	1	Channel	Flow path length
Plan Maestro de Drenajes Pluviales de Cartagena (PMDPC)	2	Channel	Flow path length and slope

The authors also estimated the average *Tc* from equations Miller, FAA, Kerby, E-W, NRCS lag equation, Kirpich, Bransby–Williams, Johnstone–Cross, Sheridan, and PMDPC so as to mimic the procedure practiced by some hydrologists (none of the *Tc* methodologies assessed herein were included as they are not typically used to estimate *Tc*). Outliers were first identified (and excluded) by means

of box plots. The resulting T_c average values were compared with those from the NRCS velocity method for each of the watersheds. The average T_c approach also showed poor agreement, with a tendency to underestimate in nine and ten watersheds for the each flow and total length approaches, respectively, under SC or NSC (gray cells in Table S6). The underestimation was less evident under non-stationary conditions.

All T_c equations (and T_c estimation methodologies) evaluated in this study showed setbacks, including those recommended by the Colombian drainage manuals MinVivienda [42] and Invías [41]. Inasmuch as these two manuals are used by many as almost indisputable references for hydrological analyses in the case of Colombia, the findings herein reaffirm the necessity of assessing the suitability of a given equation prior to its application, especially in ungauged areas. In the context of Cartagena de Indias, this becomes particularly relevant due to the use of an ad hoc T_c equation (PMDPC) to design eight concrete open channels for the first phase of the city's Stormwater Management Master Plan, one of which (7 m by 1.0 m and 400 m long channel, which was built less than two years ago) is overflowed every year during the rainy season (Figure 4). The PMDPC's performance was no better than that of the other equations, despite featuring a so-called regional constant (Table S9, $K = 4.47$) not validated by any study. Furthermore, the PMDPC equation has been utilized irrespective of the watershed's flow type, regardless of the fact that the value of K is the same for all channel surfaces. Using an equation originally developed for shallow concentrated flows for watersheds that also have sheet and/or channel flows seems, then, counterintuitive to the nature of the NRCS velocity method since it implies that all watersheds only have shallow concentrated flows. Likewise, the PMDPC's *PBIAS* poor performance indicated that the equation tended to overestimate in 60% of the selected watersheds, which leads to both smaller design rainfall intensity and peak flow values.



Figure 4. Emiliano Alcalá channel during (a) dry season and (b) rainy season.

Adjustment factors have been proposed (Table 5) for those empirical equations that showed R^2 values within the good and very good categories in Table S5. The adjustment factors are the inverse of the slope of the trend line equations obtained when T_c -simulated (empirical equations) and T_c -true (NRCS velocity method) values were scatter-plotted (trend line equations are summarized in Table S5), which makes the time of concentration values closer to the 1:1 line. This can be observed in Figure 5, where the majority of the time of concentration values obtained by means of the FAA and Kirpich equations before and after the proposed adjustment factors for the fifteen selected watersheds tended to follow the 1:1 line (black dashed line). Despite the improvement observed, the use of these adjustment factors is only recommended for the selected watersheds and/or for watersheds exhibiting similar characteristics.

Table 5. Proposed adjustment factors.

Time of Concentration Equation	Adjustment Factor	
	Stationary Conditions	Non-Stationary Conditions
FAA	2.17	2.13
NRCS Lag Equation	2.44	2.38
Kirpich	2.08	2.00
Kirpich-Adjusted	—	4.17
Johnstone-Cross	0.72	0.69
PMDPC	0.81	0.78
MinVivienda-FAA	2.27	2.22
MinVivienda-NRCS	2.38	2.33

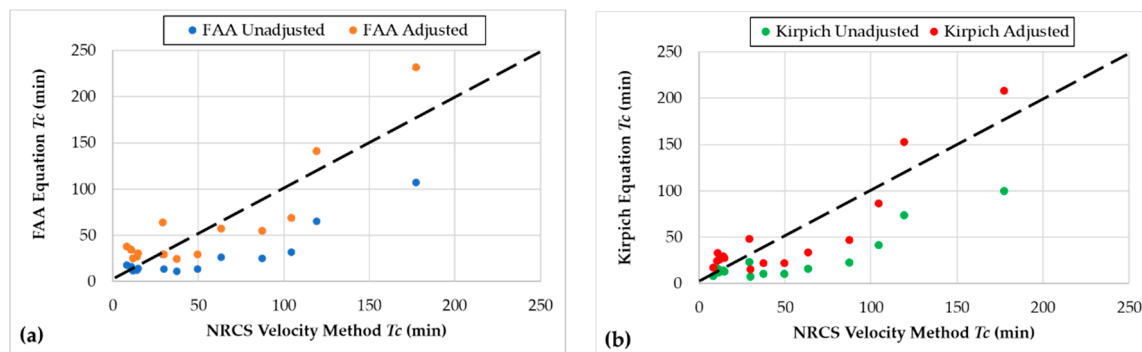


Figure 5. Time of concentration values before and after proposed adjustment factors. (a) FAA equation and (b) Kirpich equation.

Finally, the findings of this study are in line with other aforementioned studies with respect to the broad range of T_c values that designers have to typically deal with when performing a hydrological analysis. Based on the results herein and in the absence of measured rainfall and flow data, the authors recommend the estimation of T_c by means of the NRCS velocity method given the methodology's conceptual robustness and the availability of data of the methodology's variables.

5. Conclusions

Estimating the design rainfall and its associated runoff values for a given return period in ungauged watersheds can become daunting as it depends on various assumptions, such as the T_c computation among a large array of available empirical equations with different results. Thus, this task should be always accompanied by the best engineering judgement, which also involves an assessment of equations. Hydrologic and hydraulic (H&H) analysis for stormwater management such as dam-breach analyses; flood-prone area delineation and flood risk analyses; stream restoration; erosion, sediment, and contaminant load estimation, and conventional and sustainable drainage system design must be as accurate as possible when estimating all variables [10,80,81,83–86]. This includes the estimation of the T_c , which becomes relevant when using rainfall-runoff models and software programs requiring such input for the design runoff computation (e.g., rational method, HEC-HMS, WinTR-55, WinTR-20, StormCAD[®]/SewerCAD[®]). In this context, the results of this study shed light on the following: (a) the fact that selecting a given equation cannot be taken as a random process in a hydrological analysis, which implies being aware of the equation's background and limitations; (b) the identification of suitable equations for the area of study to start ruling out those that definitely do not perform well; (c) the need to implement, in any given location, a robust network of flow and rain gauges so as to have a better understanding of the local hydrology, which should include the derivation of local T_c equations for rural and urban areas given that none of equations utilized in this study performed well in all of the fifteen selected watersheds, and (d) conduct these types of studies in both urban and rural areas

given the different hydrological and hydraulic behavior of their watersheds and water courses. Finally, in the absence of measured rainfall and runoff data in a particular site, the authors recommend the use of the NRCS velocity method for T_c estimation, which, despite its own limitations [3], has proven to be reliable over the years since its inception. Validation of the resulting design runoff derived from the T_c estimation, whenever possible, should be performed via (a) hydraulic simulation using available water marks and/or (b) hydrologic simulation by means of software programs such as SWMM, which does not require the T_c as an input.

Supplementary Materials: The following are available online at <http://www.mdpi.com/2306-5338/7/3/47/s1>. Table S1: Identified flow paths information at each watershed within the study area, Table S2: Watersheds' additional variables used for T_c estimation, Table S3: T_c minimum, maximum, average, and standard deviation with P_2 under stationary conditions, Table S4: T_c minimum, maximum, average, and standard deviation with P_2 under non-stationary conditions, Table S5: Linear trend line equations between simulated and true T_c values, Table S6. Comparison between NRCS velocity method and T_c average, Table S7. Sheet flow equations, Table S8. Shallow concentrated flow equations, Table S9. Channel flow equations, Table S10. Additional T_c estimation methodologies, Table S11. Estimated T_c (min) values for each flow for P_2 under stationary conditions, Table S12. Estimated T_c (min) values for total length flow for P_2 under stationary conditions, Table S13. Estimated T_c (min) values for each flow for P_2 under non-stationary conditions, Table S14. Estimated T_c (min) values for total length flow for P_2 under stationary conditions, Table S15. Results of statistical variables with P_2 under stationary conditions, Table S16. Results of statistical variables with P_2 under non-stationary conditions.

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References

- McCuen, R.H.; Wong, S.L.; Rawls, W.J. Estimating Urban Time of Concentration. *J. Hydraul. Eng.* **1984**, *110*, 887–904. [[CrossRef](#)]
- McCuen, R.H.; Wong, S.L.; Rawls, W.J. Closure to “Estimating Urban Time of Concentration”. *J. Hydraul. Eng.* **1987**, *113*, 127–128. [[CrossRef](#)]
- U.S. Department of Agriculture Natural Resources Conservation Service (USDA-NRSC), C.E.D. *Urban Hydrology for Small Watersheds, Technical Release 55 (TR-55)*; U.S. Department of Agriculture Natural Resources Conservation Service (USDA-NRSC), C.E.D.: Washington, DC, USA, 1986.
- Huber, W.C. Discussion of “Estimating Urban. Time of Concentration”. *J. Hydraul. Eng.* **1987**, *113*, 122–124. [[CrossRef](#)]
- McCuen, R.H. Uncertainty Analyses of Watershed Time Parameters. *J. Hydrol. Eng.* **2009**, *14*, 490–498. [[CrossRef](#)]
- Sharifi, S.; Hosseini, S.M. Methodology for Identifying the Best Equations for Estimating the Time of Concentration of Watersheds in a Particular Region. *J. Irrig. Drain. Eng.* **2011**, *137*, 712–719. [[CrossRef](#)]
- Grimaldi, S.; Petroselli, A.; Tauro, F.; Porfiri, M. Temps de concentration: Un paradoxe dans l'hydrologie moderne. *Hydrol. Sci. J.* **2012**, *57*, 217–228. [[CrossRef](#)]
- Gericke, O.J.; Smithers, J.C. Review of methods used to estimate catchment response time for the purpose of peak discharge estimation. *Hydrol. Sci. J.* **2014**, *59*, 1935–1971. [[CrossRef](#)]
- Michailidi, E.M.; Antoniadis, S.; Koukouvinos, A.; Bacchi, B.; Efstratiadis, A. Timing the time of concentration: Shedding light on a paradox. *Hydrol. Sci. J.* **2018**, *63*, 721–740. [[CrossRef](#)]
- Perdikaris, J.; Gharabaghi, B.; Rudra, R. Reference Time of Concentration Estimation for Ungauged Catchments. *Earth Sci. Res.* **2018**, *7*, 58. [[CrossRef](#)]
- U.S. Department of Agriculture Natural Resources Conservation Service (USDA-NRCS). Time of Concentration. In *National Engineering Handbook (NEH)*; U.S. Department of Agriculture Natural Resources Conservation Service (USDA-NRCS): Washington, DC, USA, 2010; p. 29.
- Ontario Ministry of Transportation (OMTO). *Drainage Management Manual*; Ontario Ministry of Transportation (OMTO): Toronto, ON, Canada, 1997.

13. U.S. Department of Transportation Federal Highway Administration (USDOT-FHWA). Hydraulic Engineering Circular No. 22 (HEC-22). In *Urban Drainage Design Manual*, 3rd ed.; U.S. Department of Transportation Federal Highway Administration (USDOT-FHWA): Washington, DC, USA, 2009.
14. Texas Department of Transportation (TxDOT). *Hydraulic Design Manual (Revised)*; Texas Department of Transportation (TxDOT): Austin, TX, USA, 2019.
15. U.S. Department of Agriculture Natural Resources Conservation Service (USDA-NRCS). *Conservation Engineering Division WinTR-20 User Guide*; U.S. Department of Agriculture Natural Resources Conservation Service (USDA-NRCS): Washington, DC, USA, 2006.
16. *WinTR-55*; (Version 1.00.10); U.S. Department of Agriculture Natural Resources Conservation Service (USDA-NRCS): Washington, DC, USA, 2017.
17. *HydroCAD Stormwater*, (Version 10.1); Modeling System; HydroCAD Software Solutions LLC: Chocorua, NH, USA, 2020.
18. *Autodesk Storm and Sanitary Analysis*, (Version 2019); Autodesk Inc.: San Rafael, CA, USA, 2019.
19. *Storm Water Management Model (SWMM)*; (Version 5.1); United States Environmental Agency (USEPA): Washington, DC, USA, 2020.
20. *Hydrologic Modeling System (HEC-HMS)*; (Version 4.4.1); U.S. Army Corps of Engineers: Washington, DC, USA, 2018.
21. *SewerCAD*, (Version 10.02.00.55); Bentley Systems: Exton, PA, USA, 2019.
22. *StormCAD*, (Version 10.02.00.55); Bentley Systems: Exton, PA, USA, 2019.
23. San Diego State University (SDTU). Online Time of Concentration: Time of Concentration of Small Watersheds. Available online: <http://ponce.sdsu.edu/onlinetimeofconcentration.php> (accessed on 16 March 2020).
24. Manoj, K.C.; Fang, X. Estimating Time of Concentration of Overland Flow on Impervious Surface using Particle Tracking Model. In Proceedings of the World Environmental and Water Resources Congress 2014, Oregon, Portland, 1–5 June 2014; Volume 0300, pp. 30–37.
25. Yoo, C.; Lee, J.; Park, C.; Jun, C. Method for Estimating Concentration Time and Storage Coefficient of the Clark Model Using Rainfall-Runoff Measurements. *J. Hydrol. Eng.* **2014**, *19*, 626–634. [[CrossRef](#)]
26. De Almeida, I.K.; Almeida, A.K.; Steffen, J.L.; Alves Sobrinho, T. Model for Estimating the Time of Concentration in Watersheds. *Water Resour. Manag.* **2016**, *30*, 4083–4096. [[CrossRef](#)]
27. Sandoval-Eraza, W.; Toulkeridis, T.; Rodriguez-Espinosa, F.; Mora, M.J.M. Velocity and time of concentration of a basin—A renewed approach applied in the Rio Grande Basin, Ecuador. *IOP Conf. Ser. Earth Environ. Sci.* **2018**, *191*. [[CrossRef](#)]
28. Vahabzadeh, G.; Saleh, I.; Safari, A.; Khosravi, K. Determination of the best method of estimating the time of concentration in pasture watersheds (case study: Banadak Sadat and Siazakh Watersheds, Iran). *J. Biodivers. Environ. Sci.* **2013**, *3*, 150–159.
29. Gericke, O.J.; Smithers, J.C. Are estimates of catchment response time inconsistent as used in current flood hydrology practice in South Africa? *J. S. Afr. Inst. Civ. Eng.* **2016**, *58*, 2–15. [[CrossRef](#)]
30. De Almeida, I.K.; Almeida, A.K.; Anache, J.A.A.; Steffen, J.L.; Alves Sobrinho, T. Estimation on time of concentration of overland flow in watersheds: A review. *Geociencias* **2014**, *33*, 661–671.
31. Kaufmann de Almeida, I.; Kaufmann Almeida, A.; Garcia Gabas, S.; Alves Sobrinho, T. Performance of methods for estimating the time of concentration in a watershed of a tropical region. *Hydrol. Sci. J.* **2017**, *62*, 2406–2414. [[CrossRef](#)]
32. Salimi, E.T.; Nohegar, A.; Malekian, A.; Hoseini, M.; Holisaz, A. Estimating time of concentration in large watersheds. *Paddy Water Environ.* **2017**, *15*, 123–132. [[CrossRef](#)]
33. Wong, T.S.W.; Asce, F. Assessment of Time of Concentration Formulas for Overland Flow. *J. Irrig. Drain. Eng.* **2005**, *131*, 383–387. [[CrossRef](#)]
34. Fang, X.; Thompson, D.B.; Cleveland, T.G.; Pradhan, P.; Malla, R. Time of Concentration Estimated Using Watershed Parameters Determined by Automated and Manual Methods. *J. Irrig. Drain. Eng.* **2008**, *134*, 202–211. [[CrossRef](#)]
35. Wong, T. Evolution of Kinematic Wave Time of Concentration Formulas for Overland Flow. *J. Hydrol. Eng.* **2009**, *14*, 739–744. [[CrossRef](#)]
36. Manoj, K.C.; Fang, X.; Yi, Y.J.; Li, M.H.; Cleveland, T.G.; Thompson, D.B. Estimating time of concentration on low-slope planes using diffusion hydrodynamic model. In Proceedings of the World Environmental And Water Resources Congress 2012, Albuquerque, NM, USA, 20–24 May 2012; pp. 360–371. [[CrossRef](#)]

37. Longo, E.; Rooney, J. Development of a General Protocol to Enhance the Hydrological Analysis Techniques for Urban Catchments in Ireland. *Geosciences* **2018**, *8*, 252. [CrossRef]
38. Aristizábal, V.; Hernández, B.; Vélez, J. *Manual de Hidrología para Obras Viales Basado en el Uso de Sistemas de Información Geográfica*; Universidad Nacional de Colombia: Manizales, Colombia, 2012.
39. Otálvaro, M. *Consideraciones Generales para Realizar el Diseño Hidrológico e Hidráulico de Obras de Drenaje para Vías de Bajos Volúmenes de Tránsito. Caso de Estudio Pista de Prueba Urrao-Antioquia*; Universidad de Medellín: Medellín, Colombia, 2016.
40. Ochoa, T. *Hidrología Hidráulica y Socavación en Puentes*; ECOE Ediciones Ltda: Bogota, Colombia, 2017.
41. Instituto Nacional de Vías de Colombia (Invías). *Manual de Drenajes para Carreteras (Road Drainage Manual)*; Ministerio de Transporte de Colombia: Bogotá, Colombia, 2009.
42. Ministry of Housing, City, and Territory (MinVivienda), R. of C. *Technical Regulations for the Sector of Drinking Water and Basic Sanitation (RAS), Title D, Domestic Wastewater and Stormwater Collection Systems*; Ministerio de Vivienda de Colombia: Bogotá, Colombia, 2016.
43. Ministry of Housing, City, and Territory (MinVivienda), R. of C. *Resolution 0330 of 8 June 2017, Technical Guidelines for the Sector of Potable Water and Basic Sanitation (RAS)*; Ministerio de Vivienda de Colombia: Bogotá, Colombia, 2017.
44. Vélez, J.J.; Gutiérrez, A. Estimación del Tiempo de Concentración y tiempo de rezago en la Cuenca Experimental Urbana De La Quebrada San Luis, Manizales. *Dyna* **2011**, *78*, 58–71.
45. Torres, J.A.; Ordoñez, J.I.; Duque, R. Análisis de incertidumbre del tiempo de concentración en la modelación de eventos de tormenta en la cuenca alta del Río Magdalena. In Proceedings of the XXV Congreso Latinoamericano de Hidráulica, San José, Costa Rica, 9–12 September 2012.
46. Montoya, J.; Patiño, J. *Estudio Hidrometeorológico y Geomorfológico de Diez Quebradas del Valle de Aburrá*; Universidad EAFIT: Medellín, Colombia, 2001.
47. Avila, L.; AÁvila, H.; Sisa, A. A reactive early warning model for urban flash flood management. *World Environ. Water Resour. Congr.* **2017**, 372–382. [CrossRef]
48. Maidment, D.R. *Handbook of Hydrology*; MacGraw Hill: New York, NY, USA, 1993.
49. Welle, P.I.; Woodward, D. *Technical Note No. N4, Time of concentration (Revised)*; U.S. Department of Agriculture National Resources Conservation Service (USDA-NRCS): Chester, PA, USA, 2003.
50. Singh, J.; Knapp, H.V.; Arnold, J.G.; Demissie, M. Hydrological modeling of the Iroquois River watershed using HSPF and SWAT. *J. Am. Water Resour. Assoc.* **2005**, *41*, 343–360. [CrossRef]
51. Moriasi, D.N.; Arnold, J.G.; Van Liew, M.W.; Bingner, R.L.; Harmel, R.D.; Veith, T.L. Model Evaluation Guidelines. 2007; Volume 50, pp. 885–900. Available online: https://www.researchgate.net/publication/43261199_Model_Evaluation_Guidelines_for_Systematic_Quantification_of_Accuracy_in_Watershed_Simulations (accessed on 8 May 2020). [CrossRef]
52. Moriasi, D.N.; Gitau, M.W.; Pai, N.; Daggupati, P. Hydrologic and Water Quality Models: Performance Measures and Evaluation Criteria. *Trans. ASABE* **2015**, *58*, 1763–1785. [CrossRef]
53. Boskidis, I.; Gikas, G.D.; Sylaios, G.K.; Tsihrintzis, V.A. Hydrologic and Water Quality Modeling of Lower Nestos River Basin. *Water Resour. Manag.* **2012**, *26*, 3023–3051. [CrossRef]
54. Rauf, A.; Ghumman, A. Impact Assessment of Rainfall-Runoff Simulations on the Flow Duration Curve of the Upper Indus River—A Comparison of Data-Driven and Hydrologic Models. *Water* **2018**, *10*, 876. [CrossRef]
55. González-Álvarez, Á.; Coronado-Hernández, O.; Fuertes-Miquel, V.; Ramos, H. Effect of the Non-Stationarity of Rainfall Events on the Design of Hydraulic Structures for Runoff Management and Its Applications to a Case Study at Gordo Creek Watershed in Cartagena de Indias, Colombia. *Fluids* **2018**, *3*, 27. [CrossRef]
56. Instituto de Hidrología Meteorología y Estudios Ambientales (IDEAM) Atlas Climatológico de Colombia 1981–2010. Available online: <http://atlas.ideam.gov.co/visorAtlasClimatologico.html> (accessed on 8 May 2020).
57. Centro de Investigaciones Oceanográficas e Hidrográficas (CIOH). *General Circulation of the Atmosphere in Colombia*; CIOH: Cartagena de Indias, Colombia, 2010.
58. Perez, G.; Salazar, I. *La Pobreza en Cartagena: Un Análisis por barrios*; Banco de La República: Cartagena de Indias, Colombia, 2007; Volume 94.
59. EPA-Cartagena & UDC. Diseño del sistema inteligente de monitoreo de la calidad ambiental del distrito de Cartagena. *Estud. Integr. La Calid. Ambient. Del Sist. Caños Y Lagos* **2015**, *2*.

60. ArcGIS, (version 10.6.1); Esri Inc.: Redlands, CA, USA, 2019.
61. Crippen, R.; Buckley, S.; Agram, P.; Belz, E.; Gurrola, E.; Hensley, S.; Kobrick, M.; Lavallo, M.; Martin, J.; Neumann, M.; et al. NASADEM Global Elevation Model: Methods and Progress. *Int. Arch. Photogramm. Remote Sens. Spat. Inf. Sci.* **2016**, *XLI-B4*, 125–128. [[CrossRef](#)]
62. Azizian, A. Uncertainty Analysis of Time of Concentration Equations based on First-Order-Analysis (FOA) Method. *Am. J. Eng. Appl. Sci.* **2018**, *11*, 327–341. [[CrossRef](#)]
63. Iowa State University's Institute for Transportation (InTrans). Iowa Statewide Urban Design and Specifications Program Urban Hydrology and Runoff (2B-3 Time of Concentration). In *SUDAS Design Manual*; Iowa State University: Ames, IA, USA, 2020.
64. Miller, W.C. Evolving a shortcut for design of storm sewers. *Munic. Util.* **1951**, *89*, 42–59.
65. Cleveland, T.G.; Thompson, D.B.; Ming-Han Li, X.F. *Guidance for Estimation of Time of Concentration in Texas for Low-Slope Conditions (Report # 0-6382-P1)*; Texas Department of Transportation: Austin, TX, USA, 2012.
66. Federal Aviation Agency (FAA). *Airport Drainage*; Department of Transport Advisory Circular: Washington, DC, USA, 1970.
67. Espey, W.H.; Winslow, D.E. *The Effects of Urbanisation on Unit Hydrographs for Small Watersheds*; Tracor Incorporated: Houston, Austin, TX, USA, 1968; Report No. 68/975U and 1006/U.
68. Kerby, W. Time of concentration for overland flow. *Civ. Eng.* **1959**, *29*, 174.
69. Kirpich, Z. Time of concentration of small agricultural watersheds. *Civ. Eng.* **1940**, *10*, 362.
70. Williams, G.B. Flood discharges and the dimensions of spillways in India. *Engineering* **1922**, 134–321.
71. Johnstone, D.; Cross, W.P. *Elements of Applied Hydrology*; Ronald Press: New York, NY, USA, 1949.
72. Sheridan, J.M. Hydrograph Time Parameters for Flatland Watersheds. *Trans. ASAE* **1994**, *37*, 103–113. [[CrossRef](#)]
73. Alcaldía Mayor de Cartagena de Indias (AMCI). *Estudios y Diseños del Plan. Maestro de Drenajes Pluviales del Distrito de Cartagena*; Consorcio Consultores Cartageneros: Cartagena de Indias, Colombia, 2009.
74. Roussel, M.C.; Thompson, D.B.; Fang, X.; Cleveland, T.G.; Garcia, C.A. *Time-Parameter Estimation for Applicable Texas Watersheds*; Texas Department of Transportation: Beaumont, TX, USA, 2005.
75. Te Chow, T.; Maidment, D.R.; Mays, L.W. *Applied Hydrology*, 1st ed.; McGraw-Hill: New York, NY, USA, 1988; ISBN 0-07-010810-2.
76. ENVI, (version 5.1); Harris Geospatial Solutions Inc.: Boulder, CO, USA, 2019.
77. Gupta, H.V.; Sorooshian, S.; Yapo, P.O. Status of Automatic Calibration for Hydrologic Models: Comparison with Multilevel Expert Calibration. *J. Hydrol. Eng.* **1999**, *4*, 135–143. [[CrossRef](#)]
78. Legates David, R.; McCabe, G.J., Jr. Evaluating the use of goodness-of-fit measures in hydrologic and hydroclimatic model validation. *Water Resour. Res.* **1999**, *35*, 233–241. [[CrossRef](#)]
79. González-Álvarez, Á.; Vilorio-Marimón, O.M.; Coronado-Hernández, Ó.E.; Vélez-Pereira, A.M.; Tesfagiorgis, K.; Coronado-Hernández, J.R. Isohyetal maps of daily maximum rainfall for different return periods for the Colombian Caribbean Region. *Water* **2019**, *11*, 358. [[CrossRef](#)]
80. Duan, W.; Takara, K.; He, B.; Luo, P.; Nover, D.; Yamashiki, Y. Spatial and temporal trends in estimates of nutrient and suspended sediment loads in the Ishikari River, Japan, 1985 to 2010. *Sci. Total Environ.* **2013**, *461–462*, 499–508. [[CrossRef](#)]
81. Duan, W.; He, B.; Nover, D.; Fan, J.; Yang, G.; Chen, W.; Meng, H.; Liu, C. Floods and associated socioeconomic damages in China over the last century. *Nat. Hazards* **2016**, *82*, 401–413. [[CrossRef](#)]
82. Martins, L.; Diéz-Herrero, A.; Bodoque, J.; Bateira, C. Time of Concentration equations: The role of morphometric uncertainties in flood risk analysis and management. In Proceedings of the EGUGA, Vienna, Austria, 17–22 April 2016. EPSC2016-15660.
83. Hoogstraat, G.K. Flood hydrology and dam-breach hydraulic analyses of four reservoirs in the Black Hills, South. Dakota: U.S. In *Geological Survey Scientific Investigations Report 2011–5011*; U.S. Geological Survey (USGS): Reston, VA, USA, 2011.
84. Poleto, C.; Rutinéia, T. Sustainable Urban Drainage Systems. In *Drainage Systems*; IntechOpen: London, UK, 2012; ISBN 978-953-51-0243-4.

85. Jiang, X.; Yang, L.; Tatano, H. Assessing Spatial Flood Risk from Multiple Flood Sources in a Small River Basin: A Method Based on Multivariate Design Rainfall. *Water* **2019**, *11*, 1031. [[CrossRef](#)]
86. Federal Emergency Management Agency (FEMA). Guidance for Flood Risk Analysis and Mapping. In *Hydrology: Rainfall-Runoff Analysis*; U.S. Department of Homeland Security: Washington, DC, USA, 2019.



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