

Rainfall–runoff modelling of ephemeral streams in the Valencia region (eastern Spain)

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Abstract:

This paper presents preliminary results from the application of a transfer-function rainfall–runoff model to ephemeral streams in Mediterranean Spain. Flow simulations have been conducted for two small catchments (Carraixet and Poyo basins), located in close proximity to one another yet with significantly different geological characteristics. Analysis of flow simulations for a number of high-flow events has revealed the dominant influence of the rainfall on the catchment response, particularly for high-rainfall events. Particular success has been attained modelling the highest magnitude events in both catchments and for all events in the faster responding (Poyo) catchment. In order to investigate the viability of the model for forecasting floods in ungauged catchments, additional investigations have been conducted by calibrating the model for one catchment (donor catchment) and then applying it to another (receptor catchment). The results indicate that this can be successful when either the donor catchment is a fast response catchment or when the model is calibrated using a high-magnitude event in the donor catchment, providing that the modelled receptor catchment event is of a lower magnitude. Copyright © 2002 John Wiley & Sons, Ltd.

KEY WORDS ephemeral stream; transfer-function; rainfall–runoff model; simulation; donor catchment; receptor catchment; ungauged catchment; hydrograph; flood

INTRODUCTION: RAINFALL–RUNOFF CONVERSION IN EPHEMERAL STREAMS

The drainage basin plays a fundamental role in any hydrological study of processes concerning the conversion of rainfall to runoff. The catchment can be considered as a system that responds to certain *inputs*, the principal one being precipitation, and converting them into *outputs*, such as river flow and sediment movement. During this conversion process, the basin attempts to adapt its form to the system's energy conditions, in an attempt to reach geomorphological equilibrium (Morisawa, 1985).

The influence of drainage basin morphology and dynamics on catchment hydrographs is simulated, to a greater or lesser extent, in all rainfall–runoff conversion models, from the physically based models, which operate at a detailed scale, to the empirical, and black-box, models. In the former, the presence of the catchment is clearly apparent, because the model attempts to imitate each hydrological process, based on the laws of conservation of mass, energy and momentum. Many researchers make use of geographical information systems (GISs) for hydrological studies, applying distributed models, which operate at a pixel level (Maidment, 1993). In contrast, black-box models utilize a statistical transfer function to relate the system inputs and outputs. This function should, in principle, reflect the implicit influence of the underlying physical system, where rainfall–runoff conversion occurs (Klemes, 1981).

Although the influence of the catchment on the rainfall–runoff conversion process is evident, the way in which its influence takes effect depends on the type of fluvial system and, consequently, on the dominant

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1 discharge regime. According to Graf (1988) 'processes control forms in high magnitude events, while forms
2 control processes in low magnitude events'. The fluvial systems of humid regions respond to high-frequency
3 low-energy events with a preponderance of throughflow. Consequently, basins have a marked influence in
4 the rainfall–runoff conversion process. However, in the case of Mediterranean and semi-arid systems, the
5 dominant events occur infrequently, but with high magnitude. In these systems surface-runoff predominates
6 during floods, and the influence of the basin is reduced substantially and the influence of rainfall increases
7 markedly. In fact, the most notable hydrological characteristic of Mediterranean ephemeral streams is that
8 channels remain dry for most of the year, because they are unconnected to aquifers and therefore lack baseflow
9 (Mateu, 1988; Segura, 1990). Quickflow depends almost exclusively on the rainfall, and often is related to
10 intense 'high-energy' rainfall, leading to flash floods (Segura and Camarasa, 1996).

11 The speed of the processes and the importance of high intensity rainfall have meant that, historically,
12 simple models, very dependent upon precipitation, such as the unit hydrograph, have worked adequately
13 for predicting flood hydrographs in this environment (Marco, 1989; Abdulrazzak, 1989; García Bartual and
14 Marco, 1990). Even when more advanced unit hydrographs have been used, such as the geomorphological
15 unit hydrograph (GUH) (Rodríguez-Iturbe and Valdés, 1979; Rodríguez-Iturbe *et al.*, 1982; Corradini *et al.*,
16 1986; Rosso and Caroni, 1987; Nouh, 1990), incorporating morphometric catchment indicators, the results
17 have demonstrated the importance of the *scale* parameter—related to the catchment's rate of response and the
18 rainfall intensity—compared with the *form* parameter—related to the geomorphology of the basin (Corradini
19 *et al.*, 1986; Camarasa, 1995).

20 Thus, from the point of view of extrapolating rainfall–runoff models from gauged to ungauged basins, the
21 distinction between high- and low-magnitude events, or between fast and slow catchment responses, could
22 be more significant than the different geomorphological features between basins (always assuming that the
23 reference frame relates to the same morphoclimatic context). The magnitude and the internal structure of
24 inputs, together with the implicit type of lumped catchment reaction, could become the main variables to
25 take into account for hydrograph modelling. This observation could have very interesting implications for
26 real-time catastrophic flood forecasting, using mathematical simulations. Within the context of a region with
27 similar characteristics, it could mean significant saving in simulation models for gauged catchments and a
28 valid tool for hydrograph prediction in ungauged basins.

29 This paper shows the preliminary results of the application to Mediterranean basins of a simple
30 rainfall–runoff conversion model (*transfer-function model—TFM*), developed at the University of Salford
31 (England) for flood hydrograph simulation in humid regions. This model, which is strongly dependent on the
32 rainfall structure, was applied to several flood events, in two Valencian ephemeral streams: the Barranc de
33 Carraixet River (128 km²) and the Rambla de Poyo River (187 km²). The model's extrapolation capability
34 was tested in both temporal (extrapolations between different events in the same basin) and spatial dimensions
35 (extrapolating from one basin to another).
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38 DESCRIPTION OF RIVER CATCHMENTS

39 Mediterranean 'ramblas' and 'barrancos' are fluvial systems with ephemeral streams, where basins have steep
40 slopes, scarce vegetation and poorly developed soils. They frequently occur over a substrate composed of
41 permeable materials (limestones and dolomites), and are geomorphologically complex, which is strongly
42 influenced by underlying geological structures as a consequence of germanic relief, formed by distensive
43 forces, during the Alpine Orogeny. The drainage networks are, in many cases, still not fully organized, owing
44 to the low frequency of events with sufficient energy to affect the geomorphology. Both the basins studied,
45 the Barranc del Carraixet and la Rambla de Poyo (Figure 1), located in the eastern section of the Iberica
46 Range (east of Spain), belong to this morphoclimatic environment.

47 The first basin used in this study is the drainage area of a gauged sub-basin (128 km²) belonging to the
48 Barranc de Carraixet basin (311 km²). The stream flows into the Mediterranean Sea north of Valencia, Spain's
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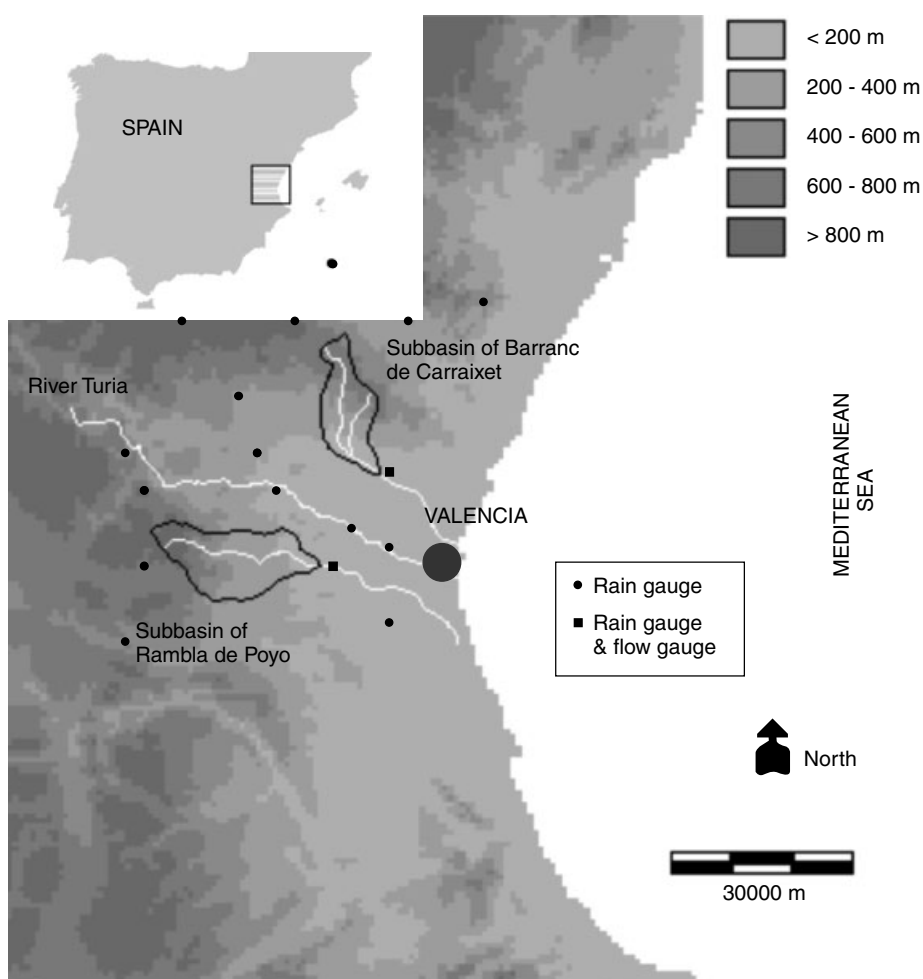


Figure 1. Area of study

third largest city, so the catchment includes areas of intensive land use. The catchment has an elongated and asymmetrical shape (elongation rate of 0.33), marked relief (over 36% of the area has slopes steeper than 30%), and is strongly influenced by underlying geological structures.

The geomorphology of the basin is very varied (Camarasa, 1995). Three main sectors can be differentiated: the headwaters, an intermediate area of alluvial fans and a calcareous subtabular platform.

1. The headwaters are in the mountainous area formed by highly resistant mesozoic materials (limestone, dolomites and sandstone). It has a typical germanic relief: highly faulted and very steep. The upper catchment has high permeability.
2. The middle catchment consists of a series of deep alluvial fans and piedmonts formed during several Quaternary phases, overlying a faulted and sunken Pliocene graben.
3. A calcareous subtabular platform borders the south of the middle basin.

The Rambla de Poyo drains the area south of Valencia city, between the basins of the rivers Turia and Jucar, and flows into the Albufera de Valencia (a coastal lagoon). The gauged sub-basin has three principal tributaries: the Barranco Grande, the Barranco de la Cueva Morica and the Rambla de Gallo-Chiva.

Q1

1 Geomorphologically it can be divided into three units, showing a similar configuration to those of the
2 Barranc de Carraixet.

- 3
- 4 1. Mountainous headwaters, developed over resistant limestone–dolomitic materials, forming steep relief.
- 5 2. An intermediate sector overlying a sunken graben, where large-scale alluvial fans mark the contact between
- 6 the Mesozoic headwaters and the graben.
- 7 3. An eastern sector, with subtabular relief, made up of highly fragmented clay-mudstones.

8
9 The fifth-order drainage network is fairly disorganized, with an elongation ratio of 0.4.

10 The main difference between both basins is the permeability: Rambla de Poyo has a greater proportion of
11 underlying impermeable rocks (35%), compared with Barranc de Carraixet (3%). Consequently, for the same
12 inputs, the Rambla de Poyo has a faster response rate than Barranc de Carraixet. For similar storm events,
13 the average response lag-time for the Poyo basin is 3 h, compared with 5 h for the Carraixet. The average
14 quickflow velocity is 2.3 m s^{-1} in the Poyo catchment compared with 0.5 m s^{-1} in the Carraixet (Camarasa,
15 1995).

17 FLOOD EVENT DATA

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19 Although in Spain the most dangerous floods occur in a few hours, the national hydrological network provides
20 data only every day. As a consequence, processes are measured with an incorrect time interval. Studies based
21 on daily data are not precise enough to understand the formation of flash floods and, therefore, forecasting
22 hydrographs for small semi-arid basins is a highly difficult task.

23 Some important flood events, which affected the Mediterranean region during 1982 and the north coast
24 during 1983, led the government to adopt some structural measures for flood prediction. A new hydrological
25 network *Sistema Automático de Información Hidrológica (SAIH*; automatic system of hydrological informa-
26 tion) is being installed by the main hydrographic authorities, starting with those areas where flood risk is more
27 important. Thus, the River Jucar Hydrographic Authority has been the first to produce reliable hydrological
28 data every 5 min.

29 This investigation is supported by rainfall (17 rain gauges) and river flow (two gauges) data provided
30 for this hydrographic authority, for three flood events occurring on 11–12 November 1988, 8–9 September
31 1990 and 16–17 April 1991. Owing to the geographical proximity of the two catchments they were affected
32 simultaneously by the same storms, enabling effective comparisons to be made between the two for the same
33 events. The two autumnal events (November 1988 and September 1990) were of higher magnitude than that
34 in April 1990 (Table I).

37 FLOOD HYDROGRAPH SIMULATION USING A SIMPLE RAINFALL–RUNOFF TRANSFER MODEL

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39 A wide range of lumped rainfall–runoff forecasting models (models that simulate river flow using a single
40 rainfall input taken to be representative of rainfall across the entire catchment) have been developed. O’Connell
41 and Clark (1981) and Reed (1984) provide reviews of many of these. These include conventional methods
42 such as the unit hydrograph (Chander and Shanker, 1984), S-curve, Clark method, linear cascade reservoir
43 model, conceptual models and non-linear storage models (Bobinski and Mierkiewicz, 1986; Corradini *et al.*,
44 1986; Corradini and Melone, 1987).

45 The hydrograph simulation model used in this work consisted of a simple rainfall–runoff transfer function
46 that has been developed at the University of Salford by Dr Tilford and his co-workers.

47 Rainfall–runoff models are widely used for predicting flood hydrographs because they have a simple
48 mathematical structure and often are relatively easy to use. Within this class of models the most conventional
49 and most widely used is the unit hydrograph, an approach first presented by Sherman (1932) and later

Table I. Flood event characteristics

Basin	Hydrological features	Events		
		November 1988	September 1990 ^a	April 1991
Barranc de Carraixet	Total rainfall (mm)	72.8	152.7	23.9
	Net rainfall (mm)	5.4	9.9	1.4
	Runoff threshold (mm)	38	83	13.5
	Runoff coefficient (%)	7	5.6	5.5
	Runoff deficit (10 ⁶ m ³)	8.6	18	2.89
	Total flood runoff (10 ⁶ m ³)	0.69	1.27	0.17
	Peak discharge (m ³ s ⁻¹)	21.7	73.55	9.5
	Lag time	6 h 5 min	3 h 15 min	5 h 55 min
	Time to peak	1 h 55 min	1 h 40 min	1 h 25 min
	Stream velocity (m s ⁻¹)	0.38	0.72	0.39
Rambla de Poyo	Total rainfall	74.8	71.9	18.8
	Net rainfall	6	0.5	2.4
	Runoff threshold (mm)	38	59.7	7.9
	Runoff coefficient (%)	8	0.7	12
	Runoff deficit (10 ⁶ m ³)	12.7	13.2	3.04
	Total flood runoff (10 ⁶ m ³)	1.2	0.09	0.44
	Peak discharge (m ³ s ⁻¹)	193	65.35	29.5
	Lag time	1 h 50 min	4 h 10 min	3 h 10 min
	Time to peak	15 min	30 min	55 min
	Stream velocity (m s ⁻¹)	3	—	1.8

^a September event in Rambla de Poyo has not been used for simulation because quality of hydrological data is inadequate.

developed in different versions by many authors (e.g. Nash, 1958; Rodriguez-Iturbe and Valdés, 1979; Gupta and Gaymire, 1983; Rosso, 1984; Corradini *et al.*, 1986).

The transfer-function model (TFM) described by Box and Jenkins (1976) is a relatively recent alternative to unit hydrographs. Although originally developed for control applications in electrical engineering, transfer functions present a number of characteristics that make them appropriate for river flow modelling, including: (i) structural simplicity; (ii) parametric efficiency ('parsimony'); (iii) self-correcting capacity; and (iv) potential for incorporating parameter updating.

In recent years a number of researchers have reported attempts to simulate the rainfall–runoff process at catchment scale using such models in temperate humid environments (e.g. Moore and O'Connell, 1978; Moore, 1980; Cluckie, 1993; Jakeman and Hornberger, 1993; Young and Beven, 1994; Nalbantis, 1995; Ramos *et al.*, 1995). Indeed, developments have been such that transfer-function rainfall–runoff models are used routinely for operational flow forecasting in several regions of the UK: in the southwest region (Birks *et al.*, 1989; Aucott *et al.*, 1992); on the River Irwell in the northwest (Cluckie and Owens, 1987); in eastern England on the River Cam and Willow Brook (Tilford, 1993). They also have been developed for real-time flood-flow modelling (e.g. Cluckie *et al.*, 1989; Wilke and Barth, 1991; Lees *et al.*, 1994; Lees, 1997). The particular characteristics of rainfall–runoff conversion processes in semi-arid environments suggest that, because of the dependence of the hydrograph on intense rainfall, this model might be applied successfully to Mediterranean basins during flood events.

The transfer-function rainfall–runoff model

Transfer-function rainfall–runoff models (TFMs) forecast future hydrographs using measurements of current and previously observed rainfall and flows. Powell (1985) and Owens (1986) demonstrated that the rainfall–runoff transformation process could be simulated satisfactorily by a single-input–single-output

(SISO) transfer function with the following structure●

Q2

$$\hat{Q}_t = a_1 Q_{t-1} + a_2 Q_{t-2} + \cdots + a_p Q_{t-p} + b_0 R_{t-\tau} + b_1 R_{t-1-\tau} + \cdots + b_q R_{t-q-\tau} \quad (1)$$

where \hat{Q}_t is the forecast flow for time t , Q_t and R_t are the observed flow and rainfall respectively at time t , τ is a pure time delay, and a and b are rainfall and flow parameters respectively. The time between successive rainfall and flow observations, i.e. the time between t and $t + 1$ defines the model time interval \mathfrak{S} .

A key feature of the model structure described by Equation (1) is the use of past observed flow values to correct the model forecasts via a closed (feedback) loop. This self-correction capability provides robustness to the model when operated in a forecasting mode because forecasts of future flow are dependent upon past observed values.

The rainfall input R_t can be defined as total or effective rainfall. When effective rainfall is used the TFM is, to a certain extent, equivalent to the unit hydrograph (Jakeman and Hornberber, 1993), albeit more efficient (or 'parsimonious') parametrically. When total rainfall is used, a factor is used to scale the rainfall input. In accordance with the model output (i.e. runoff) in order to maintain a mass balance. The scaling factor Δ is applied according to Equation (2)

$$\hat{Q}_t = a_1 Q_{t-1} + a_2 Q_{t-2} + \cdots + a_p Q_{t-p} + \Delta_t (b_0 R_{t-\tau} + b_1 R_{t-1-\tau} + \cdots + b_q R_{t-q-\tau}) \quad (2)$$

Δ_t being updated through the course of an event, according to

$$\Delta_t = \mu \Delta_{t-1} + (1 - \mu) \left[\frac{y_t - (a_1 y_{t-1} + \cdots + a_p y_{t-p})}{b_1 u_{t-1} + \cdots + b_q u_{t-q}} \right] \quad (3)$$

The pure time delay τ enables the model to simulate flow in catchments where: (i) rainfall is consistently confined to an area of the catchment upstream of the gauging station (e.g. see O'Connell and Clarke, 1981) and/or (ii) where initial losses are large (i.e. as a surrogate for initial storage). The second is especially relevant when modelling highly permeable catchments, and, together with the model's flexibility for simulating hydrographs in environments where the river flow is highly dependent on the rainfall, was the prime reason for the application of this model to ephemeral streams in Mediterranean Spain.

Specification and calibration of the transfer function model

The TFM relates total rainfall to runoff by an empirically derived input–output relationship determined by off-line calibration. This calibration process consists of three stages: determination of optimal model interval; determination of optimal model structure; and parameter estimation. Model structure, order and interval are interrelated. The model order is defined as the total number of parameters in the model. Structure is defined as the number of a parameters, b parameters and time delay τ . Finally, the model interval is the time interval that the model uses during operation (which may be equal to or greater than the data interval).

Model order determination (structure identification) aims to identify the optimal model structure, i.e. the structure combining the attributes of parsimony and forecasting accuracy. The optimal model will have sufficient parameters to describe catchment response adequately while avoiding overparameterization. In addition to the fact that it is unrealistic to estimate a large number of model parameters from a limited (and noisy) data set, parsimony is also desirable because the model structure influences computational (run) time and the number of past data required for forecasting.

Determination of optimal structure is linked intrinsically to the selection of model time interval and, by definition, the optimal model structure can be identified only if the optimal model interval has been ascertained. Failure to identify the optimal model interval will result in suboptimal forecasting performance. Small model interval will necessitate an increased number of parameters and a potentially poorer model performance (owing to superfluous additional information), whereas a larger interval will result in fewer parameters and poorer performance (owing to significant information being lost or missed). In the rainfall–runoff process the optimal

1 model interval is governed jointly by catchment response dynamics and the characteristics of the rainfall field,
2 and consequently will vary from catchment to catchment and from event to event.

3 The model (time) interval \mathfrak{S} is determined prior to parameter estimation by using an objective technique first
4 proposed for the identification of digital control systems by Isermann (1981) and developed for rainfall–runoff
5 models by Powell and Cluckie (1984). Powell (1985) showed that the step response of a system (i.e. the integral
6 of the impulse response) could be used to determine the optimal time interval $\mathfrak{S}_{\text{opt}}$ by the following rule

$$7 \quad 8 \quad 9 \quad \frac{T_{90}}{2} < \mathfrak{S}_{\text{opt}} < \frac{T_{90}}{10} \quad (4)$$

10 where T_{90} is the time for the system to rise to 90% of the steady state output as determined from the step
11 response.

12 Once the model interval has been defined, the model structure can be determined, i.e. the number of rainfall
13 and flow parameters (p and q respectively). Several methods for the determination of model structure exist.
14 For this study, an equal-order model search technique developed by Owens (1986) has been adopted by virtue
15 of simplicity and ease of use. In the search, parameters are estimated sequentially for equal-order models (i.e.
16 $p = q$) from a 2,2 model structure upwards, until an increase in model order no longer results in a significant
17 improvement in model accuracy.

18 The following evaluation criteria are used: (i) error statistics for the model convolution of the calibration
19 data; (ii) model impulse response (physically viable, i.e. positive and stable); (iii) parameter redundancy, i.e.
20 an unnecessarily high (overspecified) model order.

21 For rural catchments in the UK experience has shown that if the model interval is identified correctly, process
22 rarely has to be repeated beyond a 6,6 structure, with most catchments being modelled adequately with a
23 model order less than eight. Model instability (e.g. to a unit pulse input) usually signifies overparameterization.
24 Once the optimal equal-order model has been found the number of a parameters is reduced until an increase
25 in modelling error arises.

26 Harpin (1982) conducted a detailed analysis of the relative performance of a range of ‘conventional’
27 recursive parameter estimation algorithms. He concluded that the ordinary recursive least squares (ORLS)
28 linear estimator provided satisfactory convergence to the final (optimal) parameter values, was inherently
29 stable and robust, and was satisfactory for hydrological modelling (see also Cluckie and Harpin, 1982:
30 Cluckie *et al.*, 1980). The algorithm sequentially steps through the rainfall and flow data pairs, progres-
31 sively updating the parameter estimates, attempting to minimize the squares of the one-step-ahead forecast
32 errors.

33 Model identification/calibration produces a model that exhibits a generalized response. Assuming calibration
34 has been performed adequately, the model will have a percentage runoff closely corresponding to the
35 average percentage runoff of the rainfall/runoff calibration time-series. As the number of events used for
36 calibration increases, the model’s ability to produce high-accuracy forecasts on average increases, although
37 its performance in less typical flood conditions (i.e. those where the antecedent catchment condition or the
38 rainfall profile, distribution or intensity departs significantly from the ‘norm’) deteriorates.

40 *Application of the transfer function model*

41 In this study TFM was first applied to each event in each catchment separately, with the aim of checking
42 its ability to simulate the observed event hydrographs. The TFM model parameters were calibrated for two
43 system inputs: total rainfall and effective (net) rainfall.

44 For the first case, the catchment average total rainfall was calculated from the rain-gauge data using Thiessen
45 polygons (the calibration of the model implicitly accounting for losses, through the parameter estimation).

46 For the second case, the effective rainfall was estimated, before applying the TFM, from the total rainfall
47 using the US Soil Conservation Service (SCS) empirical loss model method (US Soil Conservation Service,
48 1972) adapted to Spanish conditions by Temez (1978). This approach has been successfully used in a variety of
49

1 applications in Mediterranean catchments (Camarasa and Garcia Bartual, 1991; Camarasa, 1995). According
 2 to this method all the rainfall at the start of the storm is lost (through evapotranspiration and infiltration).
 3 Surface runoff commences only once a threshold soil absorption level (P_0) is reached, as defined by the
 4 following expression

$$\sum E = 0 \text{ for } \sum P \leq P_0 \quad (5a)$$

$$\sum E = \frac{(\sum P - P_0)^2}{\sum P + 4P_0} \text{ for } \sum P > P_0 \quad (5b)$$

11 where $\sum P$ is the accumulated rainfall from the beginning of the storm, $\sum E$ is runoff or effective rainfall
 12 and P_0 is the runoff threshold.

13 Although P_0 can be estimated from catchment slope, soil characteristics and land use, in this study P_0 was
 14 estimated from the runoff coefficient, C , which is defined after Camarasa and Garcia-Bartual (1991) as

$$C = \frac{\sum E}{\sum P} = \frac{(\sum P - P_0)^2}{(\sum P + 4P_0) \sum P} \quad (6)$$

19 The model was applied using a model time interval of 15 min. The results of the event simulations
 20 undertaken (Table II) illustrate:

- 21 1. A model input of effective rainfall produced much better simulations than for total rainfall. The tests
 22 demonstrate the applicability of the SCS losses method. For example, for the same catchment (Rambla
 23 de Poyo) and the same event (November 1988) the results were much better using net rainfall (1, 5 order
 24 model and 9.97 RMS) than using total rainfall (30, 30 order model and 33.27 RMS).
- 25 2. The model's ability to reproduce the shape of the total hydrograph in 'high energy' events and/or in the
 26 faster response catchment (i.e. the Rambla de Poyo). In contrast, the TFM was unsuccessful in simulating
 27 smaller flood flows in the slower response catchment (i.e. the Barranc de Carraixet).

31 Table II. Results of transfer-function model calibration

Basin	Event	Input	Model structure (number of parameters a and b)	Time delay, τ (min)	Error analysis		
					Mean error	Root mean square (RMS) error	RMS of model convolution
Barranc de Carraixet	November 1988	Total rainfall	3,5	30	-0.015	0.924	5.11
		Net rainfall	5,5	30	0.003	0.877	2.79
	September 1990	Total rainfall	1,5	0	-0.053	2.318	3.85
		Net rainfall	1,5	0	-0.12	1.558	2.77
Rambla de Poyo	April 1991	Total rainfall	1,6	150	-0.029	0.763	3.43
		Net rainfall	1,6	150	-0.002	0.752	2.89
	November 1988	Total rainfall	30,30	0	0.357	10.18	33.27
		Net rainfall	1,5	0	0.207	5.95	9.97
	April 1991	Total rainfall	1,6	0	-0.053	1.811	2.69
		Net rainfall	1,2	0	0.008	1.547	2.56

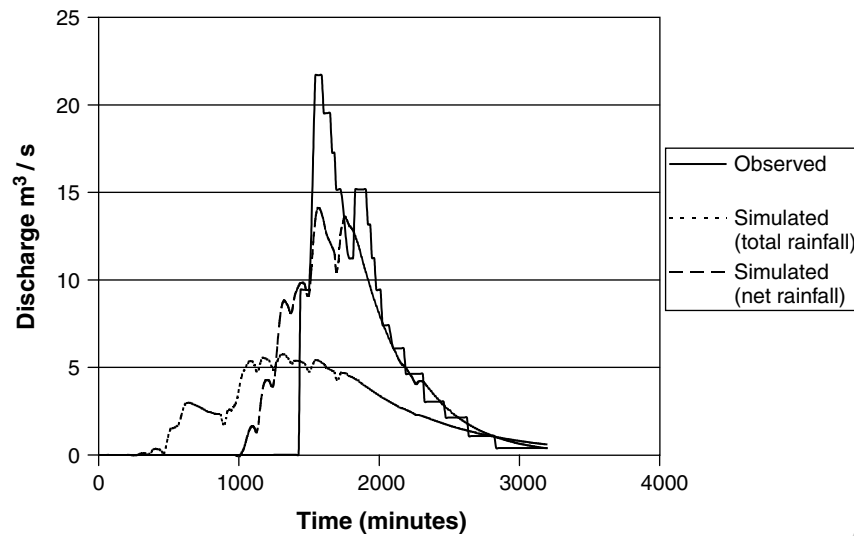


Figure 2. Comparison between simulations carried on with total and net rainfall inputs (Carraixet, November 1988 event). Error in peak estimation: using total rainfall—73.7% in peak discharge and 225 min out of phase; using net rainfall—35% in peak discharge and 30 min out of phase

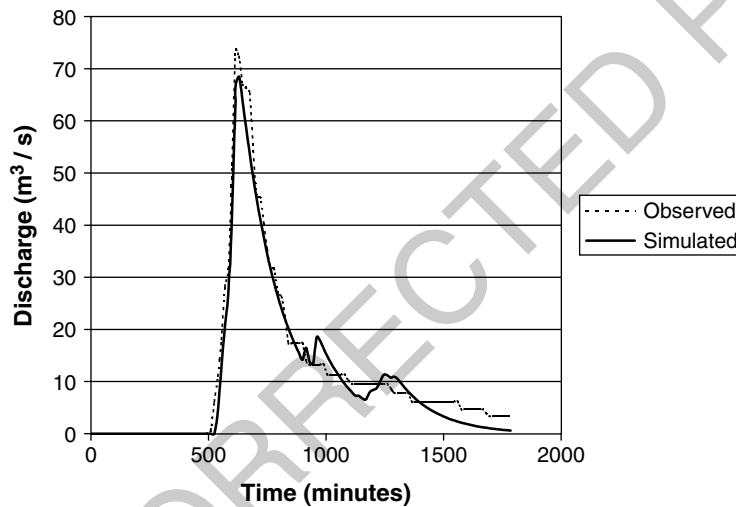


Figure 3. Simulation of a high-magnitude event (Carraixet, September 1990 event). Error in peak estimation: 6.8% in peak discharge and 15 min out of phase

Figure 2 shows an example of the improved simulations arising from the use of (SCS) effective rainfall compared with those using total rainfall. It is worth noting that, although the catchment characteristics may be of secondary importance for very high magnitude events, their influence was still significant in two key aspects: (i) the high volume of losses to the subsurface flow (Segura and Camarasa, 1996), and the tendency to form flash floods, with moving wave fronts, as a result of the rapid concentration of flows moving down a dry river bed. (Woolhiser, 1971). For these reasons, when the catchment response was immediate, either because it was a fast-response catchment, as in the case of the Rambla de Poyo, or because of high intensity rainfall (as in the case of the Carraixet September 1990 event (Figure 3), or the Poyo November 1988 event,

1 the TFMs produced excellent simulations. However, when low-energy events in slow-response catchments
 2 were considered (e.g. the Carraixet April 1991 event), model performance was poor. An intermediate example
 3 was given by the November 1988 Carraixet event, where, for an intermediate magnitude event, even when
 4 a delay time between the inputs and outputs of 30 min was used, the results were no better than mediocre
 5 (Figure 2)

8 ASSESSMENT OF FORECASTING PERFORMANCE

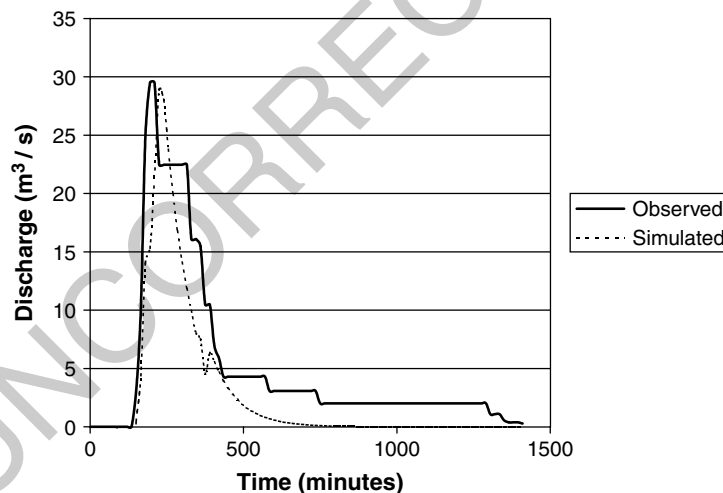
9 *Event—based forecast performance assessment*

11 The performance of the TFMs for both test catchments has been assessed using a small number of events
 12 (Table III). Owing to sparsity of historic data, the magnitude of the events varied greatly—however, this
 13 provided an opportunity to investigate the relationship between model forecasting performance and event
 14 magnitude. It should be noted that the test events were not used to calibrate the model and therefore provide
 15 an unbiased test of model forecasting performance.

16 The study indicates that the model performed well for both catchments when the TFM, calibrated using
 17 a single high-magnitude event, was used to forecast the hydrographs of a lower magnitude event. This is
 18 illustrated with two examples: the Rambla de Poyo model for a (low magnitude) test event (April 1991)
 19 and the Carraixet model for the (intermediate magnitude) November 1988 event. Figures 4 and 5 present the
 20 forecast performance for these two events.

21 Table III. Event-based forecast

23 Basin	24 Event used for calibrating model	25 Event used for simulation	26 Delta factor	27 Forecast root mean square errors (RMSE)
28 Barranc de Carraixet	29 November 1988	September 1990	0.98	2.397
	September 1990	November 1988	1.1	1.076
	September 1990	April 1991	1.11	0.909
30 Rambla de Poyo	31 November 1988	April 1991	0.45	2.196
	April 1991	November 1988	0.97	23.345



32 Figure 4. Example of temporal extrapolation of the model in a fast-response catchment (Poyo): simulation of a low magnitude event (April
 33 1991), using the TFM calibrated for a high-magnitude event (November 1988). Error in peak estimation: 2% in peak discharge and 30 min
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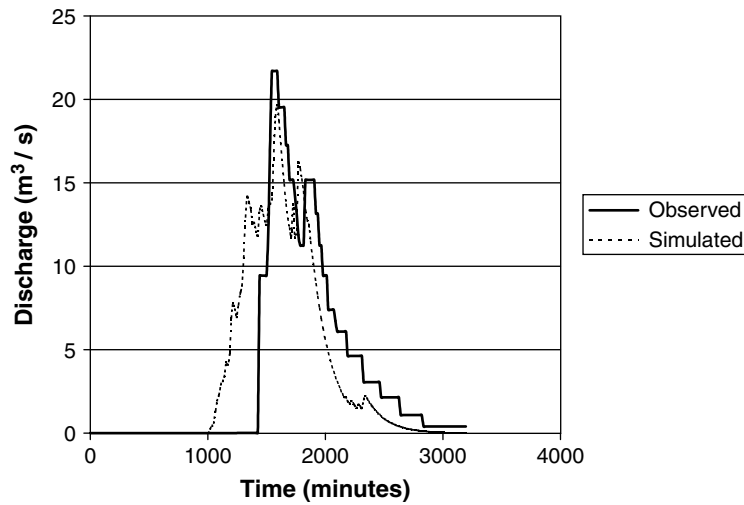


Figure 5. Example of temporal extrapolation of the model in a slow-response catchment (Carraixet): simulation of a medium magnitude event (November 1988), using the TFM calibrated for a high-magnitude event (September 1990). Error in peak estimation: 9.4% in peak discharge and 45 min out of phase

It is interesting to note that for both catchment models, forecasting performance is poor when the model is used to forecast flows for a high-magnitude event when the models have been calibrated using a lower magnitude events. This is illustrated by two examples: the Poyo catchment model for the (high magnitude) November 1988 event, and the Carraixet catchment model for the (high magnitude) September 1991 event. These are shown in Figures 6 and 7.

Spatial model transposition

The forecasting of flows in ungauged catchments represents a significant challenge. In order to investigate the potential of the TFM for forecasting flows in such circumstances, the forecasting performance of a TFM

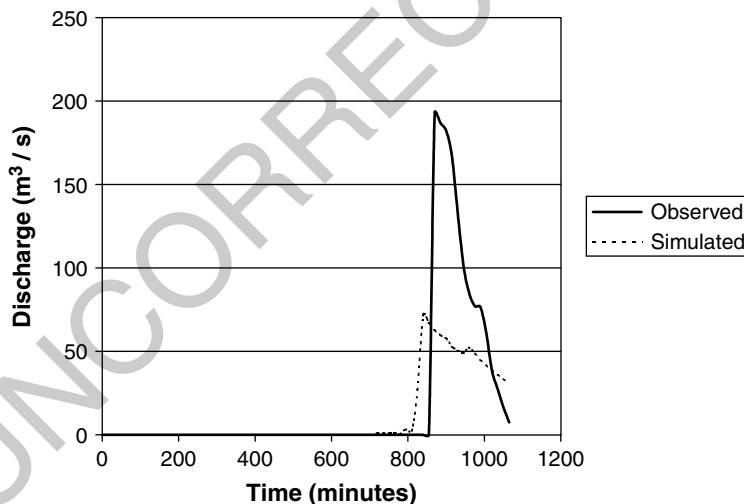


Figure 6. Example of temporal extrapolation of the model in a fast-response catchment (Poyo): simulation of a high-magnitude event (November 1988), using the TFM calibrated for a low-magnitude event (April 1991). Error in peak estimation: 62.7% in peak discharge and 30 min out of phase

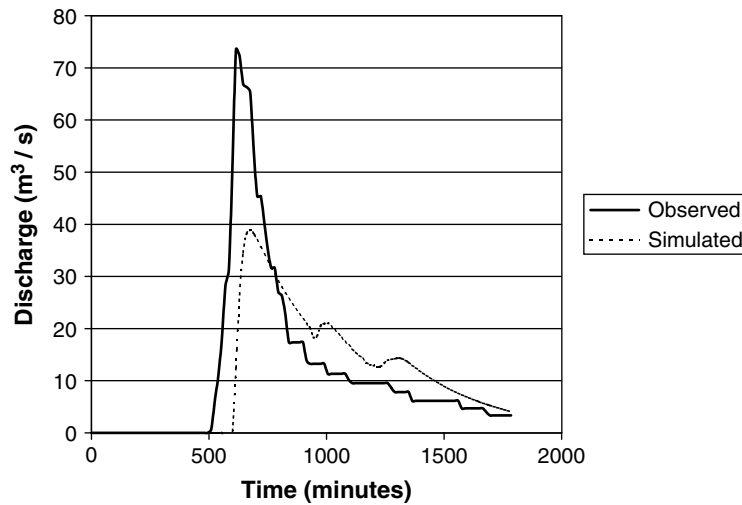


Figure 7. Example of temporal extrapolation of the model in a slow-response catchment (Carraixet): simulation of a high-magnitude event (September 1990), using the TFM calibrated for a medium magnitude event (November 1988). Error in peak estimation: 47% in peak discharge and 60 min out of phase

calibrated using data in a gauged ('donor') catchment was assessed when applied to a second ('receptor') catchment. For the purposes of this investigation, the receptor was also gauged—providing the opportunity to assess the quality of the model forecasts. The same two test catchments were used for this study (Poyo and Carraixet). Both catchments (donor and receptor) belong to the same environment, having the same geological structure and very similar geomorphological features. They are located just to the north and to the south of the city of Valencia, so they are very close and they are affected by the same storms. The only difference is the speed of the basin response to the rainfall, derived from their different permeability and slope. The model has been tested for different impulse–response functions, fitted in different events, as can be seen in Table IV.

The assessment reveals the rate of catchment response to storm rainfall as a key factor in the quality of model forecasting performance in the receptor catchment. The TFMs calibrated in the fast response Poyo catchment produced good quality forecasts when applied to the slower response Carraixet catchment, even when a low-magnitude event had been used to calibrate the model in the fast basin. This was highlighted by the simulation carried out on the Carraixet September 1990 event, using the TFM calibrated for the Poyo April 1991 event. Figure 8 illustrates how the model calibrated for the fast response catchment, even during a low-energy event such as that of April 1991, could be extrapolated to a slower response catchment, provided the simulated event had a certain magnitude. It is thought that this is because high-magnitude events in the receptor catchment reduce the importance of physical catchment processes in runoff generation, leading to good model performance.

Table IV. Spatial model transposition

Calibration		Simulation		Delta factor	Forecast root mean square errors (RMSE)
Donor basin	Event	Receptor basin	Event		
Barranco de Carraixet	September 1990	Rambla de Poyo	November 1988	1.1	18.547
Rambla de Poyo	November 1988	Barranc de Carraixet	September 1990	0.28	4.626
Rambla de Poyo	April 1991	Barranc de Carraixet	September 1990	0.7	4.368

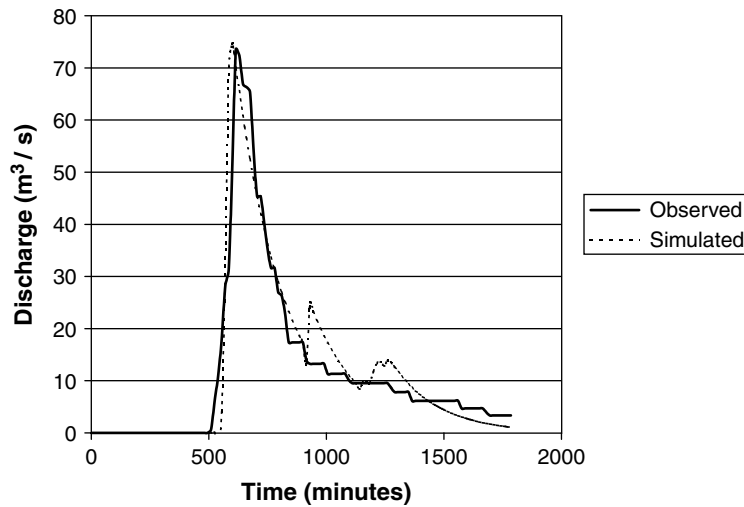


Figure 8. Example of areal extrapolation from a donor fast catchment to a receptor slow catchment: simulation of September 1990 in Carraxiet, using the TFM calibrated in Poyo, for the April 1991 event. Error in peak estimation: 1.56% in peak discharge and 15 min out of phase

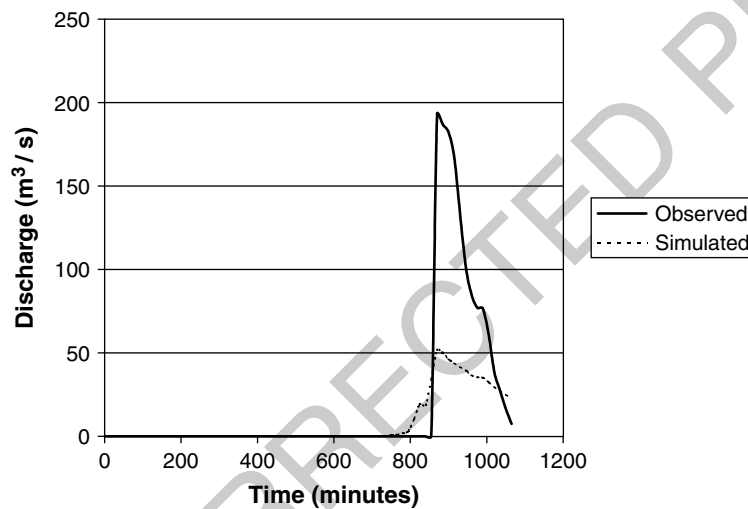


Figure 9. Example of areal extrapolation from a donor slow catchment to a receptor fast catchment: simulation of November 1988 in Poyo, using the TFM calibrated in Carraxiet, for the September 1990 event. Error in peak estimation: 51.71% in peak discharge and 0 min out of phase

Conversely, application of the TFM calibrated for the slower response Carraxiet catchment when applied to the faster response Poyo catchment performed poorly for all events. Figure 9 shows, as an example, the simulation of the November 1988 event in the Rambla de Poyo, using the TFM calibrated for the Carraxiet event of September 1990.

CONCLUSIONS AND DISCUSSION OF RESULTS

Although the results of this work are only preliminary, the various simulations have demonstrated the potential of the TFM for rainfall-runoff modelling in Mediterranean catchments. Other simulations were carried out in

1 these catchments for the same events (Camarasa, 1995), using the geomorphological unit hydrograph (GUH)
2 developed by Rosso (1984) and based on the Nash (1954) • unit hydrograph. This model uses morphometric
3 characteristics of the basin to estimate the scale and the form parameters of the classic Nash unit hydrograph. Q3
4 Results of simulations showed the important role of the scale parameter (in comparison to the form parameter),
5 related to rainfall intensity and structure, as well as the speed of quickflow. The form parameter, related to
6 basin morphology, had less influence on the prediction of the final hydrograph (Camarasa, 1995).

7 In these catchments, which lack baseflow and subsurface regulating inputs, the quickflow depends almost
8 exclusively on the rainfall. When the floods are of lower magnitude, the influence of the catchment on the
9 processes of runoff dominates, and the dependence on the rainfall is reduced. In such cases it was more
10 difficult to reproduce the hydrographs using simple linear TFMs. However, when high-magnitude events are
11 considered, the influence of the catchment is minimized, the dependence on the rainfall increases considerably
12 and model performance improves significantly. Simulation performance improves markedly when an input of
13 effective rainfall is used. The estimation of effective rainfall, which is of primary importance in any semi-arid
14 catchment, has been shown to be undertaken more effectively using an empirical model such as that of the
15 Soil Conservation Service, rather than a simple linear method scaling approach as used by the TFM.

16 The response rate of the catchment also has been shown to be a very important factor in the forecasting
17 performance of the TFM. Although both catchments have similar features and are located in the same
18 morphoclimatic environment, the rainfall–runoff conversion processes in the Poyo River are faster than in the
19 Carraixet River, owing to differences in permeability and slope. The lithology of the Poyo basin produces a
20 lower permeability compared with the Carraixet basin. Slope is greater in the Poyo basin than in the Carraixet
21 basin. Consequently, for similar inputs, the Poyo basin hydrograph shows shorter lag-time (3 h compared with
22 5 h in Carraixet) and time to peak. The shorter the runoff time, the less the influence of the catchment on the
23 hydrograph. That is why the Poyo hydrograph reproduces better the rainfall structure and, consequently, the
24 TFM is able to simulate this hydrograph better than for the Carraixet. Overall, model performance was best
25 for the faster responding Poyo catchment than in the slower response Carraixet. In the latter, the influence
26 of the physical catchment processes increases and, consequently, the results of the simulations were not as
27 convincing.

28 The TFM was most effective when calibrated on high-magnitude events or, in the case of model
29 transposition, from a fast response donor catchment. Once again, the study highlighted the importance of
30 the rainfall as a factor in model performance, observations that are consistent with the statement of Graf
31 (1988) that ‘processes dominate forms’, and the catchment’s response reproduces, in a purer manner, the
32 system’s inputs. The model could capture this type of response, which some authors, referring to the unit
33 hydrograph, have called ‘the catchment’s finger print’ (Marco and Reyes, 1981).

34 This is not the case, however, for low-magnitude events, or for slow response catchments, where the output
35 hydrograph reflects much more the influence of the catchment processes, the process ‘noise’ reducing the
36 models performance.

37 The results suggest that there is considerable potential for the use of a simple linear rainfall–runoff transfer-
38 function model for real-time flood forecasting in Mediterranean environments. The findings indicate that the
39 most important factor in producing satisfactory forecasting performance is to ensure that the models are
40 calibrated using high-magnitude flow events. If transposition of a calibrated model to an ungauged catchment
41 is necessary, the most important issue is to ensure that the donor catchment is a faster response system than
42 the receptor catchment. The greatest errors have been observed to occur when forecasting low-magnitude
43 events in slow response catchments: this is of limited consequence in a flood forecasting and warning context.
44

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49

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REFERENCES

- 3
4
5
6 Abdulrazzak MJ, Sorman AU, Alhames AS. 1989. Water balance approach under extreme arid conditions: a case study of Tabalah basin,
7 Saudi Arabia. *Hydrological Processes* **3**: 107–122.
- 8 Aucott LM, Grigg WL, Han D, Cluckie ID. 1992. Developing applications of weather radar in the Wessex flood forecasting system.
9 *Proceedings 2nd International Symposium on Hydrological Applications of Weather Radar*, University of Hanover, September.
- 10 Birks C, Bootman A, Cluckie ID, Han D. 1989. Wessex flood forecasting system. In *Hydrological Applications of Weather Radar*,
11 Cluckie ID, Collier CG (eds). Ellis-Horwood: Chichester. ● Q4
- 12 Bobinski E, Mierkiewicz M. 1986. Recent developments in simple adaptive flow forecasting models in Poland. *Hydrological Sciences*
13 *Journal* **31**: 297–320.
- 14 Box GEP, Jenkins GM. 1976. *Time Series Analysis: Forecasting and Control*. Prentice-Hall. ● Q5
- 15 Camarasa AM. 1995. *Génesis de crecidas en pequeñas cuencas semiáridas: Barranc del Carraixet y Rambla de Poyo*. PhD thesis, MOPT-
16 Confederación Hidrográfica del Júcar, Valencia, 252 pp.
- 17 Camarasa AM, García Bartual R. 1991. Estimación del hidrograma de crecida a partir de un modelo conceptual de base geomorfológica.
18 *Tecnología del Agua* 49–55. ● Q6
- 19 Chander S, Shanker H. 1984. Unit hydrograph based forecast model. *Hydrological Sciences Journal* **31**: 279–291.
- 20 Cluckie ID. 1993. Real-time flood forecasting using weather radar. In *Concise Encyclopedia of Environmental Systems*, Young PC (ed.).
21 Pergamon Press: Oxford.
- 22 Cluckie ID, Harpin R. 1982. A real-time simulator of the rainfall–runoff process. *Mathematics and Computers in Simulation* **XXIV**: 131–139.
- 23 Cluckie ID, Owens M. 1987. Real-time rainfall–runoff models and use of weather radar information. In *Weather Radar and Flood*
24 *Forecasting*, Collinge VK, Kirby C (eds). Wiley: Chichester; 171–190.
- 25 Cluckie ID, Harwood DA, Harpin R. 1980. Three systems approaches to real-time rainfall–runoff forecasting. In *Hydrological Forecasting*.
26 *Proceedings of the Oxford Symposium*, April, IAHS Publication 129, International Association of Hydrological Sciences: Wallingford;
27 389–396.
- 28 Cluckie ID, Yu PS, Tilford KA. 1989. Real-time flood forecasting, model structure and data resolution. In *Weather Radar Networking*,
29 Collier CG, Chapius M (eds). Proceedings, COST-73 Seminar on Weather Radar Networking, Brussels, Belgium, September. Kluwer:
30 Dordrecht; 459–472.
- 31 Corradini C, Melone F. 1987. On the structure of a semi-distributed adaptive model for flood forecasting. *Hydrological Sciences Journal*
32 **32**(2): 227–242.
- 33 Corradini C, Melone F, Ubertini L, Dingh UP. 1986. Geomorphologic approach to synthesis of direct runoff hydrograph from the upper
34 Tiber basin. In *Scale Problems in Hydrology*. Kluwer: Dordrecht; ● 57–79. Q7
- 35 García Bartual R, Marco J. 1990. A stochastic model of the internal structure of convective precipitation in time at a raingauge site. *Journal*
36 *of Hydrology* **118**: 129–142.
- 37 Graf WL. 1988. *Fluvial Processes in Dryland Rivers*. Springer-Verlag: Berlin.
- 38 Gupta VK, Waymire E. 1983. On the formulation of an analytical approach to hydrologic response and similarity at the basin scale. *Journal*
39 *of Hydrology* **65**: 95–123.
- 40 Harpin R. 1982. *Real time flood routing with particular emphasis on linear methods and recursive estimation techniques*. PhD thesis,
41 University of Birmingham, Department of Civil Engineering.
- 42 Isermann R. 1981. *Digital Control Systems*, Springer-Verlag: Berlin; 566 pp.
- 43 Jakeman AJ, Hornberger GM. 1993. How much complexity is warranted in a rainfall–runoff model? *Water Resources Research* **29**(8):
44 2637–2649.
- 45 Klemes V. 1981. Stochastic models of rainfall–runoff relationships. Pre-symposium proceedings, *International Symposium of Rainfall–Runoff*
46 *Modelling*; ● 36–36. Q8
- 47 Lees MJ. 1997. Modelling and automatic control of flow regulation for multipurpose catchment management. *Proceedings, 5th British*
48 *Hydrological Society National Hydrology Symposium*, Salford; 1.1–1.12. ● Q9
- 49 Lees MJ, Young PC, Ferguson S, Beven K, Burns J. 1994. An adaptive flood warning scheme for the River Nith at Dumfries. In *2nd*
50 *International Conference on River Flood Hydraulics*, White WR, Watts J (eds). Hydraulics Research; Wallingford. Wiley: Chichester. ● Q10
- 51 Maidment DR. 1993. Developing a spatially distributed unit hydrograph by using GIS. In *Application of Geographic Information Systems*
52 *in Hydrology and Water Management*. ● IAHS 211, International Association of Hydrological Sciences: Wallingford; 181–192. Q11
- 53 Marco J. 1989. La defensa integral frente a las crecidas en la Comunidad Valenciana. In *El agua en la Comunidad Valenciana*. Universidad
54 Politécnica de Valencia; 61–69.
- 55 Marco J, Reyes M. 1981. *Hidrología*. Escuela técnica Superior de ingenieros de caminos, canales y puertos de Valencia.
- 56 Mateu JF. 1988. Crecidas e inundaciones en el País Valenciano. In *Guía de la Naturaleza de la Comunidad Valenciana*. Edicions Alfons el
57 Magnànim: Diputacion Provincial de Valencia; 595–636.
- 58 Moore RJ. 1980. *Real-time Forecasting of Flood Events using Transfer Function Noise Models. Part 2*. Contract Report to the Water Research
59 Centre, Medmenham, UK. Institute of Hydrology, Wallingford.
- 60 Moore RJ, O'Connell PE. 1978. *Real-time Forecasting of Flood Events using Transfer Function Noise Model. Part 1*. Contract Report to the
61 Water Research Centre, Medmenham, UK. Institute of Hydrology, Wallingford.
- 62 Morisawa M. 1985. *Rivers: Form and Process*. Longman: London.
- 63 Nalbantis I. 1995. Use of multiple time-step information in rainfall–runoff modelling. *Journal of Hydrology* **165**: 135–159.

- 1 Nash JE. 1958. Determining runoff from rainfall. *Institute of Civil Engineering Proceedings* **10**: 163–184.
- 2 Nouh M. 1990. Flood hydrograph estimation from arid catchment morphology. *Hydrological Processes* **4**: 103–120.
- 3 O'Connell PE, Clark RT. 1981. Adaptive hydrological forecasting—a review. *Hydrological Sciences Bulletin* **26**(2): 179 pp.
- 4 Owens MD. 1986. *Real-time flood forecasting using weather radar data*. PhD thesis, University of Birmingham, Department of Civil Engineering.
- 5 Powell SM. 1985. *River basin models for operational forecasting of flow in real-time*. PhD thesis, University of Birmingham, Department of Civil Engineering.
- 6 Powell SM, Cluckie ID. 1984. On the sampling interval of discrete transfer function models of the rainfall runoff process. *Proceedings 7th IFAC/IFOR Symposium on Identification and System Parameter Estimation*.
- 7 Ramos J, Mallants D, Feyan J. 1995. State-space identification of linear deterministic rainfall–runoff models. *Water Resources Research* **31**(6): 1519–1531.
- 8 Reed DW. 1984. *A Review of British Flood Forecasting Practice*. Report No. 90, Institute of Hydrology 42 pp. • Q12
- 9 Rodriguez-Iturbe I, Valdés JB. 1979. The geomorphological structure of hydrologic response. *Water Resources Research* **15**(6): 1409–1420.
- 10 Rodriguez-Iturbe I, Gonzalez Sanabria M, Bras RL. 1982. The geomorphoclimatic theory of the instantaneous unit hydrograph. *Water Resource Research* **18**(4): 877–886.
- 11 Rosso R, Caroni E. 1987. Analysis estimation and prediction of the hydrologic response from catchment geomorphology. *Memorie e studi dell'istituto di idraulica e costruzioni idrauliche del politecnico di milano* **1**: 93–108.
- 12 Rosso R. 1984. Nash model relation to Horton order ratios. *Water Resources Research* 914–920. • Q6
- 13 Segura FS. 1990. *Las ramblas valencianas*. PhD thesis, Universitat de València, 229 pp.
- 14 Segura FS, Camarasa A. 1996. Balances hídricos de crecidas en ramblas mediterráneas: pérdidas hídricas. In *Clima y agua: la gestión de un recurso climático*, Marzol MV, Dorta P, Valladares P (eds). Universidad de La Laguna: Tenerife; 235–245.
- 15 Sherman LK. 1932. Streamflow from rainfall by unitgraph method. *Engineering New Records* **103**: 501–505.
- 16 Témez J. 1978. *Cálculo hidrometeorológico de caudales máximos en pequeñas cuencas naturales*. MOPU Dirección General de Carreteras, 113 pp. • Q14
- 17 Tilford KA. 1993. *Weather radar data for operational hydrology*. PhD thesis, Telford Institute of Environmental Systems, Department of Civil and Environmental Engineering, University of Salford, Manchester, 330 pp.
- 18 US Soil Conservation Service. 1972. *National Engineering Handbook*, Section 4, Supplement A, *Hydrology*. US SCS: Washington, DC.
- 19 Wilke K, Barth F. 1991. Operational river-flood forecasting by Wiener and Kalman filtering. In *Hydrology for the Water Management of Large River Basin. Vienna Symposium*. IAHS Publication 201, International Association of Hydrological Sciences: Wallingford. • Q15
- 20 Woolhiser DA. 1971. Deterministic approach to watershed modelling. *Nordic Hydrology II* •146–166.
- 21 Young PC, Beven KJ. 1994. Data-based mechanistic modelling and the rainfall-flow non-linearity. *Environmetrics* **5**: 335–363. Q6
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- 23
- 24
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