INSTITUTE OF HYDROLOGY PREDICTION OF RUNOFF VOLUME FROM FULLY-SEWERED URBAN CATCHMENTS HUG TEXTRASE

by

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# ABSTRACT

This report describes the development of a mathematical model to predict the volume of runoff from a given rainfall event on a fullysewered catchment. The model will ultimately be incorporated into a new design method for storm sewer systems. Regression analysis on existing urban catchment data was used in the model development because the translation of rainfall into runoff is a complex physical process. Data from 368 storms on 14 catchments were collated and processed from an archive being established at the Institute. A data set was compiled of rainfall and runoff volumes and pertinent catchment and storm characteristics for each of the 368 events. Analyses performed on the data from individual catchments demonstrated that there was no clear trend from one catchment to another. Analyses performed on the whole data set identified an additive 3-variable form of regression equation, using percentage runoff as dependent variable, as being the most appropriate. This equation will be updated as further data become available.



REPORT NO 41

September 1977



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### INTRODUCTION

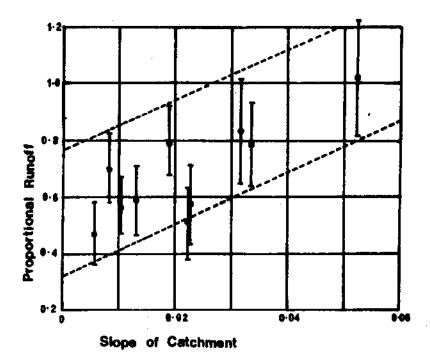
Complementary research programmes at the Hydraulics Research Station and the Institute of Hydrology are concerned with the development of rainfall-runoff simulation models for the improved design of urban storm water systems. The Institute's contribution relates to the above-ground phase of the rainfall-runoff process, the conversion of the rainfall hyetograph into a runoff hydrograph at the inlet to a sewer system (Helliwell  $et\ al.$ , 1976).

The hydrological investigation may be subdivided into three parts: (a) the derivation of the volume of runoff (or the losses relating to any given total rainfall volume), (b) the distribution of these losses in time, and (c) the distribution of the runoff volume in time or the attenuation of the effective rainfall profile. Progress in parts (b) and (c) relies on subcatchment experiments in progress at Bracknell, Stevenage, Southampton and Wallingford, using a flow meter developed to measure the discharge through a road gully (Blyth and Kidd, 1977) at the interface between the above and below ground phases of the urban runoff process. However, this report is concerned with part (a) only. And, whilst valuable insights into the volume of runoff may be gained from these experiments, catchment-averaged results may be obtained from existing urban rainfall-runoff data from larger catchments up to 200 ha. Indeed, as indicated in the next section, the random spatial distribution of different surface types and predominance of unquantifiable losses leads to the conclusion that a catchment-averaged approach to the derivation of runoff volume is probably more appropriate than an approach using subcatchment data.

The traditional approach to estimating the volume of runoff stems from the runoff coefficient in the Rational (Lloyd-Davies) Formula. Escritt (1950) has amalgamated the work of a number of researchers who have tabulated runoff coefficients for various types of contributing area and, in some cases, have related overall coefficients to some measure of population density. These coefficients were determined by intuition and empiricism and require subjective judgement to apply in a given situation.

An alternative approach has been to take the runoff coefficient as equal to the proportion of impervious surfaces in the catchment. The Road Research Laboratory (Watkins, 1962) perpetuated this practice, despite the fact that the data they collected suggests lower runoff volumes than the assumption implies. An alternative way of looking at this concept is to assume that there is 100% runoff from impervious surfaces and 0% from pervious surfaces. Both assumptions are doubtful but the errors have a tendency to cancel each other out.

Sarginson (1973) used the summary of the RRL data (Watkins, 1962) to observe the variation of average observed percentage runoff from catchment to catchment. He noted a significant correlation between percentage runoff and average catchment slope, as indicated in Figure 1.



Note: error bars represent one standard deviation of RRL data.

FIGURE 1 Variation of proportional runoff with average catchment slope (after Sarginson)

A number of urban runoff models such as the Chicago Hydrograph Model (Tholin & Keifer, 1960), the University of Cincinnatti Urban Runoff Model (Papadakis & Preul, 1972) and the Storm Water Management Model (Metcalf & Eddy Inc., 1971), attempt to model the volume of runoff deterministically. In general, they assume 100% runoff from impervious surfaces after deducting an allowance for depression storage, and they employ a Horton (1940) type of infiltration model to allow for the losses from the pervious surfaces.

Kaltenbach (1963) related percentage runoff to percentage impervious and maximum five-minute rainfall intensity based on data collected by the Johns Hopkins University in the Eastern United States.

The first part of the work described in this report is concerned with a purely qualitative appraisal of the physical phenomena which control the losses in a sewered catchment. This section concludes with: (i) a list of variables upon which the volume of runoff might be expected to depend, and (ii) a justification for the statistical approach adopted in this work. This is followed in Section II by a description of the data set compiled for the specific purposes of this study and in Section III, the development of a regression model for the prediction of the runoff volume from a given rainfall input. Section IV contains general conclusions drawn from the study.

# I: A QUALITATIVE APPRAISAL OF THE LOSSES IN SEWERED CATCHMENTS

The volume of runoff may be defined as being equal to the volume of rainfall minus losses. The losses in urban catchments are caused by the same hydrological processes as in natural catchments, albeit in different proportions. Each of these processes will be discussed so as to identify those catchment and storm variables upon which the total loss might depend. With respect to the volume of runoff, the most significant physical phenomenon in the transition of a catchment from a rural to an urbanised state is the increase in the proportion of impervious surfaces. This increase has the effect of decreasing infiltration, resulting in a reduction of total loss and a corresponding increase in the total volume of runoff. In discussing each of the hydrological loss processes, the catchment area will be divided into pervious and impervious surfaces, although it is realised that there could be subgroups within this division. It is considered that the only significant subgrouping is between roofed and paved areas within the impervious category.

# Interception and evaporation

Both interception and evaporation are negligible in fully-sewered urban catchments and therefore unimportant in the determination of runoff volumes. The only effect of evaporation is on the depletion of depression storage between rainfall events and as such, it will affect the immediate antecedent conditions.

### Infiltration

The infiltration process in the pervious areas of an urban catchment may be assumed to be identical to that taking place in natural catchments. Assuming a Horton (1940) type of infiltration model (ie. loss rate subtracted from rainfall rate), the infiltration loss from pervious surfaces will be a complex function of the pervious area, the rainfall duration, total depth of rainfall, some measure of rainfall intensity over all or some critical part of the rainfall event, and the Horton coefficients: the initial infiltration capacity ( $f_0$ ); the equilibrium infiltration capacity ( $f_0$ ) and the decay constant ( $f_0$ ). These in turn are dependent on the soil type and the long-term antecedent wetness of the catchment.

The term 'impervious' as traditionally used is undoubtedly something of a misnomer since it is likely that, of the rain falling on a paved area, some small proportion infiltrates, either through the surface itself or through cracks, joints or faults in the paved surface. The nature and temporal distribution of this infiltration is difficult to determine but may be assumed to be constant. If this assumption is adopted, then the loss due to infiltration over the paved surfaces is a function of the paved area and the storm duration. On the other hand, assuming that the roofed areas are directly connected to the sewer system, any infiltration loss from this source is negligible.

### Depression storage

Storage of rainwater on the surface of the catchment, preventing its contribution to runoff, is known as depression storage. In the case of paved surfaces it may be seen as a combination of (i) that layer of water which is being held on the ground by surface tension (initial wetting) and (ii) that water which is being held in surface depressions where the configuration of the ground prevents the water from reaching an inlet. These two components are best combined and depression storage treated as a depth of water spread evenly over the surface. Depression storage is a function only of physical variables, the predominant ones being catchment slope and paved area. Depression storage for pitched roofs may be seen as initial wetting only, and is relatively insignificant. The likely ratio of magnitude of the two components of depression storage would suggest that storage loss for roofed areas is negligible. Depression storage on pervious surfaces will similarly depend on catchment slope and pervious area. However, a complication arises because the depression storage will be depleted by infiltration when the rainfall intensity falls below the infiltration rate. Thus, the loss due to depression storage over pervious surfaces is, to some extent, also a function of those variables affecting the infiltration losses over the pervious areas.

The depression storage on both pervious and impervious surfaces also depends on the immediate antecedent conditions (not to be confused with the longer-term antecedent conditions). If the event is a single cell of a much larger event, then one might expect the depression storage over the paved surfaces, if not the pervious surfaces, to be satisfied. Conversely, an event occurring over a dry catchment might have to satisfy the full depression storage before runoff occurs. In this study the immediate antecedent conditions are treated as a dichotomy, such that depression storage is taken to be either satisfied or unsatisfied corresponding to a wet or dry catchment. This is clearly a simplification, since it is possible to have an event where, immediately prior to the start, the catchment conditions are somewhere between the two extremes. However, the vast majority of events used in this study appeared to fall clearly into one of the two categories outlined above.

### Other losses

Perhaps the greatest cause of uncertainty in estimating runoff volumes lies in the interaction of runoff from impervious and pervious surfaces. In most cases, the runoff contribution from pervious areas will travel over the impervious surfaces to reach the sewer system. If this contribution is considered in the same way as rain falling on the impervious surfaces, then it will be subject to the same losses as the latter. In some cases, the above order of occurrence may be partially reversed, as in the case of a footpath separated from road gullies by a grass verge. In this instance, the footpath's contribution is bound to be influenced by the physical characteristics of the grass verge and vice versa. It is impractical to postulate a distributed model for such a complex interaction for use in a design situation.

The other major uncertainty of this type corresponds to the assessment of contributing area. In any given catchment there will almost certainly be a proportion of the pervious surfaces which may not be expected to contribute to the sewer system (in effect, all rainfall will be lost to infiltration). In addition, there will be areas of impervious surface which do not contribute to the sewer system, either because the impervious surfaces were not constructed strictly according to the designer's specifications or due to subsequent effects such as blockage and by-passing of gullies, blockage of downpipes etc. Such losses as these are also not amenable to distributed modelling.

Another cause of rainfall loss is due to a reduced catch by roofs. It is current opinion that a roof will catch a reduced quantity of rainfall due mainly to air turbulence around a building. The reduced catch by a roof will result, to some extent, in an increased catch by other surfaces, although it is unlikely that the magnitude of the increased catch will have as significant an effect on the volume of runoff as the reduced roof catch, because much of it will fall on the walls of buildings and on the back-gardens, from where it will be unlikely to reach the sewer system.

The last loss to be considered does not occur on the ground surface at all but in the sewer system itself where infiltration (in through the pipe joints) or exfiltration (out through the pipe joints) may occur. There is usually a base flow present in almost all storm sewer systems throughout dry periods. This may be due to infiltration, an effective 'negative' loss to the system. Such a baseflow is indicative of some interaction between the system and the surrounding groundwater. To compound the problem, there is no indication as to whether this infiltration is present at a constant rate through all depths of flow or whether it may change into exfiltration at a flow depth greater than the baseflow resulting in an effective loss to the system.

When considered individually, each of the above factors might be regarded as relatively trivial in the context of a design model. Considered together, they represent a major source of error which hinders model calibration on most catchments and justifies the use of simplified concepts.

There are two major conclusions arising out of the above discussion. The first comprises a list of catchment and storm characteristics upon which the losses in fully-sewered urban catchments may be expected to depend. Table 1 is a summary of these characteristics. Some, such as rainfall depth, duration and catchment area, may be reflected directly as catchment or storm variables; others, such as antecedent catchment conditions or soil type, require indexing.

The other major conclusion concerns the complexity of processes in sewered catchments. If a deterministic solution were to be attempted, then all significant aspects of the process should be considered. While some aspects, such as infiltration and depression storage, are amenable to such an approach, others, such as the interaction of surfaces and pipe blockages, are not. It seems that

a statistical approach to runoff volume prediction, such as has been used recently for natural catchments (NERC, 1975), is more appropriate.

TABLE 1 Factors affecting rainfall-runoff losses

	IMPERVIOUS AREAS	PERVIOUS AREA	S	
Cause of loss	Catohment characteristics	Storm characteristics	Catalmens shirasteristics	Storm characteristics
INFILTRATION	Tubetaronz gres	storm duration	pervicus area soil type	storm duration rainfall volume long-term catchment wetness rainfall intensity
DEPRESSION STORAGE	impervious area	imediate antecedent wetness	pervious area catchment slope	immediete antecedent wetness infiltration factors
OTHER LOSSES	umpervious area	storm duration rainfall volume		storm duration

### II: THE DATA SET

Data from fourteen catchments have been collected and assembled in a suitable form for analysis. Data from seven of the catchments had been used previously by the Transport and Road Research Laboratory in their work on the design of urban sewer systems (Watkins, 1962). Data from the other seven catchments have been collected more recently during university research or under the sponsorship of the Department of the Environment.

Table 2 gives brief details of all fourteen catchments which are fully-sewered and gauged for rainfall and runoff at one or more points. The location of the catchments is shown in Figure 2.

# Rainfall-runoff data preparation

In general, rainfall was measured by one or more autographic raingauges and both rainfall and runoff were recorded on open-scale charts. The objective, in all cases, was to obtain temporal distributions of rainfall and runoff for high intensity storms of up to three hours duration. Data from a certain number of catchments were rejected due to poor quality; only data from catchments with reliable discharge measurements were used.

For all catchments, raw data were available in the form of rainfall and runoff charts. Individual storm events were extracted for analysis with regard to two main criteria:

- (i) Discrete (but not necessarily single-peaked) hydrographs resulting from relatively short and intense rainfall
- (ii) Runoff in excess of a specified stage for each flow gauge, ie. 'significant' events.

The chosen events were reduced to a digital record using the Institute's D-MAC digitiser. These records were then transferred to a disc-file record on the Institute's UNIVAC computer as a time series of rainfall (intensity in mm/hour) and runoff (discharge in litres/second) values at one minute intervals throughout each event.

The data for each event were then synchronised (according to chart time) as a rainfall-runoff record and output in both tabular and graphical form. The above data collection exercise was not done solely with the runoff volume studies in mind but with the purpose of creating an archive of urban hydrological data.

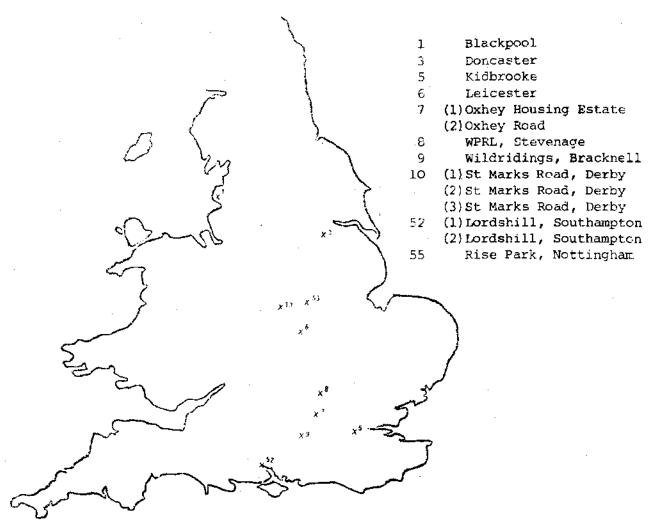


FIGURE 2 Location of catchments used in analysis

TABLE 2 Details of catchments employed in analysis

	CATCHMENT NUMBER AND NAME	BRISEF DESCRIPTION OF CATCHMENT	RAINFALL	RUNOFF	SPONSOR	TOTAL AREA	treery (ha)	PERIOD OF DATA	NO. OF EVENTS
1	BLACKPOOL	medium density residential development-circa 1939	RRL auto- graphic gauge	standing wave flume in combined sever	現職し	4.42	#1.9	1953-1958	46
3	DONCASTER	low density resi- dential develop- ment - probably pre-war	RRL auto- graphic gauge	standing wave flume in seapar- ate sewer	RRL	5.14	30.0	1955-1958	15
5	K108ROOKE, KENT	small factory area with large concrete yards	RRL auto- graphic gauge Pines auto- graphic gauge	wave flume in separa-	RRL	3.42	68.2	1953-1958	70
6	LEICESTER	mixed alder-type residential devel- opment. High density close to outfall - low density at top of catchment	ZRRL auto- graphic gauges		RRL	59.50	36.1	1958	39
7(1)	STATES DALEUCH YEMAD	large residential development	3 RRL auto- graphic gauges		ARL	247,00	19.8	1953-1959	19
7(2)	DEMEN PIND	500 m length of moad and grass verge in shallow cutting	RRL auto- graphic gauge Dines auto- graphic gauge	Y-motch weir	RAL	. 73	60.3	1954-1959	6
8	WPRL, STEVENAGE	small factory area similar to Kidbrook		Y-notch weir	RRL	1,39	59.0	1955-1959	26
9	WELDPICINGS, BRACKMELL	Modern development residential, shops and school		Standing wave flume at outfall	ቦርነዊ	11.60	46,2	1974-1975	20
12(1)	ST. MARKS RD 1, DERBY	3 nested catchments		stage is	DGWE	10.40	52.9		151
9(2)	ST. MARKS TO 2, DERBY	mixed-age residenti development	Dines auto-	neasured at 1 noints in		8.55	\$1.0	1971+1975 1971-1975	19
0(3)	ST. MARKS PO 3, DERBY		graphic gauge	the system	DGWE	7.23	48.7		39
<b>52</b> (1)	LORDSHILL 1, SOUTHAMPTON	2 small adjacent	Tipping	Arkan air-		.8	41.3		29
52 ° 2 °	LORDSHILL 2, SOUTHAMPTON	ratchments, modern residential devel- opment	hucket gauge	purge system in prefabri- cated flumes at outfall	of S'ton	.6	41,7	1974-1975	31
55	RISE PARY NOTTINGHAM	modern residential	Tipping bucket gauge		Trent Poly,	62.0	31,1	1974	7

Some of the rainfall-runoff data were poorly synchronised, but this deficiency (although important for modelling the complete process) is irrelevant for modelling volume relationships as long as the causative rainfall can be identified. At the time of digitisation the start and finish points of a rainfall event were chosen to cover adequately the complete storm; later, a reduction of the event was made so that only the period of most intense rain was included. The start of this intense rainfall was arbitrarily defined as the point when the rainfall intensity continuously exceeded 1 mm/hr and the end of the event was defined as the point when the intensity dropped below 1 mm/hr for more than five minutes. The resulting hyetograph was integrated to give a total rainfall volume in mm over the catchment. If there were records from more than one raingauge in or near the catchment, an average rainfall hyetograph was generated using Thiessen polygons.

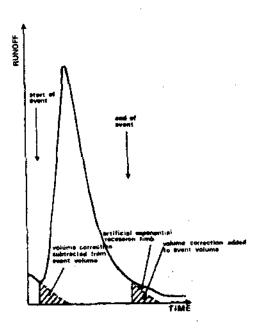


FIGURE 3

Typical corrections made in the derivation of runoff volume

Similarly, the runoff data were reduced to give a single hydrograph for each event. The start of the event was defined as the point where the hydrograph began to rise and the end at a point on the falling limb before it had flattened out (hydrographs from sewered catchments tend to flatten out at a low discharge into long recession limbs not associated with the direct catchment runoff). A typical hydrograph is shown in Figure 3.

A correction was made to account for the volume of direct runoff still in storage at the nominal end of the event, assuming an exponential recession limb from the cutoff point. For a given catchment, the value of the recession constant K (minutes) was obtained from a study of the approximate time of concentration and logarithmic plots of the recession limbs of selected events. For an exponential decay, the volume under a recession associated with a given discharge Q (litres/s) is equal to 60KQ (litres). A similar correction was deducted from the runoff volume where the rising limb of the hydrograph commenced from a non-zero discharge. These volume corrections are illustrated in Figure 3.

The volume of runoff in litres for each event was thus calculated by integration of the hydrograph from start to end, and the above corrections applied. The volume was then converted to an equivalent depth in mm over the paved area so that it was dimensionally comparable with rainfall depth.

Together with the two variables, rainfall volume (PPT) and runoff volume over the paved surface (IMPQ), the basic data set comprises another twelve variables. Of these, six may be defined as catchment characteristics and eight as storm characteristics. The definition of these characteristics together with a number of subsidiary ones (calculated from the basic set), will be treated individually.

# Catchment characteristics

The first catchment variable in the data set is the catchment number (CNO), and is used solely for identification.

The second characteristic is the average catchment slope. A number of slope measurements were considered. Finally, the variable SLOPE was defined as the weighted average pipe slope, which was felt to provide a good objective approximation to an average ground slope and was capable of exact replication. Details of lengths and gradients were extracted from a plan of the pipe network and used in the following equation to derive SLOPE:

SLOPE = 
$$\frac{\sum_{j=1}^{n} \sum_{j=1}^{L_{j}} \sum_{j=1}^{L_{j}} \sum_{j=1}^{L_{j}} \dots (1)}{\sum_{j=1}^{L_{j}} \sum_{j=1}^{L_{j}} \sum_{j=1}^{L_{j}} \dots (1)}$$

where S and L are the slope (%) and length respectively of each individual pipe j in the system, and n is the total number of pipes in the system.

The area characteristics are defined by three variables, TOTA, IMPA, and RATIO. TOTA and IMPA are the total area and impervious area respectively, and RATIO is the ratio of roofed to paved surfaces within the impervious category. A subsidiary variable, defined from the basic set, is the percentage of impervious area, PIMP, obtained from the equation

$$PIMP = \frac{100 \times IMPA}{TOTA} \qquad ... (2)$$

The soil type is described using the soil index, SOIL, as employed in the UK Flood Studies Report (NERC, 1975). The soil index is calculated by finding the area of the catchment in each soil type, using the national soil map at a scale of 1 inch to 10 miles (NERC, 1975), which defines five soil classifications from class 1 (very high acceptance, low runoff) to class 5 (very low acceptance, high runoff). A weighted mean of soil fractions was adopted as a soil index with a range of .15 to .50, as given by

$$SOIL = \frac{0.15S_1 + 0.30S_2 + 0.40S_3 + 0.45S_4 + 0.50S_5}{S_1 + S_2 + S_3 + S_4 + S_5}$$
 ... (3)

where S is the catchment area within soil classification n. In practice, each catchment, because of its small relative size, fell wholly within one soil type, and so the calculation was simplified.

### Storm characteristics

The first storm variable is the storm number (SNO), and is used for identification only. The next two are the rainfall volume (PPT) and

the runoff volume over the paved area (IMPQ), the derivation of which is described above. A subsidiary variable is the runoff volume expressed as a depth over the total catchment area (TOTQ), and is calculated from

$$TOTQ = IMPA \times IMPQ/TOTA$$
 ... (4)

The immediate antecedent conditions, relating to whether or not the depression storage is satisfied, are indexed by the variable IFWET, which has a value of 1 when the antecedent conditions are dry (depression storage unsatisfied) and of 0 when the reverse is true. It is obtained by a subjective scrutiny of the rainfall charts for the period leading up to the commencement of the event.

The long-term antecedent conditions are indexed by one of three variables. The first is the five-day antecedent precipitation index, API5, and is obtained from the rainfall record prior to a given event. Assuming the daily rainfall to be evenly distributed, the API5<sub>9</sub> (the index at 9 a.m. on any given day) was obtained as follows:

$$API5_9 = API5_O \times C + P \times \sqrt{C} \qquad ... (5)$$

where API5 $_{\odot}$  is the value of the index at 9 a.m. on the previous day, P is the rainfall in the preceding 24 hours, and C is the depletion constant taken as 0.5. The value of the API5 at the start of an event was then obtained from the following:

$$API5_{T} = API5_{9} \times C^{\frac{T-9}{24}} + P_{9,T} \times C^{\frac{T-9}{48}}$$
 ... (6)

where T is the start time of the event. T-9 is the time (in hours) between 9 a.m. on the day of the event and the start of the event.  $P_{0,T}$  is the rainfall during T-9 taken to be evenly distributed. If the distribution was known and the rainfall was very unevenly distributed, the adjustment to obtain API5, was done manually (using a depletion constant of  $C^{1/24}$  for every hour).

SMD, the soil moisture deficit, is a further index of antecedent conditions. The SMD value for 9 a.m. on the day of the event was obtained for the nearest SMD station(s) to the catchment. (Data were obtained through the Meteorological Office, Bracknell). If data from more than one station were used, an average was obtained by weighting the individual values on the basis of proximity to the catchment. This value was then reduced by the quantity of rain which fell between 9 a.m. and the start of the event (P<sub>9.T</sub> above).

CWI, the catchment wetness index, is a subsidiary variable describing

antecedent conditions. It was developed in the Flood Studies (NERC, 1975) to provide a compromise between SMD and the rather shorter term index, API5. It is obtained from API5 and SMD using the following equation:

$$CWI = 125 + API5 - SMD$$
 (All values in mm) ... (7)

The figure of 125 is introduced with the sole purpose of keeping the CWI positive.

The next variable is the rainfall duration, DUR, defined as the time in minutes between start and end times of intense rainfall as discussed in the section on rainfall-runoff data preparation.

The rainfall intensity is characterised in one of two ways. Kaltenbach (1963) considered that the most significant variable was the average rainfall intensity during the severest 5 minutes of the event. Thus, M5I, the maximum 5-minute rainfall intensity, was extracted from the rainfall record. A subsidiary variable, AVINT, the average rainfall intensity during the event was included as an alternative to the M5I, and is given by:

$$AVINT = 60 \times PPT/DUR \qquad ... (8)$$

### Dependent variables

Having chosen multiple linear regression analysis to generate a mathematical model, the number of different forms of the model are limited to those in which the parameters occur linearly. Firstly, an additive form (y = B + B + B x + B x + D x +

For instance, the runoff-volume (in litres) might be the initial choice for the dependent variable. But intuition would suggest that this volume is so strongly (and directly) dependent on the area of the catchment that a better dependent variable could be obtained by dividing the volume by the area - this is because the runoff volume from a given catchment might be expected to be approximately double that from a catchment of half the size, all other things being equal. Thus, a more appropriate dependent variable would be the runoff volume expressed as an equivalent depth over the catchment (as for rainfall volume). This transformation does not mean that catchment area should be excluded from the list of independent variables.

This line of argument can be carried a step further by applying the same reasoning to the rainfall volume as to the catchment area, based on the knowledge that the runoff depth is so strongly dependent on the

rainfall depth that benefit could be derived by including the latter in the dependent variable. This is traditionally achieved in one of two ways: (a) by dividing the runoff depth by the rainfall depth to give the proportional runoff (xloo to give the percentage runoff), or (b) by subtracting the runoff depth from the rainfall depth to give the loss.

Finally, the watershed boundaries in an urban catchment may be more closely aligned to the limits of the impervious area alone than to the limits of the whole catchment area. It is possible, therefore, that benefit could be derived from taking the catchment area as equal to the area of impervious surfaces, and deriving the dependent variables accordingly (obtaining a runoff depth by dividing the runoff volume (in litres) by IMPA instead of TOTA).

Based on the above discussion and with no prior evidence to suggest which of the forms of dependent variable might be the most appropriate, regression analyses were performed independently on the four variables as below:

the loss over the total area,

TOTLOSS = PPT - TOTQ ... (9)

the notional loss over the paved area,

IMPLOSS = PPT - IMPQ ... (10)

the percentage runoff over the total area,

TOTPQ = TOTQ/PPT x 100 ... (11)

the notional percentage runoff over the paved area,

IMPPQ = IMPQ/PPT x 100 ... (12)

The variables, TOTLOSS and TOTPQ, are logical and physically understandable. It should be noted that the values of TOTLOSS are always positive whereas values of IMPLOSS range from -17.58 to +11.68. Similarly, TOTPQ has a range from 8.87% to 88.29% whereas IMPPQ sometimes exceeds 100%. Although these anomolies are not desirable from a logical viewpoint, it is felt that the inclusion of all these dependent variables is justified in determining the best form of the equation.

It is difficult to decide at this stage which of the 4 dependent variables is appropriate. A similar exercise for natural catchments (NERC, 1975) concluded that the percentage runoff form of the equation was more appropriate than the loss form. It was decided to perform the analyses using all the dependent variables, and then to compare the four equations obtained.

Table 3 contains a complete list of dependent and independent variables. The complete data set and its associated correlation matrix may be found in the Appendix.

TABLE 3 Definition of variables

	Variatle	Definition	Derivation
<b> </b>	CNO SLOPE TOTA IMPA RATIO	catchment number catchment slope (weighted average pipe slope) (%) total catchment area (ha) impervious (paved and roofed) area (ha) ratio of roofed area to paved area	
A SET	SOIL	soil type index	
BASIC DATA	SNO PPT IMPQ IFWET API5 SMD DUR M5I	storm number rainfall volume (in mm) runoff volume (in mm over impervious area immediate antecedent wetness dichotomy 5-day antecedent precipitation index soil moisture deficit (in mm) duration of rainfall event (minutes) maximum 5-minute rainfall intensity (mm/h)	
≿.,ç	PIMP	percentage impervious (%)	IMPA/TOTA X 100
SUBSIDIARY VARIABLES	AVINT TOTQ CWI	average rainfall intensity (mm/hr) runoff volume (in mm over total area) catchment wetness index	60 PPT/DUR IMPA X IMPQ/TOTA 125 +API5 - SMD
NT ES	TOTLOSS IMPLOSS	rainfall loss over total area (mm) notional rainfall loss over impervious	PPT-TOTO PPT-IMPO
DEPENDENT VARIABLES	TOTPQ IMPPQ	area (mm) percentage runoff from total area (%) notional percentage runoff from impervious area (%)	TOTQ/PPT X 100 IMPQ/PPT X 100

### III: THE ANALYSIS

Multiple linear regressions were fitted using the ASCOP computer package (NCC, 1972) on the Institute's UNIVAC computer, to examine the relationships between variables and thus to derive an equation to predict the volume of runoff that could be expected from a given storm on a given catchment.

The multiple determination coefficient,  $R^2$  (defined as the proportion of explained variance) is a convenient measure of the usefulness of the independent variables for predicting the dependent variable; for this reason  $R^2$  values are quoted throughout.

# Individual catchments

In the first stage of analysis, two dependent variables (TOTLOSS and TOTPQ) were chosen and the linear relationship between these and each independent storm variable were examined for each catchment individually. This identified the storm variables which exert the greatest apparent control on the dependent variables.

The most significant variables, overall, for explaining TOTIOSS were PPT and DUR, (PPT being the single most significant for every catchment) followed by M5I, AVINT, SMD, API5, CWI and IFWET in descending order of importance. Details of the appropriate correlations for each catchment can be found in Table 4; R<sup>2</sup> values are given.

Explanation of TOTPQ was not dominated by any one storm variable and it seemed that storm variables were much less useful in the prediction of TOTPQ than they were for predicting TOTLOSS. The three antecedent wetness variables, CWI, SMD and API5 were the most significant followed by M5I, IFWET, PPT, DUR and AVINT. There was very little difference in the ability of any of these variables to predict TOTPQ; they were all relatively poor. Table 4 gives details of these correlations.

The second stage of the analysis on the individual catchment was to do regressions on the 'best' one, two, three and four variables. (The 'best' regression equation being defined as that one having the highest sum of squares due to the regression). It should be noted that the 'best' one variable does not necessarily appear in the 'best' two variable equation and so on (for example see Table 5, showing the TOTPQ analysis for Blackpool). The R<sup>2</sup> results for the Leicester and Nottingham catchments should be noted with caution as the number of variables approaches the number of points, reducing the degrees of freedom to such an extent that it is virtually impossible not to obtain an R<sup>2</sup> value close to 1.0. Table 5 gives details of these regressions.

By obtaining a 'score' for each variable by giving it a weighting according to its frequency of occurrence as the best first, second, third or fourth variable in the regression equations, we find that the

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TABLE 4 Multiple determination coefficients for individual catchment analyses, expressed as percentages

	GLACKP	DOL	DONCAS	TER	KIDBRO	OKE	LETCES	TER	CXHE	Y 1	OXHE	Y 2	STEVEN	MGE
	TOTLOSS	TOTPO	TOTLOSS	TOTPO	TOTLOSS	TOTPO	TOT, OSS	TOTPO	TOTLOSS	тотро	TOTLOSS	TOTPQ	TOTLOSS	TOTPO
PPT	96.8	0.0	98.3	1.1	95.6	0.6	70.6	64.5	96.5	14,1	70.0	14,9	99.2	1.6
IFWET	15.3	9.6	1.7	0.3	0.2	14.1	0.8	0.2	1.7	1.6	0.4	2.4	0.2	0.0
4915	28.1	3.4	2.3	0.3	1.7	13.2	17.0	0.0	2.2	20.0	0.0	0.8	1.0	2.7
CPM2	0.0	17.2	3.9	15.2	0.1	0.7	28.5	85.5	8.7	42.7	0.9	3.6	1.4	2.3
CUI	1.6	18.7	5.3	16.2	0.0	1.7	18.7	90.2	0.4	47,1	0.9	2.6	1.7	1.3
DUR	77.2	0.1	45.4	2.8	61.0	0.5	26.6	93.6	44.7	14.8	24.6	8.5	27.0	0.1
AVINT	1.0	1.2	8.5	9.4	0.1	0.7	1.1	50.9	15.0	0.0	24.6	4.1	2.3	0.2
M51	17.4	1.0	33.1	21.8	13.0	1.3	15.9	1.8	36.3	0.4	36.0	5.1	29. <i>7</i>	6.4
	BRACKI	Ect	DERSI	1	DERBI	2	DERSY	3	SOUTHAM	TOR 1	SOUTHAN	TON 2	HOTTLNS	HWH
	TOTLOSS	TOTPO	TOTEDSS	101P)	TOTLOSS	10120	TOTLOSS	TOTPQ	TOTLOSS	CALOL	TOTLOSS	TOTPO	TOTLOSS	<u>rate</u> (
PPT	88.9	0.1	96.4	2.7	98.3	0.5	98.5	0.3	93.3	29.1	99.7	5.0	100.0	78.2
FUET	4.5	24.9	1.2	9.5	1.3	0.5	0.6	0.3	1.2	34.4	19.2	10.6	1.0	7.5
AP15	3.6	55.0	0.3	1.7	4.1	10.3	0.8	0.1	2.3	17.1	1.0	33.6	26.2	16.1
SMD	0.4	32.2	9.4	19.3	0.0	23.1	0.0	5.5	3,7	0.9	2.0	0.0	44.7	19.5
C.:1	0.8	41.2	9.0	18.2	0.1	24.2	0.0	6.2	2.2	2.3	1.7	0.6	20.9	11.5
DUR.	54.0	0.1	87.3	2.7	72.6	0.8	71.2	1.1	84.6	17.2	54.3	2.3	25.1	47.4
S/INT	5.9	2.2	2.9	1.3	ó.a	0.1	0.9	9.3	14,4	11.3	0.7	0.2	70.0	39.9
								12.1	3.7	0.1	7.7	2.7	72.8	31.6

TABLE 5 Best n-variable regressions for individual catchment analyses

CATCHMENT	- 1	n=2	BEST N VARTABLI n×3	E5	3 VARIABLE REGRESSION EQUATION SHOWL	NG R <sup>2</sup> VALU
	n=l			***	50.3-3.82 IFWET183 SMD177 MSI	R2+. 29
SLACKPOOL	CHI	IFHET, SMO		PPT, DUR. MS1. CW1	30.1+.092 DUR506 M51+.628 AY [MT	R2+.36
OONCASTER	MSI	M5[, CV]	DUR, MSI, AVINT	DUR, M51, CNI, AVINT	43.7+.329 PPT-5.61 IFMET+.536 API5	R2 = . 17
(LDBROOFE		1FWET, APIS	PPT, IFWET, APIS	PPT, IFWET, DUR, MSI	44045.0 IFMET-7.14 APT5-2.42 DUR	R <sup>2</sup> =1.00
EICESTER	Dijis	APIS, DUR	IFNET, APIS, DUR	IFHET, APTS, SMD, DUR	4.55+3.04 TFMET+.073 DUR+.148 CWI	R2=.61
DXHEY H.E	CAL	PP?, CW1	IFNET, DUR, CNT	TENET, BUR, MST, CHI		27 - 20
XMEY ROAD	PPT	PPI, SMD	SMD, DUR, AVINT	PPT, IFWET, APIS, SHO	44,5+,065 SMD+,306 DUR+,570 AVINT	R2 28
STEVENAGE	H51	MSI, AVINT	OUR, MS1, AVINT	IFWET, APIS, MSI, AVINT	20,6+.034 DUR109 MSL+.286 AVENT	
BRACKNEL L	AP15	SMD, CWI	SMO, DUR, CWI	IFHET, SHO, DUR, CHI	-108.+1.05 940077 DUR+1.19 CWI	R2=.69
DERBY 1	SMD	IFWET, CWI	IFWET, MSI, CWI	(FWET, SMD, MSI, CHI	23.2+8.60 1FWET+.406 N51+,136 CMT	R2 = . 31
DERBY 2	CM1	CWI. AVINT	IFWET, APIS, SMD	IFWET, APIS, SMD, AVINT	36.3+4.80 JFWET+.841 AP15096 SMD	R2 + . 27
ER8Y 3	NS1	SMD, M51	SMD, MSI, AVINT	SEWET, SHO, MSI, AVINY	47.5052 SHD127 M51-,384 AVINT	R2 18
SOUTHAMPTON 1	IFWET		PPT, IFWET, APIS	PPT, IFWET, SMD, CWI	20,5+1.61 PPT-7.66 1FWET+,748 APIS	R2×.60
					23.5-2.31 IFWET+.530 APIS+.087 AVINT	R4= 47
	APIS	AP15, M51		IFWET, APIS, DUR, AVINT		
KOTT LINGHAM	PPT	PPT, HS:	PPT, MSJ. AVENT	PPT, IFWEY, APIS, MSI HDIVIDIAL CATCHMENTS	65331.1 PPT+11.2 M51-26.8 AVINT	92-95
REGRESSION OF	TOTLOSS	PPT, 45:	PPT, MSI, AVINT TORM VARIABLES FOR II BEST N VARIABLE	PPT, IFWEE, APIS, MSI HD[VIDHAL CATCHMENTS	65331.1 PPT+11.2 M51-28.8 AVINT	92- 95 — —
CATCHMENT	PPT TOTLOSS	PPT, 45; ON BEST 1-4 \$ n=2	PPT, MSI. AVENT  TORM VARIABLES FOR II  BEST M VARIABLE N=3	PPT, TEMET, AP15, MS1  HD[VIDIAL CATCHMENTS  ES  n=4	65331.1 PPT+11.Z M51-28.8 AVINT  3 VARIABLE REGRESSION EQUATION SHOW	9295 
REGRESSION OF CATCHMENT BLACKPOOL	PPT TOTLOSS n=1 PPT	PPT, 45: ON BEST 1-4 S n=2 PPT, CWI	PPT, MSI. AVINT  ***CORM VARIABLES FOR II  ***BEST M VARIABLE  ***3  ***PPT, IFHET, CHI	PPT, IFWEY, APIS, MSI MD[V3[NIAL CATCHMENTS ES n=4 PPT, DUR, MSI, CWI	65331.1 PPT+11.Z M51-26.8 AVINT  3 VARIABLE REGRESSION EQUATION SHOWL .599573 PPT+.175 IFMET005 CM1	9295 
REGRESSION OF CATCHMENT BLACKPOOL DONCASTER	PPT TOTLOSS n=1 PPT PPT	PPT, H5: ON BEST 1-4 S n=2 PPT, CWI PPT, M51	PPT, MSI, AVINT  TORM VARIABLES FOR 11  BEST M VARIABLE  N=3  PPT, HENT, CMI PPT, MSI, AVINT	PPT, IFWEY, APIS, MSI MD[VIDIAL CATCHMENTS ES m=4  PPT, DUR, MSI, CWI PPT, DUR, MSI, AVINT	65331.1 PPT+11.2 M5E-28.8 AVINT  3 VARIABLE REGRESSION EQUATION SHOWN 599+.573 PPT+.175 IFWET005 CN1335+.690 PPT+.038 M5I036 AVINT	92- 95 
REGRESSION OF CATCHMENT BLACKPOOL DONCASTER KIDBROOKE	PPT TOTLOSS n=1 PPT PPT PPT	PPT, M5; ON DEST 1-4 S n=2 PPT, CWI PPT, M51 PPT, IFWEY	PPT, MS1, AVINT  TORM VARIABLES FOR 1  BEST N VARIABLE  n-3  PPT, IFHET, CHI  PPT, MS1, AVINT  PPT, IFHET, OUR	PPT, IFWET, APIS, MSI MD[VI]MAL CATCHMENTS ES n=4 PPT, DUR, MSI, CWI PPT, DUR, MSI, AVINT PPT, IFWET, DUR, MSI	65331.1 PPT+11.Z M51-26.8 AVINT  3 VARIABLE REGRESSION EQUATION SHOW: .599+.573 PPT+.175 IFWET005 CM:355+.690 PPT+.038 M51036 AVINT203+.346 PPT+.228 IFWET003 OUR	92 - 95 
REGRESSION OF CATCHMENT BLACKPOOL DONCASTER KIDBROOKE LEICESTER	PPT TOTLOSS  n=1 PPT PPT PPT PPT	PPT, 45; ON BEST 1-4 S n=2 PPT, CWI PPT, M51 PPT, IFWEY PPT, SMD	PPT, MSI, AVINT  YORM VARIABLES FOR 11  BEST N VARIABLE  N=3  PPT, IFHET, CMI  PPT, MSI, AVINT  PPT, SPW, MSI	PPT, IFWEY, APIS, MSI MD[VIDMAL CATCHMENTS ES	3 VARIABLE REGRESSION EQUATION SHOWS 599-573 PPT-175 IFWET-005 CMI -135+690 PPT-038 M5I-036 AVINT -203+546 PPT-228 IFWET-003 OUR	92 - 95 
REGRESSION OF CATCHMENT BLACKPOOL DONCASTER KIDBROOKE ERICESTER OXHEY H.E.	PPT  TOTLOSS  n=1  PPT  PPT  PPT  PPT	PPT, 45:  ON BEST 1-4 S  n=2  PPT, CWI PPT, M51 PPT, IFWET PPT, SMD PPT, CWI	PPT, MSI, AVINT  TORM VARIABLES FOR II  BEST N VARIABLE  n=3  PPT, IFNET, CHI  PPT, MSI, AVINT  PPT, IFNET, OUR  PPT, SHO, MSI  PPT, OUR, CMI	PPT, IFWEY, APIS, MSI MD[VIDIAL CATCHMENTS ES  M=4  PPT, DUR, MSI, CWI PPT, DUR, MSI, AVINT PPT, IFWEY, DUR, MSI PPT, IFWEY, DUR, MSI PPT, IFWEY, DUR, CWI	3 VARIABLE REGRESSION EQUATION SHOWN \$994.573 PPT+.175 IFWET005 CNI3354.690 PPT+.038 MSI036 AVINT2034.546 PPT+.028 IFWET+.003 OUR8294.533 PPT+.025 SWED010 MSI 3.924.784 PPT015 DWR017 CMI	R2= 95 NG R2 VAL R2= 97 R2= 99 R2= 96 R4= 97 R2= 99
REGRESSION OF CATCHMENT BLACKPOOL OONCASTER KIDBROGKE LEICESTER OOKHEY H.E. OOKHEY ROAD	PPT PPT PPT PPT	PPT, MS: ON BEST 1-4 S n=2 PPT, CMI PPT, MS: PPT, ISMD PPT, CMI PPT, CMI PPT, CMI PPT, MS:	PPT, MSI, AVINT  TORM VARIABLES FOR II  BEST N VARIABLE  A-3  PPT, IFMET, CWI  PPT, MSI, AVINT  PPT, SHO, MSI  PPT, OUR, CWI  PPT, SWD, MSI	PPT, IFMET, APIS, MSI  MD[VIDHAL CATCHMENTS  ES n=4  PPT, DUR, MSI, CMI PPT, DUR, MSI, AVINT PPT, IFMET, SMD, AVINT PPT, IFMET, SMD, CMI PPT, APIRT, BMT, CMI PPT, APIRT, MSI, MSI, CMI PPT, APIRT, MSI, CMI PPT, APIRT, MSI, CMI	3 VARIABLE REGRESSION EQUATION SHOWN 5994.573 PPT+.175 IFMET005 CM13354.690 PPT+.036 MSI036 AVINT2034.546 PPT+.228 [FMET003 CUR8294.533 PPT+.025 SMD010 MSI 3.924.784 PPT035 SMD017 CM1 5.499214 PPT037 SMD017 CM1 5.499214 PPT037 SMD017 CM1	R2= 95 NG R2 VAL R2= 97 R2= 99 R2= 96 R4= 97 R2= 99 R2= 99
REGRESSION OF CATCHMENT BLACKPOOL DONCASTER KIDBROOKE EICESTER DXHEY H.E. DXHEY ROAD STEVENAGE	PPT PPT PPT PPT	PPT, 45:  ON DEST 1-4 S  n=2  PPT, CWI PPT, M51 PPT, IFWEY PPT, SMD PPT, CWI PPT, M51 PPT, M51 PPT, M51	PPT, MSI, AVINT  TORM VARIABLES FOR 11  BEST N VARIABLE  PPT, IFHET, CMI  PPT, MSI, AVINT  PPT, IFWET, OUR  PPT, SWD, MSI  PPT, SWG, ANGWY	PPT, IFWEY, APIS, MSI  MD[VIDIAL CATCHMENTS  ES  M=4  PPT, DUR, MSI, CMI PPT, DUR, MSI, AVINT PPT, IFWET, DUR, MSI PPT, IFWET, SUR, CMI PPT, IFWET, BUR, CMI PPT, APIS, MSI, CMI PPT, APIS, MSI, CMI PPT, APIS, MSI, CMI PPT, DUR, MSI, CMI	3 VARIABLE REGRESSION EQUATION SHOWN 599+.573 PPT+.175 IFWET005 CW1335+.690 PPT+.038 MSI036 AVINT203+.546 PPT+.228 IFWET+.003 OWR829+.593 PPT+.025 SW0010 MSI 1.924784 PPT015 DWR017 CM1 .549+.214 PPT003 SW0+.010 MSI0558214 PPT003 SW0+.010 MSI	R2 = 95 NG R2 VAL R2 = 97 R2 = 99 R2 = 96 R4 = 97 R2 = 100
REGRESSION OF CATCHMENT BLACKPOOL DONCASTER KIDBROOKE LEICESTER DOXHEY H.E. DIXHEY ROAD STEVENAGE BRACKHELL	PPT  TOTLOSS  n+1  PPT  PPT  PPT  PPT  PPT	PPT, 45:  ON DEST 1-4 S  0=2  PPT, CWI PPT, M51 PPT, SMD PPT, CWI PPT, M51 PPT, M51 PPT, M51 PPT, M71 PPT, AP15	PPT, MSI, AVINT  TORM VARIABLES FOR II  BEST N VARIABLE  A-3  PPT, IFWET, CMI  PPT, MSI, AVINT  PPT, IFWET OUR  PPT, SWO, MSI  PPT, OUR, CMI  PPT, MSI, AVINT  PPT, MSI, MSI  PPT, MSI, MSI  PPT, MSI, AVINT  PPT, MSI, AVINT	PPT, IFMEY, APIS, MSI  MD[VIDNAL CATCHMENTS  ES	3 VARIABLE REGRESSION EQUATION SHOWN 599-573 PPT+.175 IPHET005 CH1335+.690 PPT+.038 MSI036 AVINT203+.584 PPT228 IPHET003 OUR829+.593 PPT015 DIM017 CM1 .549214 PPT015 DIM017 CM1 .549214 PPT015 DIM017 CM1 .549214 PPT018 DIM010 MSI0550749 PPT018 DIM010 MSI 0510629 PPT099 APIS020 AVINT	R2= 95 R0 R2 VAL R2= 97 R2= 99 R2= 96 R4= 99 R2= 1.00 R2= 96
REGRESSION OF CATCHMENT BLACKPOOL DONCASTER KIDBROOKE LEICESTER DXHEY H.E. REMENT H.E. REMENTER BROOKE BROO	PPT	PPT, MS:  ON DEST 1-4 S  PPT, CHI PPT, MSI PPT, IFWEY PPT, SMD PPT, CHI PPT, MSI PPT, MSI PPT, APIS PPT, APIS PPT, OUR	PPT, MSI, AVINT  YORH VARIABLES FOR 11  BEST N VARIABLE  n=3  PPT, IFHET, CWI  PPT, MSI, AVINT  PPT, IFWET, OUR  PPT, OUR, CMI  PPT, SWE, MSINT  PPT, SWE, ANSWY  PPT, APTS, SWD  PPT, MR, MSI	PPT, IFWEY, APIS, MSI  MD[VIDNAL CATCHMENTS  ES  PPT, DUR, MSI, CMI PPT, DUR, MSI, AVINT PPT, IFMET, SUR, CMI PPT, AFIET, SUR, CMI PPT, AFIET, MSI, CMI PPT, APIS, MSI, CMI PPT, APIS, DUR, AVINT PPT, APIS, DUR, AVINT PPT, APIS, DUR, AVINT	3 VARIABLE REGRESSION EQUATION SHOWS 599-573 PPT-175 IFWET-005 CMI -135-690 PPT-038 M5I-036 AVINT -203-546 PPT-228 IFWET-003 OW -829-593 PPT-025 SMO-010 M5I -929-784 PPT-035 DMR-017 CMI 549-214 PPT-003 SMO-030 M5I -058-749 PPT-1015 M5I-020 AVINT 051-629 PPT-069 API5-008 SMO 0799-728 PPT-075 UMR-055 M5I	R2= 95 R6 R2 VAL R2= 97 R2= 99 R2= 99 R2= 97 R2= 100 R2= 96 R2= 96
METTINGHAM  REGRESSION OF  CATCHMENT  BLACKPOOL  DONCASTER  KIDBROOKE  LEICESTER  DONNEY H.E.  DISHEY ROAD  STREY ROAD  SROCKHELL  DCRBY 1  DCRBY 2	PPT TOTLOSS  n=1 PPT PPT PPT PPT PPT PPT PPT PPT PPT	PPT, MS:  ON DEST 1-4 S  D=2  PPT, CWI PPT, MSI PPT, IFWEY PPT, SMD PPT, CWI PPT, MSI PPT, MSI PPT, MSI PPT, APIS PPT, OUR PPT, OUR	PPT, MSI, AVINT  TORM VARIABLES FOR II  BEST N VARIABLES  PPT, IFHET, CMI  PPT, MSI, AVINT  PPT, SMO, MSI  PPT, CMR, SMI  PPT, CMR, MSI  PPT, CMR, MSI  PPT, CMR, MSI  PPT, CMR, MSI  PPT, CMR, AVINT	PPT, IFWEY, APIS, MSI  MD[VI]MAL CATCHMENTS  ES  n=4  PPT, DUR, MSI, CWI PPT, DUR, MSI, AVINT PPT, IFWET, DUR, MSI PPT, IFWET, SWD, AVINT PPT, APIS, MSI, CWI PPT, APIS, MSI, AVINT PPT, DUR, MSI, AVINT PPT, DUR, MSI, AVINT PPT, DUR, MSI, CWI PPT, DUR, MSI, CWI PPT, DUR, MSI, CWI	3 VARIABLE REGRESSION EQUATION SHOW! 3 VARIABLE REGRESSION EQUATION SHOW! 5994.573 PPT+.175 IFWET005 CNI335+.690 PPT+.038 FIF036 AVINT -203+.545 PPT+.208 FIF036 AVINT203+.545 PPT+.025 SMD010 M51924.784 PPT015 DMR017 CMI 5494.214 PPT015 DMR017 CMI 0514.629 PPT003 SMD030 M510554.749 PPT+.015 M51020 AVINT 0514.629 PPT099 API54.008 SMD .7994.728 PPT012 CMM056 M51 4324.584 PPT012 CMH015 M51	R2 = 95 NG R2 VAL R2 = 97 R2 = 99 R2 = 96 R2 = 99 R2 = 100 R2 = 96 R4 = 96 R6 = 97
MOTTINGHAM  REGRESSIOM OF  CATCIPMENT BLACKPOOL DONCASTER KIDBROOKE LEICESTER DXMEY H.E. DXMEY ROAD STEVENINGE BROZEMELL DERBY 1 DERBY 2 DERBY 2	PPT	PPT, MS:  ON DEST 1-4 S  ON DEST 1-4 S  PPT, CWI PPT, MSI PPT, SMD PPT, CWI PPT, MSI PPT, MSI PPT, MPI PPT, APIS PPT, CWI	PPT, MSI, AVINT  TORM VARIABLES FOR II  BEST M VARIABLE  A-3  PPT, IFWET, CWI  PPT, MSI, AVINT  PPT, SHO, MSI  PPT, SOUR, CMI  PPT, SPE, MSI  PPT, MSI, AVINT  PPT, MSI, AVINT  PPT, CMI, AVINT  PPT, CMI, AVINT  PPT, SPN, DUR	PPT, IFWEY, APIS, MSI  MD[VIDIAL CATCHMENTS  ES  N=4  PPT, DUR, MSI, CMI PPT, DUR, MSI, AVINT PPT, IFWET, SMD, AVINT PPT, IFWET, SMC, CMI PPT, AVINT PPT, AVINT PPT, AVINT PPT, DUR, MSI, AVINT PPT, DUR, MSI, AVINT PPT, DUR, MSI, CMI PPT, DUR, MSI, CMI PPT, DUR, MSI, AVINT PPT, IFWET, DUR, MSI, CMI PPT, IFWET, DUR, MSI, CMI PPT, IFWET, DUR, MSI	3 VARIABLE REGRESSION EQUATION SHOWN 5994-573 PPT+-175 IFMET005 CNI -355+-690 PPT+-038 MSI036 AVINT -203+-546 PPT+-028 IFMET003 OUR -8294-533 PPT+-015 DUR017 CMI 5494-214 PPT033 9MS4030 MSI -0554-749 PPT015 DIR070 AVINT -0518-749 PPT095 APISH-008 SMO -7594-728 PPT012 DUR055 MSI -8394-834 PPT012 DUR055 MSI -8394-834 PPT012 DUR055 MSI -8394-834 PPT012 DUR050 MSI -9594-728 PPT012 DUR050 MSI -9594-739 PPT007 DUR050 MSI -9594-739 PPT007 DUR050 DUR	R2= 95 R6 R2 VALI R2= 97 R2= 96 R4= 97 R2= 96 R4= 97 R2= 96 R4= 97 R2= 96 R4= 97 R2= 96
METTINGHAM  REGRESSION OF  CATCHMENT  BLACKPOOL  DONCASTER  KIDBROOKE  LEICESTER  DONNEY H.E.  DISHEY ROAD  STREY ROAD  SROCKHELL  DCRBY 1  DCRBY 2	PPT TOTLOSS  n=1 PPT PPT PPT PPT PPT PPT PPT PPT PPT	PPT, MS:  ON DEST 1-4 S  D=2  PPT, CWI PPT, MSI PPT, IFWEY PPT, SMD PPT, CWI PPT, MSI PPT, MSI PPT, MSI PPT, APIS PPT, OUR PPT, OUR	PPT, MSI, AVINT  TORM VARIABLES FOR II  BEST N VARIABLES  PPT, IFHET, CMI  PPT, MSI, AVINT  PPT, SMO, MSI  PPT, CMR, SMI  PPT, CMR, MSI  PPT, CMR, MSI  PPT, CMR, MSI  PPT, CMR, MSI  PPT, CMR, AVINT	PPT, IFWEY, APIS, MSI  MD[VI]MAL CATCHMENTS  ES  n=4  PPT, DUR, MSI, CWI PPT, DUR, MSI, AVINT PPT, IFWET, DUR, MSI PPT, IFWET, SWD, AVINT PPT, APIS, MSI, CWI PPT, APIS, MSI, AVINT PPT, DUR, MSI, AVINT PPT, DUR, MSI, AVINT PPT, DUR, MSI, CWI PPT, DUR, MSI, CWI PPT, DUR, MSI, CWI	3 VARIABLE REGRESSION EQUATION SHOW! 3 VARIABLE REGRESSION EQUATION SHOW! 5994.573 PPT+.175 IFWET005 CNI335+.690 PPT+.038 FIF036 AVINT -203+.545 PPT+.208 FIF036 AVINT203+.545 PPT+.025 SMD010 M51924.784 PPT015 DMR017 CMI 5494.214 PPT015 DMR017 CMI 0514.629 PPT003 SMD030 M510554.749 PPT+.015 M51020 AVINT 0514.629 PPT099 API54.008 SMD .7994.728 PPT012 CMM056 M51 4324.584 PPT012 CMH015 M51	R2 = 95 NG R2 VAL R2 = 97 R2 = 99 R2 = 96 R2 = 99 R2 = 100 R2 = 96 R4 = 96 R6 = 97

only result which confirms those made earlier in this section is that PPT is overwhelmingly the most significant variable in explaining the variation in TOTLOSS. The other variables used in explaining TOTLOSS are, in order of frequency of occurrence, M5I, DUR, API5, IFWET, CWI, AVINT and SMD.

The reason for DUR becoming less important than was suggested earlier is that it is strongly correlated with PPT, so that once PPT has been included, there is little benefit to be derived from the inclusion of DUR. This is true also of the three antecedent wetness variables and gives a biassed view of their importance. If API5, SMD and CWI are considered together for the purposes of the above 'scoring' then this combined catchment wetness variable becomes more significant than all the other variables except PPT.

The frequency of occurrence of each storm variable in the 'best' regression equations to predict TOTPQ is as follows: IFWET, M5I, CWI, API5 and SMD, PPT, DUR and AVINT. If the three wetness variables are again considered together, this combined description then becomes the most important variable.

In conclusion to this part of the analysis it is emphasised that, for the explanation of TOTLOSS, the most appropriate variables seem to be PPT, a catchment wetness indicator and M5I, and for the explanation of TOTPQ, a catchment wetness indicator followed by M5I and IFWET seem most appropriate.

A further conclusion that can be made at this stage is the inconsistency of coefficients between catchments. It had been hoped that some pattern of regression coefficients from one catchment to another might have emerged, but no such pattern was observed.

### The complete data set

In the second stage of analysis all catchments were considered together, allowing the introduction of catchment characteristics. These regressions were carried out with both loss (TOTLOSS and IMPLOSS) and both percentage runoff (TOTPQ and IMPPQ) descriptions as dependent variables. The only constraint on the selection of the best prediction equations, using from one to five variables, was the exclusion of the variables CNO, SNO, IMPQ, API5, SMD, RATIO and the other dependent variables. API5 and SMD were excluded to save time and cost because only one antecedent wetness variable was considered necessary in the final equation - CWI is a combination of the other two and also appeared most significant in the examination of individual catchments. RATIO was excluded because it was not considered a useful variable on its own. Tables 6 to 9 show how the regressions altered for the successive introduction of significant variables for the four dependent variables.

An assessment of the magnitude of the residual mean square, as the number of variables increases, indicates the best cut-off point for the number of variables to be included in the prediction equation. A plot

TABLE	6	Regression	οf	TOTLOSS	on	best	variables

VARIABLE	COEFFICIENT	ST. ERROR	T-VALUE	R	$\mathbb{R}^2$
CONSTANT PPT	00408 .61395	.08528 .01066	0478 57.5675	.9490	.9005
CONSTANT PPT PIMP	2.1312 .59403 04094	.20514 .00939 .00367	10.3889 63.2358 -11.1545	.9622	<b>.92</b> 58
CONSTANT SOIL PPT PIMP	4.6950 -4.1885 .59986 06169	.32943 .44607 .00846 .00397	14.2519 - 9.3897 70.8819 -15.5408	.9697	<b>.94</b> 03
CONSTANT SOIL PPT DUR PIMP	4.2762 -3.9827 .65132 00408 05671	.33101 .43485 .01339 .00084 .00399	12.9186 - 9.1588 48.6481 - 4.8665 -14.2306	.9716	<b>.94</b> 39
CONSTANT SOIL PPT DUR CWI PIMP	4.441 -3.5834 .64953 00393 00383 - 0.5738	.32989 .44448 .01321 .00083 .00113 .00393	13.4623 - 8.0620 49.1743 - 4.7450 - 3.3982 -14.5887	.9725	.9457

TABLE 7	Regression of	IMPLOSS on bes	t <b>vari</b> ables		
VARIABLE	COEFFICIENT	ST. ERROR	T-VALUE	R	R <sup>2</sup>
CONSTANT TOTA	1.1396 01060	.12263 .00183	9.2936 -5.7779	.2891	.0836
CONSTANT TOTA CWI	2.1551 01024 01405	.23329 .00178 .00278	9.2380 -5.7644 -5.0565	. 3789	.1436
CONSTANT TOTA PPT CWI	1.6367 01204 .10390 01486	.24833 .00176 .02067 .00270	6.5911 -6.8531 5.0271 -5.5128	. 4463	. 1992
CONSTANT TOTA DUR M51 CWI	1.1582 01228 .00580 .03463 01506	.27522 .00174 .00126 .00693 .00264	4.2083 -7.0699 4.5942 4.9966 -5.6991	.4834	.2337
CONSTANT TOTA SOIL DUR M5I CWI	1.8926 01162 -2.6455 .00610 .03614 01239	.37686 .00174 .93734 .00125 .00689 .00278	5.0219 -6.6977 -2.8223 4.8601 5.2472 -4.4499	. 5002	. 2502

TABLE 8	Regression of	TOTPQ on best	variables	· · · · · · · · · · · · · · · · · · ·	
VARIABLE	COEFFICIENT	ST. ERROR	T-VALUE	Ř	R <sup>2</sup>
CONSTANT PIMP	9.8353 .59820	2.4960 .04884	3.9404 12.2472	. 5392	.2907
CONSTANT SOIL PIMP	-30.115 63.867 .92127	4.2433 5.8312 .05168	- 7.0969 10.9527 17.8274	.6828	. 4661
CONSTANT SOIL CWI PIMP	-33.353 55.081 .08262 .93192	4.1202 5.8229 .01483 .04971	- 8.0951 9.4594 5.5696 18.7466	. 7128	. 5081
CONSTANT IMPA SOIL CWI PIMP	-46.628 .23850 62.569 .08579 1.1090	5.2768 .06096 6.0238 .01457 .06652	- 8.8363 3.9126 10.3869 5.8869 16.6709	.7266	. 5280
CONSTANT IMPA SOIL IFWET CWI PIMP	-44.820 .23720 62.016 - 2.3856 .08290 1.1129	5.3268 .06069 6.0033 1.1613 .01458 .06625	- 8.4142 3.9084 10.3303 - 2.0543 5.6865 16.7974	.7304	. 5334
TABLE 9	Regression of	IMPPQ on best	variables	<del></del>	<del></del>
VARIABLE	COEFFICIENT	ST. ERROR	T-VALUE	R	R <sup>2</sup>
CONSTANT SOIL	43.86 <b>4</b> 100.65	4.1138 10.409	10.6626 9.6694	.4511	,2035
CONSTANT SOIL CNI	37.449 75.901 .21584	3.9913 10.463 .03161	9.3827 7.2545 6.8277	. 5419	. 2937
CONSTANT IMPA SOIL CWI	37.083 .42840 69.288 .21789	3.8897 .09465 10.298 .03080	9.5336 4.5260 6.7282 7.0734	.5756	.3313
CONSTANT IMPA SOIL CWI PIMP	- 8.6932 .81206 102.20 .22679 .61553	10.896 .12587 12.438 .03009 .13735	7979 6.4518 8.2167 7.5364 4.4814	.6053	.3664
CONSTANT IMPA SOIL IFWET CWI	- 6.0369 .81015 101.39 - 3.5057 .22253	11.030 .12568 12.431 2.4047 .03019	5473 6.4463 8.1558 - 1.4579 7.3716		
PIMP	.62135	.13720	4.5289	.6083	.3701

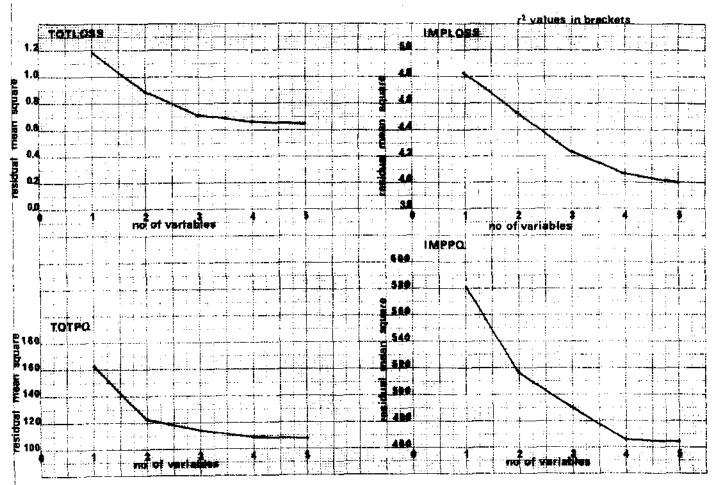


FIGURE 4 Appropriate cut-off points for the loss and percentage runoff regressions

of the residual mean square against the number of variables is shown in Figure 4 for the loss and for the percentage runoff dependent variables. From this it appears that three variables are most appropriate in the prediction of TOTLOSS and four variables in the prediction of IMPLOSS, TOTPQ and IMPPQ.

This gives the following 'best' equations:

Looking at the R<sup>2</sup> value (from Tables 6 to 9) for these four equations, we find that 94.0% of the variation in TOTLOSS is explained, 23.4% for IMPLOSS, 52.8% for TOTPQ and 36.6% for IMPPQ. This suggests that the variables calculated using runoff over the total area (rather than the depth of runoff over the paved area) are easier to predict. It is also evident that a far larger percentage of the variance in TOTLOSS is explained compared to the variance explained for TOTPQ.

# Comparison of equations

It was felt that the selection of the best equation for predicting the runoff from urban catchments should not be made solely with regard to the R<sup>2</sup> values. A comparison of the four different equations (13 to 16 shown above) was made by using each equation to predict each of the other dependent variables in turn. At first sight, the manipulation of regression equations might seem theoretically unattractive. However, if the four equations are considered as mathematical models (the fact that they are derived through regression analysis is irrelevant), then their performance may be tested against a number of objective functions. The objective functions are, in this case, the proportion of explained variance of each of the dependent variables. In addition, a further objective function was used to test the four models deriving from a measurement and prediction of the absolute volume of runoff in litres. Table 10 shows the results of this comparison, a maximisation of the first four and a minimisation of the last objective functions being desirable. At first sight, the negative R2 values seem erroneous; the explanation is that the model is inferior to an estimate equal to the average value of the dependent variable. The circled values of the objective functions in the leading diagonal are the multiple determination coefficients derived from the regression analyses shown in Tables 6 to 9.

The chief conclusion to be drawn from Table 10 is that the percentage runoff equations are more appropriate than the loss equations. The fact that they do better after manipulation at explaining the variance of the loss dependent variables means that the percentage runoff form of the regression provides a more realistic model of the actual processes. Conversely, the loss equations provide a poor prediction of the percentage runoff variables. In terms of the objective functions, there is little to choose between the two percentage runoff equations. The higher value of the multiple determination coefficient would seem to favour the TOTPQ equation.

TABLE 10 Comparison of regression equations

		<del></del>		OBJECTI	VE FUNCTION	
		R <sup>2</sup> TOTLOSS	R <sup>2</sup> IMPLOSS	R <sup>2</sup> TOTPQ	R <sup>2</sup> IMPPQ	$\sum \left[ \frac{(Q_{PRED} - Q_{OBS})}{Q_{OBS}} \right]$
	TOTLOSS	(94)	.04	19	-1.96	179
ENT EE	IMPLOSS	.94	(23)	.01	26	67
END	TOTPQ	.96	.42	(53)	.32	42
DEPENDENT VARIABLE	IMPPQ	.96	. 40	.52	37	. 38

# Examination of the residuals

As a final aid in the examination of the most appropriate equation the residuals for all equations were studied. For the TOTPQ equation, plots of (a) residuals against observed rainfall volume (PPT), (b) residuals against observed TOTPQ, and (c) residuals against predicted TOTPQ, are shown in Figures 5 to 7. For further examination, each catchment is plotted using a different symbol. The two major points arising out of these plots are (i) the tendency of the model to overestimate low percentage runoffs and to underestimate high percentage runoffs (see Figure 6), which is to be expected, and (ii) the fact that the larger residuals tend to occur for 'smaller' rainfall events (see Figure 5).

# Refinement of the final equation

The analyses described above concluded that the most appropriate model was the TOTPQ regression given in equation 15. Beyond this, a number of adjustments and refinements were made to this equation before a final form was adopted.

Firstly, the form of the four-variable equation leaves something to be desired in that the combination of IMPA and PIMP is intuitively very unsatisfactory (there is an inverse correlation of .63 between them). Table 8 shows that dropping IMPA from the final equation makes very little difference to the performance of the model (as indexed by the multiple determination coefficient). Thus the three-variable form was adopted, given by:

$$TOTPQ = -33.4 + .932PIMP + 55.1SOIL + .0826CWI$$
 ... (17)

It was further observed that the statistical population from which the values of SMD and API5 were derived has a marked tendency to embrace summer storm events (whereas the natural catchment data set from which the CWI was derived - NERC, 1975 - comprises mainly winter season events). During the summer season, SMD values are generally high and API5 values low (in the data set used here, average values of API5 and SMD are 5.67 mm and 57.8 mm respectively). It was thus felt that CWI was too heavily weighted in favour of SMD and that some modified index might be more appropriate. A modified definition of the catchment wetness index (and renamed the urban catchment wetness index, UCWI) was used given by:

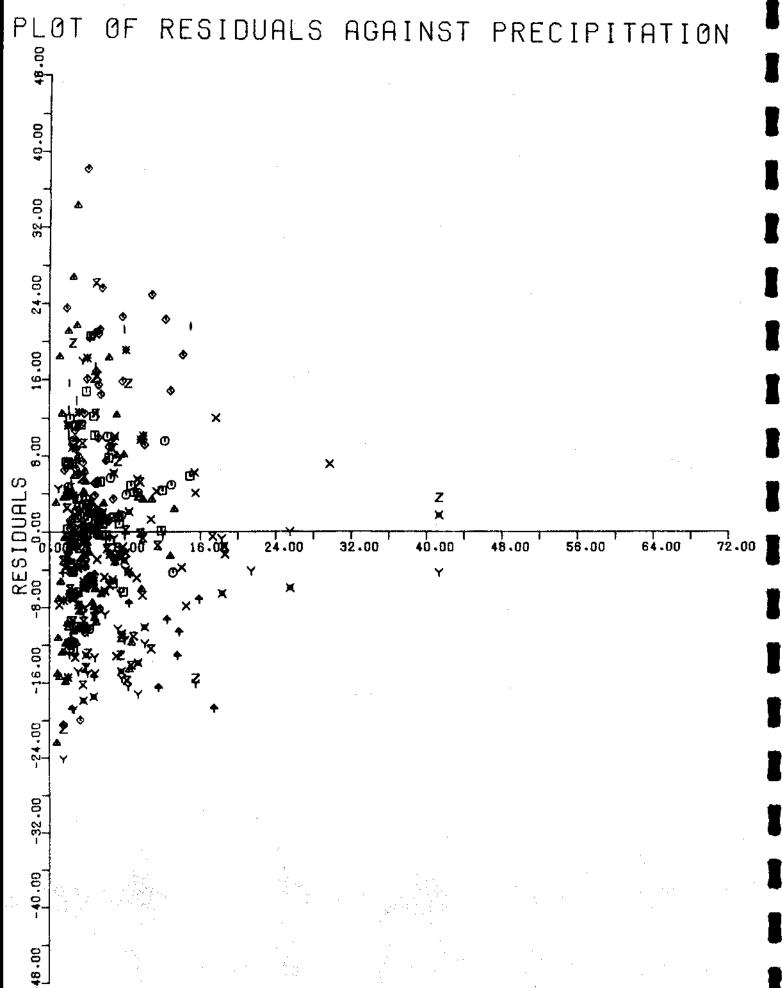
$$UCWI = 125 + (n * API5) - SMD$$
 ... (18)

where n is some weighting factor.

Regressions were done for integer values of n between 1 and 12. It was found that the R<sup>2</sup> value was highest for a value of n equal to 8. ASCOP outputs for the original definition (n=1) and for the n=8 case are shown in Figures 8 and 9 respectively. Using this modified definition, the following equation was adopted as being more suitable than that given in equation 17:

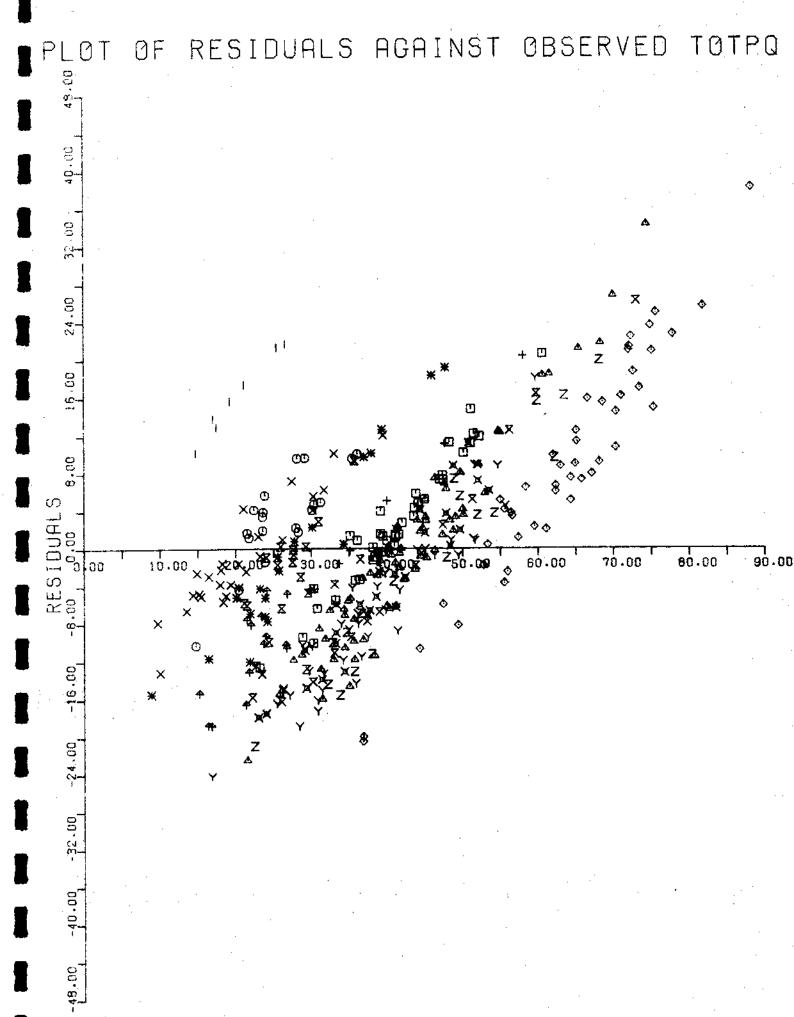
FIGURES 5, 6 and 7Plots of residuals against precipitation, observed TOTPQ and predicted TOTPQ Key Blackpool Doncaster Kidbrooke Leicester Oxhey Housing Estate Oxhey Road W.P.R.L., Stevenage Wildridings, Bracknell Derby 1 Derby 2 Derby 3 Southampton 1 Southampton 2

Nottingham

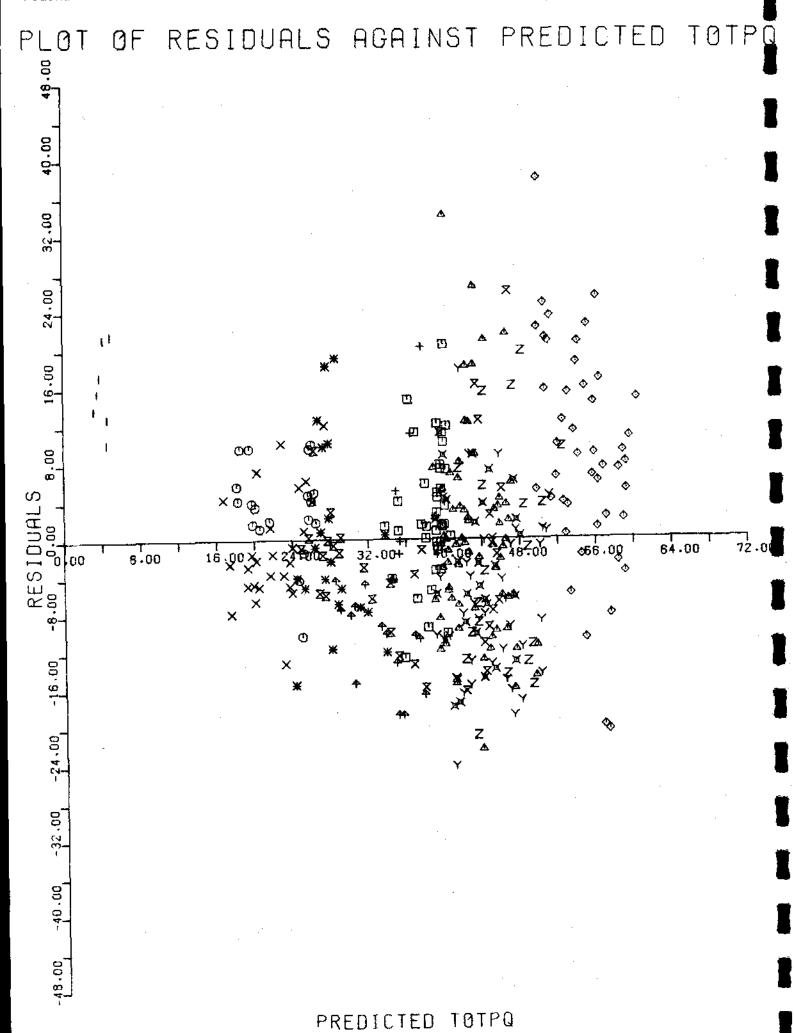


PRECIPITATION

FIGURE 6



OBSERVED TOTPO



$$TOTPQ = -33.6 + .924PIMP + 53.4SOIL + .0649UCWI$$
 ... (19)

The plot of the residuals against predicted TOTPQ for equation 15 shown in Figure 7 do not display any trend, indicating that no better fit was likely to be achieved by transforming the variables. However, a multiplicative form of the model was tested by performing a regression analysis on the logarithmic transforms of the variables. The variables came into the analysis in the same order as that shown in Table 8. This resulted in a three-variable equation given by:

$$TOTPQ = .640 \text{ PIMP}^{.953} \text{SOIL}^{.243} \text{ UCWI}^{.139} \dots (20)$$

This equation explained 51% of the variance in log (TOTPQ) and, undertaking the same kind of comparison test demonstrated in Table 10, explains 47% of the variance in TOTPQ. As such, the multiplicative form of the equation is considered slightly inferior to the additive version.

A partial multiplicative form of the model was then considered. A complete analysis of all possible combinations is clearly impracticable but two possibilities appeared to show some promise. Firstly, a multiplicative combination of the soil index with the percentage of pervious area resulted in the following equation:

TOTPQ = 
$$-55.6 + 1.32PIMP + 1.13SOIL (100-PIMP) + ... (21)$$
  
.0655UCWI

This equation explains only 51% of the variance in TOTPQ. Secondly, a multiplicative combination of SOIL and UCWI resulted in:

$$TOTPQ = -10.3 + .790PIMP + .237SOIL.UCWI$$
 ... (22)

This equation explains 49% of the variance of TOTPQ. This result is not altogether surprising bearing in mind the fact that it has one less degree of freedom than equation 19.

Finally, it was considered prudent to investigate whether the percentage runoff was dependent on the return period of the rainfall event (intuition suggests that the percentage runoff might increase for rarer events). A return period was generated (NERC, 1975) for each event based on the total rainfall volume and the storm duration (a more rigorous approach would be to use the maximum rainfall volume during some critical duration for each catchment - but the approach used was much simpler and was considered to provide a satisfactory approximation). The return period and the percentage runoff have a correlation coefficient of -.13. A regression of the residuals of equation 19 on the return period yielded a correlation coefficient of .01. Based on the above analyses of this data set, it was concluded that the percentage runoff was independent of the return period.

ASCOP output for TOTPO regression (UCWI weighting factor n = 1) FIGURE 8

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PROB	00:00	00000	0000	•0000
T-VALUE	-8.0951	#654°6	5.5696	18.7460
STANDARD ERACR	.412.116+001	.582294+001	148345-001	.497115-621
COEFFICIENT	3335304302+**	.550814+005***	.826215-001***	*431919+600***
VAFIABLE	CONSTANT	2016	C W I	dwld

MATRIX CONTAINING VARIANCES AND COVARIANCES OF REGRESSION COEFFICIENTS

SOIL	.339067+002234007-301	234007-Q01 .E20062-C03	

# ANALYSIS OF VARIANCE OF FIT

DUE TO REGRESSION .426780+065 3 .142260+065 .119273+063 125-3147 .0000 ABOUT REGRESSION .413220+065 364 .113522+063 .136547+062 125-3147 .0000 total	SOURCE	SUM DE SOUARES	i.	MEAN SUCARE	KOOL MEAN SQUARE VAR-KATIO		80 ad
.413220+005 364 .113522+003 .84604005 367 .228883+003	DUE TO REGRESSION	. 426780+365	M)	+142260+005	*119273*003	125-3147	0000
.845040+665 367 .228883+803	ABOUT REGRESSION	.413220+055	364	*113522+003	+136547+002		
	TOTAL	.8489400+665	367	.228883+803	*151284+C62		

# UNKEPLICATED DEPENDENT VARIABLE - NO GOOGNESS OF FIT TEST IS POSSIBLE

.5 98 07	71279
•	•
<b>.</b>	11
COEFFICIENT	COEFFICIENT
DETERMINATION	CORRELATION
MULTIPLE	MULTIPLE

DETERMINANT = .592432+NGO

29

ASCOP output for TOTPO regression (UCWI weighting factor n = 8) FIGURE 9

TITEE

VOLUME OF RUNOFF BUALYSIS

DATA MATRIX ONE POINTS 368 VARIABLES EACH LITH MULTIPLE REGRESSION ANALYSIS USING

REGRESSION EQUATION DESCRIBING VARIABLE TOTPO

MUL COR COF		9403316°	. 9885908	*9757724
PROB	0000.	00000	0000.	00000
T-VALUE	-4.3639	6-4023	7.0480	19.0329
STANDARD ERPCR	. 401631+001	.567621+001	.920176-002	.485#31-001
COEFFICIENT	-*335922+002***	.533706+002×**	***I (I)-bb58b9*	*452010+01020*
VAKIABLE	CONSTANT	SOIL	UCHI	PIME

MATRIX CONTAINING VARIANCES AND COVARIANCES OF REGRESSION COEFFICIENTS

<b>山東田山</b>	*151202+000	+345541-006	*235644-002
(.cm	137057-001	.846724~064	. 345841-005
SOIL	.322193+002	137037-001	.151202+000
	5016	JCW1	JKP

ANALYSIS OF VARIANCE OF FIT

	3000° 8			
VAR-RATIO	136.9628			
ROOT MEAN SOUARE	•121840+003	.194117+002	*15128°+062	
AEAN SOUARE	*148471+C38	*108403+003	+228883+003	
90	\$5.1	364	Sol	
SHA OF SQUARES	4445434+005	304986408.	. 846000+005	
SOURCE	DUE TO REGRESSION	ABOUT REGRESSION	TOTAL	

UNREPLICATED REPENDENT VARIABLE - NO GOODNESS OF FIT TEST IS POSSIBLE

	•
Į,	ŧ,
COEFFICIENT	CORFFICIENT
DETERMINATION	CORRELATION
MULTIPLE	MULTIPLE

530255

. 6055H4+600 DETERMINANT =

### IV: CONCLUSIONS

# General conclusions

Some specific conclusions were discussed in the previous section. This section draws more general conclusions related to the whole analysis exercise.

By way of summary, the final equation adopted as most suitable for prediction of the volume of runoff is given by:

$$TOTPQ = -33.6 + .924PIMP + 53.4SOIL + .0649UCWI$$
 ... (23)

where TOTPQ is the percentage runoff (runoff expressed as mm over the total catchment area),

PIMP is the percentage impervious,

SOIL is the soil index,

and UCWI is the urban catchment wetness index, given by (125 + 8API5 - SMD).

This equation is that deemed most appropriate from the analyses discussed in Section III. It explains 53% of the variance of TOTPQ (correlation coefficient .73). The values of Student's t quoted in Figure 9 demonstrate that the regression coefficients of the dependent variables are highly significant (much greater than at the .01% level).

The unexplained variance is 47%. Of this, it is likely that approximately half can be ascribed to error in the data. The remainder is due to the inability of the regression model to simulate the very complex processes highlighted in Section I.

It is important that equation 23 is not used outside the limits of the data set. The limits of the data are as follows:

PIMP: 20-70% SOIL: .15-.45

UCWI 0-330 (API5 0-32 mm SMD 0-125 mm)

Soil index values cover almost the whole range as defined by equation 3 in Section II, and there are therefore no limitations in this respect. Values of the catchment wetness indices cover a range which one might expect to encompass most situations, although some caution might be necessary under very wet conditions. PIMP has the strongest influence of the three dependent variables, and it is essential that the equation is not used outside the 20-70% limitations.

The discussion in Section I indicated a number of variables upon which the volume of runoff might depend. In the regression analysis which followed, a number of these proved to be insignificant. It would have

been more satisfactory if storm variables had exerted a stronger control on the final equation adopted - in fact the inferior equation 14, using IMPLOSS as the dependent variable, contains more of the variables upon which one might intuitively have hoped that the volume of runoff might depend. The exclusion of these variables does not mean that they do not in some degree control the runoff volume process, but that their influence becomes insignificant in the context of the model adopted.

The inclusion of a catchment wetness index in the final equation has significant implication. Except where calibration of the model has been allowed, even the more sophisticated urban simulation models do not make allowance for the variation of antecedent catchment wetness conditions. These analyses suggest that the antecedent wetness exerts quite a strong control on the runoff volume.

Two of the variables excluded merit special attention, the first of which is the depression storage dichotomy, IFWET. The implication of its exclusion is that the volume of depression storage is small compared to the rainfall volume, which is especially true for large storm events. This is not to say that a depression storage submodel is not an important component of a complete above-ground model, because the fact that it needs to be satisfied at the beginning of an event has a profound effect on the timing of the response hydrograph.

The second excluded variable which deserves some mention is the Catchment slope. Bearing in mind the findings of Sarginson (1973), its exclusion is at first sight somewhat surprising. But the correlation matrix in the Appendix shows that there is a strong inverse correlation between slope (SLOPE) and percentage impervious (PIMP) (the implication being that society has a tendency to build in flat valleys rather than on the sides of hills!). It is likely then that the effect identified by Sarginson is included implicitly by the inclusion of the percentage of impervious area.

### Future work

The number of catchments (14) contributing to the data set is smaller than is desirable for a statistical analysis of this type. For this reason, it is intended to extend the data set to include data from a number of other catchments. In mind at the moment are data from catchments at Stevenage, Crawley, Nottingham and Hendon. It is intended to incorporate further catchments without degrading the quality of the data. Further data can be obtained by extending the existing archive to include further rainfall events on some of the catchments used in this study. It is further hoped that some catchments having data of dubious quality at present might be incorporated when theoretical discharge calibrations have been checked by dilution gauging. When these data become available, the final equation will simply be amended without the extensive analyses described here.

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THE APPENDIX

A complete listing
of the 368 data points
used in the analysis
and
a correlation matrix
for the complete data set.

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7105	.45	455	.45	45	1	- d	7	א ה		7 E	ក <u>ឃ</u> វិ ដ	) 	ก แ * -	) u	n u ‡ ≥	) i	າ ກໍ	c <del>t</del> .	. 45	54.5	<u>ن</u> ج	10.	.45	45	645	ហ a	.45	#55	. 45	, #5	្រុះ	04		0 1	O T	) 	4.0	0.4	? •
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SLOPE	600	80.	α 2 -	) C	0.0	<b>&gt;</b> 0	GD • T	30°I	φ,	•	1,08	•	1.08	0	0	0	C	Ç	•		1.08	•	, ,	•	, -						•	•	• ~	•	• ]	•	N 0	•	•
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19	.73	┣~	ው	27	4	$\alpha$	#	9	4.5	Ç	4	О	0	N	-	┣~	86.	. 3	#	Φ)	#	Φ	<.n	C.	С	#	S.	0	0	<b>.</b> 74	Φ)	œ	• 95	• <b>6</b> 6	•79	
χ. ιζ	O1	$\sim$		#	_	~	0.1	$\odot$		Ю	~	:+	Œ	N)	V.		20	_	-	~	$\alpha$	$\odot$	0	0	M	<b>►</b>	യ	œ	Ø	**	Q.	12	12	12	σ	
DUR	26	95	17	74	50	16	<del>1</del> 6	α	ភ	Œ	Œ	<del>.</del> †	ĸ	41	56	95	11	74	<b>2</b> 0	16	16	α	ប	<b>8</b>	œ	Œ	20	Ю	28	52	31	<b>5</b> 6	34	21	116	
SMD										Ç.		25.					10(.77								(V		<b>.</b>					97.	08		9	
APIS	•	5.69			•	•	95	10. 10.	.73	.24	1.90	. •	•	•	•	•	1.37	•			. 95	J.	٠	2.87		3	6.57	3	86.	.27		6.12	•		<b>†9</b> •	
IFWET	0	0	بر.،	c	0	<b>0</b>	C.	0	<b>4-</b> -!	<b>~</b> ~	O	يسو	<del>, ,</del>	<b>-</b>	Ċ		C	0	c	0	0	0	vН	0	+-1	0	0	<del>-</del> -1	4	<del>, -1</del>	-	۵	<b>~</b> ⊷1	<b>4</b> -4	<b>-</b>	
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RATIO	- ⇒	84.	ŧ	<b>寸</b>	⇉	#	877*	⇉	⇉	ⅎ	⇉	⇉	⇉	1.0	0	?	1.03	•	<b>-</b>	0	0.	0	•	•	9	0	0	•	6	σ	6	σ	96.	86.	• 98	
IMPA	.33	₩•	<u>်က</u> ်		, · K)	3.5	<b>.</b> 35	3	Ω <b>10</b>	ĸ	10	က <b>်</b>	3	Q	Q	.25	.25	.25	.25	.25	20	. 25	.25	. 25	Ø	, 25	. 25	Ø	9.2	9.2	9.2	19.2	9.2	9.2	9.2	
TOIA	.80				.80			.80	. 80		.80	.80	.80		• 60	. 60	.60	. 60	09.	99.		09.	. 60	09.	.60	9.		.60	å	Ň	å	62.00	å	å	å	
SLOPE	•	•	•	•	•	•	•	•	•	•	•	•			•	•	1:0°1	•	•		•	•	•	•	•		•		•			•	•	•		
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	T0TPQ	IMPPQ	TOTPQ IMPPQ TOTLOSS IMPLOSS	S IMPLO:	SS SLOF	SLOPE TOTA	A IMPA	A PIMP		RATIO SC	SOIL P	ppT I	IFWET A	API5 S	SMD	DUR	M5I	CWI	AVINT	
OTFO	00.	.63	.30	39	ı	. 37	.36	.36	54 -	.49	.04	- 50.	90	.23	16	.03	10	.18	16	
Oddini	63	1.00	00.	73	Ÿ.	. 05	. 26 .	25	- 92	.29	.45	4.	. 15	. 32	41	.04	.08	.44	01	
70TI 05S	•	00	1,00	.27		. 05	.33	32	34	.12	.15	.95	- 70.	. 70,	03	79.	.47	.02	.16	
WPI 055		73	.27	1.00	ı	90.	. 29 - ,	. 82	.18	- 60.	.22	.16	.01.	26	.23	.18	.13	26	90.	
SLOPE	37	.05	.10	90	_	8.	. 22 .	. 22	54	.19	. 16	8	13	.04	20	12	.07	.19	.21	
TOTA	- ,36	. 26	.33	. 29		.22 ].	.8	66	.64	.34	9(.	.Z.	80.	.05	03	· .0	.24	.04	,14	
IMPA	36		.32	28		. 22	.99 1.	00.	.63	.35	<del>ن</del> ب	.20	90.	.03	03	ō.	.21	.04	60.	_
D I M D	40.	ı	34	.18	•	. 54	49.	63 1.	- 00.	- 15 -	. 57	19	=	07	.23	90.	22	23	23	
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5011	40			22		.16	. 16.	٠ ۲:	- 73.	.61	1.00	.17	12	.17	34	60.	Ξ.	.35	80.	~
7 T D D	50° -			91.	1	8	. 12.	.20 -	- 61.	.02	٦١.	1.00	.07	.03	07	.78	.42	.07	S.	_
   180	( ) ( )	1		.10		13 -	. 80	90.	Ξ.	- 70.	12	.07	1.00	٦. 61	90.	.12	07	٤١	60° -	_
1 15 1 15 1 15 1 15	.23	•	í	- ,26		.04	.05	.03 -	- 70,	.16	. 17	03	61	1.00	22	- 09	.13	.34	.12	۸.
- <del>S</del>	16	1	ı	.23	ı	- 02.	.03 -	.03	.23	.13 -	.34	07	90.	22	1.00	• .08	03	99	90.	
Silo	რ	•	•	.18	1	. 12 -	.0	<u>ا</u>	- 00.	90.	60.	.78	.12	60	08	3.00	- 98	90.	29	_
15	01			.13		.07	.24	- 21 -	.22	00.	=	.42	07	.13	01	08	1.00	.03	.73	~
: is	81.	_		- ,26		<u>6</u> .	.04	- 04	. 23 -	.15	.35	,0 <b>7</b>	- 3	.34	99	90.	.03	1.00	- 04	<b>5</b> +
AVINT	96.	1		.00		.21	,14	- 60.	.23	.02	.08	60°	- 00	.12	90.	29	.73	04	1.00	_
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