

# 1 Development of a Streamflow Drought Severity-Duration-Frequency 2 Curve Using a Threshold Level Method

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## 10 11 Abstract

12 This study developed a streamflow drought severity-duration-frequency (SDF) curve  
13 that is analogous to the well-known depth-duration-frequency curve used for rainfall. Severity  
14 was defined as the total water deficit volume to target threshold for a given drought duration.  
15 The fixed and variable threshold level methods were introduced to set the target instreamflow  
16 requirement, which can significantly affect the streamflow drought severity. The four  
17 threshold levels utilized were the fixed, monthly, daily, and desired yield for water use. The  
18 fixed threshold level in this study is the 70<sup>th</sup> percentile value ( $Q_{70}$ ) of the flow duration curve  
19 (FDC), which is compiled using all the available daily streamflows, and the monthly  
20 threshold level is the monthly-varying  $Q_{70}$ s of each month's FDC. The daily variable  
21 threshold is the  $Q_{70}$  of the FDC obtained from the antecedent 365 daily streamflows. The  
22 desired yield threshold determined by the central government consists of domestic, industrial,  
23 and agricultural water uses as well as environmental instreamflow. As a result, the desired  
24 yield threshold can identify streamflow droughts using the total water deficit to the  
25 hydrological and socioeconomic targets, while the fixed, monthly, and daily streamflow  
26 thresholds derive the deficiencies or anomalies from the average of historical streamflow.  
27 Based on individual frequency analyses, SDF curves for four thresholds were developed to  
28 quantify the relation among severities, durations, and frequencies. For additional specification,  
29 the drought duration-frequency curve was developed. This curve can be an effective tool for  
30 identifying streamflow droughts using severities, durations, and frequencies.

31  
32 *Keywords:* frequency analysis, streamflow drought, severity-duration-frequency (SDF)  
33 curve, threshold level method

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## 35 **1. Introduction**

36 A drought implies a period of time when the supply of water cannot meet its typical  
37 demand. Rainfall deficiencies of sufficient magnitude over prolonged durations and  
38 subsequent reductions in the streamflow interfere with the normal agricultural and economic  
39 activities of a region, leading to a decrease in agriculture production and thus affecting  
40 everyday life. Dracup et al. (1980) defined a drought using the following properties: 1) nature  
41 of water deficit (e.g., precipitation, soil moisture, or streamflow); 2) basic time unit of data  
42 (e.g., month, season, or year); 3) threshold for distinguishing low flows from high flows,  
43 while considering the mean, median, mode, or any other derived thresholds; and 4)  
44 regionalization and/or standardization. Based on these definitions, various indices were  
45 proposed over the years to identify drought. Recent studies have focused on such multi-  
46 faceted drought characteristics using various indices (Palmer, 1965; Rossi et al., 1992; McKee  
47 et al., 1993; Byun and Wilhite, 1999; Tsakiris et al., 2007; Pandey et al., 2008a; 2008b; 2010;  
48 Nalbantis and Tsakiris, 2009; Wang et al., 2011; Tabari et al., 2013; Tsakiris et al., 2013).

49 The American Meteorological Society (1997) groups drought definitions and types into  
50 four categories: meteorological or climatological, agricultural, hydrological, and  
51 socioeconomic droughts. The meteorological drought is a result of the absence or reduction of  
52 precipitation and short-term dryness results in an agricultural drought that severely reduces  
53 crop yields. Precipitation deficits over a prolonged period that reduce streamflow,  
54 groundwater, reservoir, and lake levels will result in a hydrological drought. If hydrological  
55 droughts continue until the supply and demand of numerous economic goods are damaged, a  
56 socioeconomic drought occurs (Heim, 2002).

57 Hydrological and socioeconomic droughts are very difficult to be approached.  
58 Hydrological drought is defined as a significant decrease in the availability of water in all its  
59 forms, appearing in the land phase of the hydrological cycle. These forms are reflected in  
60 various hydrological variables such as streamflows, which include snowmelt and springflow,  
61 lake and reservoir storage, recharge of aquifers, discharge from aquifers, and baseflow  
62 (Nalbantis and Tsakiris, 2009). Therefore, streamflow is the key variable to analyze in  
63 describing hydrological droughts because it take into account outputs of four different  
64 subsystems, i.e., surface runoff from the surface water subsystem, subsurface runoff from the  
65 upper and lower unsaturated zone, and baseflow from the groundwater subsystem (Tsakiris et  
66 al., 2013). Furthermore, streamflow crucially affects the socioeconomic drought for several  
67 water supply activities such as hydropower generation, recreation, and irrigated agriculture,  
68 where crop growth and yield are largely dependent on water availability in the stream (Heim,  
69 2002). Hence, hydrological and socioeconomic droughts are related to streamflow deficits  
70 with respect to hydrologically normal conditions or target water supplies for economic growth  
71 and social welfare.

72 For additional specification, Tallaksen and van Lanen (2004) defined streamflow drought  
73 as a “sustained and regionally extensive occurrence of below average water availability”.  
74 Thus, threshold level approaches to define the duration and severity of a drought event while  
75 considering the daily, monthly, seasonal, and annual natural runoff variations are widely  
76 applied for drought analyses (Yevjevich, 1967; Sen, 1980; Dracup et al., 1980; Dalezios et al.,  
77 2000; Kjeldsen et al., 2000; American Meteorological Society, 2002; Hisdal and Tallaksen,  
78 2003; Wu et al. 2007; Pandey et al., 2008a; Yoo et al., 2008; Tigkas et al., 2012; van  
79 Huijgevoort, 2012). These approaches provide an analytical interpretation of the expected  
80 availability of river flow; a drought occurs when the streamflow falls below the threshold  
81 level. This level is frequently taken as a certain percentile flow for a specific duration and is  
82 assumed to be steady during the considered month, season, or year. Kjeldsen et al. (2000)  
83 extended the steady threshold concept to the variable method, employing seasonal, monthly  
84 and daily streamflows.

85 Therefore, there is a growing need to integrate drought severity duration, and frequency  
86 based on multivariate theory (Bonaccorso et al., 2003; Gonz´alez and Vald´es, 2003; Mishra  
87 et al., 2009; Song and Singh, 2010a, b; De Michele et al., 2013). Thus, based on the typical  
88 drought characteristics (water deficit and duration) and threshold levels, this study developed  
89 quantitative relations between drought parameters, namely severity, duration, and frequency,  
90 and used them to plot drought iso-severity curves for certain return periods and durations.  
91 This study quantified the streamflow drought severity, which is closely related to hydrological  
92 and socioeconomic droughts, using fixed, monthly, daily, and desired yield threshold levels.  
93 Furthermore, this study proposed a streamflow drought severity-duration-frequency (SDF)  
94 curve using traditional frequency analyses. This framework was applied to the Seomjin River  
95 basin in South Korea.

96

## 97 **2. Methodology**

### 98 **2.1 Procedure**

99 This study consists of five steps, as shown in Fig. 1. Step 1 is to determine the threshold  
100 levels for the fixed, monthly, daily, and desired yield for water use. The threshold selection  
101 description is shown in section 2.3. Step 2 is to calculate the severities (total water deficits)  
102 and durations for all the drought events at the four threshold levels. The methodology to  
103 derive the severity and the duration is shown in section 2.2. Step 3 is to derive the annual  
104 maxima of severity and duration and to identify the best-fit probability distribution functions  
105 using L-moment ratio diagrams (Hosking and Wallis, 1997). The calculation procedure is  
106 shown in section 2.4 using related equations and descriptions. Step 4 is to calculate the  
107 streamflow drought severities using the selected probability distribution with best-fit

108 parameters and to develop SDF curves. This is described in Section 2.5. Step 5 is to develop  
109 the duration-frequency curves of the four threshold levels using an appropriate probability  
110 distribution.

111

112 Fig. 1

113

## 114 2.2. Streamflow drought severity

115 In temperate regions where the runoff values are typically larger than zero, the most  
116 widely used method to estimate a hydrological drought is the threshold level approach  
117 (Yevjevich, 1967; Fleig et al., 2006; Tallaksen et al., 2009; Van Loon and Van Lanen, 2012).  
118 The streamflow drought severity with threshold level method has the following advantages  
119 over the standardized precipitation index (SPI) in meteorology (Yoo et al., 2008) and the  
120 Palmer drought severity index (PDSI) in meteorology and agriculture (Dalezios et al., 2000):  
121 1) no a priori knowledge of probability distributions is required and 2) drought characteristics,  
122 such as frequency, duration, and severity, are directly determined if the threshold is set by  
123 sectors impacted by the drought.

124 A sequence of drought events can be obtained using the streamflow and threshold levels.  
125 Each drought event is characterized by its duration  $D_i$ , deficit volume (or severity)  $S_i$ , and  
126 time of occurrence  $T_i$  as shown by the definition sketch in Fig. 2. With a prolonged dry  
127 period, the long drought spell is divided into several minor drought events. Because these  
128 droughts are mutually dependent, Tallaksen et al. (1997) proposed that an independent  
129 sequence of drought events must be described using some type of pooling as described below.

130

131 Fig. 2

132

133 If the ‘inter-event’ time  $t_i$  between two droughts of duration  $d_i$  and  $d_{i+1}$  and severity  $s_i$   
134 and  $s_{i+1}$ , respectively, are less than the predefined critical duration  $t_c$  and the pre-allowed  
135 inter-event excess volume  $z_c$ , the mutually dependent drought events are pooled to form a  
136 drought event as (Zelenhasic and Salvai, 1987; Tallaksen et al., 1997)

$$137 \quad d_{pool} = d_i + d_{i+1} + t_c$$

$$138 \quad s_{pool} = s_i + s_{i+1} - z_c. \quad (1)$$

139 This study assumed  $t_c = 3$  days and  $z_c = 10\%$  of  $d_i$  or  $d_{i+1}$  for simplicity. These numbers  
140 will be studied in the future.

141

### 2.3 Threshold selection

The threshold might be fixed or vary over the course of a year. A threshold is regarded as fixed if a constant value is used for the whole series, and a variable threshold is a value that varies over the year, using monthly and daily variable levels (Hisdal and Tallaksen, 2003). If the threshold is derived from the flow duration curve (FDC), it implies that the whole streamflow record is used in its derivation. As shown in Fig. 3, obtained from the study area, fixed and monthly thresholds can be obtained from an FDC and twelve monthly FDCs based on the entire record period. The daily varying threshold can be derived using the antecedent 365 day streamflow.

The threshold choice is influenced by the study objective, region, and available data. In general, a percentile from the can be used as the threshold. Relatively low thresholds in the range of  $Q_{70}$  to  $Q_{95}$  are often used for perennial rivers (Kjeldsen et al., 2000). This study selected  $Q_{70}$  for the fixed threshold, considering the Korean hydrologic condition, namely, a monsoon climate.  $Q_{70}$  means a 70% flow of the FDC. Specifically, 70% is the percentage of time that the streamflow  $Q_{70}$  is exceeded. However, the threshold selection should be further analyzed because it is not clear that  $Q_{70}$  should be used as a representative threshold for rivers in a monsoon climate.

The time resolution, i.e., where to apply series of annual, monthly, or daily streamflows, depends on the hydrologic regime in the region of interest. In a temperate zone, a given year might include both severe droughts (seasonal droughts) and months with abundant streamflow, meaning that annual data would not often reveal severe droughts. Dry regions are more likely to experience droughts lasting for several years, i.e., multi-year droughts, which support the use of a monthly or annual time step. Hence, different time resolutions might lead to different results regarding the drought event selection. This study used daily streamflow data, and the time resolutions were selected from 30 days to 270 days because droughts in the region of interest have never been studied.

Fig. 3

The variable threshold approach is adapted to detect streamflow deviations for both high and low flow seasons. Lower than average flows during high flow seasons might be important for later drought development. However, periods with relatively low flow either during the high flow season, for example, caused by a delayed onset of a snow-melt flood, is not commonly considered a drought. Therefore, the events defined with the varying threshold should be called streamflow deficiencies or streamflow anomalies rather than streamflow

177 droughts (Hisdal et al., 2004). In contrast, the desired yield for sufficient water supply and  
 178 environmental instreamflow can be an effective way to identify a streamflow drought by  
 179 considering hydrological and socioeconomic demands because environmental instreamflow  
 180 has become important in recent years.

181

## 182 **2.4 Probability distribution function**

183 An L-moment diagram for various goodness-of-fit techniques was used to evaluate the  
 184 best probability distribution function for datasets in several recent studies (Hosking, 1990;  
 185 Chowdhury et al., 1991; Vogel and Fennessey, 1993; Hosking and Wallis, 1997). The L-  
 186 moment ratio diagram is a graph where the sample L-moment ratios, L-skewness ( $\tau_3$ ), and L-  
 187 kurtosis ( $\tau_4$ ) are plotted as a scatterplot and compared with the theoretical L-moment ratio  
 188 curves of candidate distributions. L-moment ratio diagrams were suggested as a useful  
 189 graphical tool for discriminating amongst candidate distributions for a dataset (Hosking and  
 190 Wallis, 1997). Two representations used to assist in the selection of statistical distributions are  
 191 the sample average and line of best fit, which can be plotted on the same graph to facilitate  
 192 selecting the best-fit distribution.

193 When plotting an L-moment ratio diagram, the relation between the parameters and the L-  
 194 moment ratios  $\tau_3$  and  $\tau_4$  for several distributions are required. In the case of a GEV  
 195 distribution, the three-parameter GEV distribution described by Stedinger et al. (1993) has the  
 196 following probability density function (PDF) and cumulative distribution function (CDF):

$$197 \quad f(x) = \frac{1}{\alpha} \left\{ 1 - \frac{\kappa}{\alpha} (x - \xi) \right\}^{1/\kappa - 1} \cdot \exp \left[ - \left\{ 1 - \frac{\kappa}{\alpha} (x - \xi) \right\}^{1/\kappa} \right] \quad \kappa \neq 0 \quad , \quad (3a)$$

$$198 \quad f(x) = \frac{1}{\alpha} \exp \left\{ - \frac{x - \xi}{\alpha} - \exp \left( - \frac{x - \xi}{\alpha} \right) \right\} \quad \kappa = 0 \quad , \quad (3b)$$

$$199 \quad F(x) = \exp \left[ - \left\{ 1 - \frac{\kappa}{\alpha} (x - \xi) \right\}^{1/\kappa} \right] \quad \kappa \neq 0 \quad , \quad (4a)$$

$$200 \quad F(x) = \exp \left\{ - \exp \left( - \frac{x - \xi}{\alpha} \right) \right\} \quad \kappa = 0 \quad , \quad (4b)$$

201 where  $\xi + \alpha/\kappa \leq x \leq \infty$  for  $\kappa < 0$ ,  $-\infty \leq x \leq \infty$  for  $\kappa = 0$ , and  $-\infty \leq x \leq \xi + \alpha/\kappa$   
 202 for  $\kappa > 0$ . Here,  $\xi$  is a location,  $\alpha$  is a scale, and  $\kappa$  is a shape parameter. For  $\kappa = 0$ , the  
 203 GEV distribution reduces to the classic Gumbel (EV1) distribution with  $\tau_3 = 0.17$ . Hosking  
 204 and Wallis (1997) provided more detailed information regarding the GEV distribution. The  
 205 relation between the parameters and  $\tau_3$  and  $\tau_4$  for the shape parameter's GEV distribution  
 206 can be obtained as follows (Hosking and Wallis, 1997):

207 
$$\tau_3 = \frac{2(1-3^{-\kappa})}{(1-2^{-\kappa})} - 3 \quad (5a)$$

208 
$$\tau_4 = \frac{5(1-4^{-\kappa}) - 10(1-3^{-\kappa}) + 6(1-2^{-\kappa})}{(1-2^{-\kappa})}. \quad (5b)$$

209

## 210 **2.5 Development of SDF relationships**

211 IDF or depth-duration-frequency (DDF) curves can be defined to “allow calculation of the  
 212 average design rainfall intensity (or depth) for a given exceedance probability over a range of  
 213 durations” and results from the rainfall frequency analysis (Okonkwo and Mbajiorgu, 2010).  
 214 Statistical frequency analyses, such as rainfall analyses, are frequently utilized for drought  
 215 events. This, however, cannot fully explain droughts without any consideration of severity  
 216 and duration, which has resulted in the development of the SDF curve. Thus, extreme drought  
 217 events can be specified by frequency, duration and either depth or mean intensity (i.e.,  
 218 severity). The frequency is usually described by its return period, which is defined as the  
 219 average interval of time within which the magnitude of the event is reached or exceeded once.  
 220 Because its magnitude is given by the total depth occurring in a particular duration, the SDF  
 221 relation can be derived. For the estimation of the return periods for drought events of a  
 222 particular depth and duration, the frequency distributions can be utilized (Dalezios et al.,  
 223 2000).

224

## 225 **3. Study region**

226 The Seomjin River basin is located in southwestern Korea (Fig. 4). The area and total  
 227 length of the Seomjin River are approximately 4,911.9 km<sup>2</sup> and 212.3 km, respectively. The  
 228 altitude range is rather large, spanning from approximately 0 to 1,646 m (Fig. 4). The climate  
 229 of South Korea is characterized by extreme seasonal variations. Winter is cold and dry under  
 230 the dominant influence of the Siberian air mass, whereas the summer is hot and humid, with  
 231 frequent heavy rainfall associated with the East Asian monsoon. In the Seomjin River basin,  
 232 the measured precipitation is mainly concentrated in summer, and the measured mean annual  
 233 precipitation varied from < 1,350 mm/yr<sup>-1</sup> (in the north region) to > 1,600 mmyr<sup>-1</sup> (in the  
 234 southeastern region) during the 1975-2012 observation period. In general, approximately 60%  
 235 of the annual precipitation occurs during the wet season (July through September) in South  
 236 Korea. This extreme seasonality in the precipitation causes periodic shortages of water during  
 237 the dry season (October through March) and flood damage during the wet season.

238

239 Fig. 4.

240

241 The administrative districts where the basin is located cover three provinces, four cities,  
242 and 11 counties (Namwon City, Jinan County, Imsil County, and Sunchang County in the  
243 Northern Jeolla Province; Suncheon City, Gwangyang City, Damyang County, Gokseong  
244 County, Gurye County, Hwasun County, Boseong County, and Jangheung County in the  
245 Southern Jeolla Province; and Handing County in the Southern Gyeongsang Province). Influx  
246 rates into the basin by province are 47% (Southern Jeolla Province), 44% (Northern Jeolla  
247 Province), and 9% (Southern Gyeongsang Province), and a total of 321,104 residents  
248 occupying 129,322 households live in these areas.

249 The land use consists of arable land (876.29 km<sup>2</sup>), forest land (3,400.61 km<sup>2</sup>), urban area  
250 (67.12 km<sup>2</sup>), and other land uses (567.86 km<sup>2</sup>). Additionally, 69.2% of the entire basin area  
251 (4,911.89 km<sup>2</sup>) is forest land. Major droughts occurred in the Southern Jeolla Province from  
252 1967 to 1968 and from 1994 to 1995. The Seomjin River basin had < 1,000 mm of  
253 precipitation on average in 1977, 1988, 1994, and 2008. Among these years, the annual  
254 precipitation in 1988 was only 782.7 mm (56.5%) of the annual average of 1,385.5 mm from  
255 1967 to 2008, representing a serious drought.

256

## 257 **4. Results**

### 258 **4.1. Determination of the threshold levels**

259 This study used four threshold levels. The fixed threshold level is  $Q_{70}$  of the FDC, which  
260 resulted from 37-year daily streamflows. The monthly thresholds are twelve  $Q_{70}$ s of monthly  
261 FDCs, which incorporated data of all the daily streamflows of January through December for  
262 the past 37 years. The daily threshold is  $Q_{70}$  of the FDCs, which resulted from the antecedent  
263 365 daily streamflows. Thus, the daily threshold level smoothly varies every day. The desired  
264 yield threshold for a sufficient water supply and environmental instreamflow was determined  
265 by the Korean central government. This is related to social and economic droughts because it  
266 associates the supply and demand of a number of economic goods and environmental safety.  
267 The desired yield threshold differed considerably from the other levels and represented more  
268 realistic conditions because the desired yield is equivalent to the planned water supply.

269 The four calculated thresholds are presented in Fig. 5, and the specific monthly-averaged  
270 values are listed in Table 1. The average levels were 1.9, 2.5, 2.8, and 13.8 m<sup>3</sup>s<sup>-1</sup> for the fixed,  
271 monthly, daily and desired yields, respectively. The daily threshold levels fluctuated  
272 significantly because of the natural streamflow variations for the antecedent 365 days and  
273 were the largest of the four threshold levels because a summer period (June, July, and August)  
274 was considered. The desired yield level was larger than the fixed, monthly, and daily  
275 thresholds. This phenomenon occurred during the winter in Korea, and, as a result, both the

276 water demand and natural runoff during the winter (December, January, and February) were  
277 quite small. However, the threshold levels for the daily, monthly, and desired yields during  
278 the summer were much higher than those during the other seasons. The threshold levels for  
279 the desired yield during May and June were much higher than the levels for the other  
280 thresholds because the agricultural water demand was the highest in this season.

281

282 Fig. 5

283

284 Table 1

285

286

## 287 **4.2 Calculations of streamflow drought severity and duration**

288 The durations and severities for all the streamflow drought events were calculated based  
289 on the streamflow drought concept and threshold levels. The annual maxima values of  
290 duration and severity are shown in Fig. 6, and the summarized values are listed in Table 2.  
291 The maximum durations from the desired yield threshold approach were considerably higher  
292 than those from the other thresholds because the desired yields were highest during June and  
293 July because of agricultural water use. Similar to the results for the drought duration, the  
294 severities showed much higher values.

295

296 Table 2

297

298 Fig. 6

299

300 To compare the differences between the four threshold levels, the correlation coefficients  
301 were calculated as shown in Table 3. Similar trends were observed for the monthly and daily  
302 threshold levels. However, the durations and severities from the desired yield threshold level  
303 were completely different from those for the fixed, monthly and daily levels. Therefore, it is  
304 possible that the drought identification techniques based on general threshold levels cannot  
305 reflect the socioeconomic drought in terms of water supply and demand. Therefore, two-way  
306 approaches categorized by the time periods (fixed, monthly, and daily) for hydrological  
307 drought and desired yield threshold for socioeconomic drought should be separately included  
308 for specific drought characteristics identification.

309

310 Table 3

311

312 **4.3 Identification of the probability distribution function**

313 The L-moment diagrams of various goodness-of-fit techniques were used to evaluate the  
314 best probability distribution function for the datasets. To develop a streamflow drought SDF  
315 curve, the proper probability distribution function should be determined based on the  
316 statistical results as described in Section 2.4.

317 The L-moment ratio diagrams were derived for the four threshold approaches and are  
318 displayed in Fig. 7. Of the distribution models tested, three parameter distributions such as the  
319 Pearson type 3 (PT3), generalized normal (GNO), and generalized extreme value (GEV)  
320 distributions appeared to be consistent with their datasets. In the frequency analysis  
321 addressing extreme values, distributions that use more than three parameters are required to  
322 express the upper tail. The PT3, GNO, and GEV distributions can be applied in this study. As  
323 shown in Fig. 7, this study selected the GEV distribution for a representative probability  
324 distribution because most observations are appropriate for the GEV. This corresponds to  
325 Dalezios et al. (2000) for the PDSI and Yoo et al. (2008) for the SPI.

326

327 Fig. 7

328

329 **4.4 Development of SDF curves**

330 Streamflow drought SDF curves were developed using the derived probability distribution  
331 functions as shown in Fig. 8. For these plots, 10-, 20-, 50-, 80-, and 100-year-frequency  
332 severities were calculated at 30-, 60-, 90-, 120-, 150-, 180-, 210-, and 270-day durations.  
333 Because the amount of available data only corresponds to 37 years, we calculated up to a 100-  
334 year frequency. However, SDF curves from fixed, daily, and monthly thresholds were  
335 calculated using comparatively short durations because the annual maximum durations vary  
336 from 30 to 96 days. The SDF described the streamflow drought severities with respect to  
337 durations and frequencies. The severity increases with increasing frequency and duration. For  
338 a specific description, Table 4 compares all of the severities to specific frequencies and  
339 durations for the desired yield threshold. As the duration becomes larger, the difference ratio  
340 between return periods becomes much larger. Therefore, because the streamflow drought  
341 severity should be more crucial when the drought continues for a longer period, the frequency  
342 to long-drought should be approached with caution.

343

344 Fig. 8

345

346 Table 4

347

#### 348 **4.5 Development of duration-frequency curve**

349 The drought can be characterized by a deficit volume using the threshold levels. However,  
350 using the deficit volume is not sufficient to explain the extreme droughts. Thus, analyzing  
351 streamflow drought durations can be another useful tool in identifying the drought event.  
352 Therefore, occurrence probabilities of various duration events were also estimated using a  
353 general frequency analysis. As a result, using the GEV distribution previously shown in Fig. 7,  
354 duration-frequency curves for four threshold levels were developed as shown in Fig. 9. For  
355 these plots, 2-, 3-, 5-, 10-, 20-, 30-, 50-, 70-, 80-, and 100-year-frequency severities were  
356 calculated. Similar to the SDF curves, the durations for the desired yield were much higher  
357 than those for the other three thresholds.

358

359 Fig. 9

360

#### 361 **5. Conclusions**

362 This study developed a useful concept to describe the characteristics of streamflow  
363 droughts using frequency analyses. SDF curves for streamflow droughts were developed to  
364 quantify a specific volume based on a specific duration and frequency. In addition, duration-  
365 frequency curves were used to derive the relationship between drought duration and  
366 frequency. This study used severity, which represents the total water deficit for specific  
367 durations. Using the L-moment diagram method, the GEV was selected for the best-fit  
368 probability distribution. As a result, SDF curves were derived to identify the relationship  
369 among streamflow drought severity, duration and frequency. The severities increased with  
370 increasing duration and frequency. However, these values were quite different because the  
371 four threshold level approaches defined the streamflow drought differently.

372 The fixed threshold level in this study is the 70<sup>th</sup> percentile value ( $Q_{70}$ ) of FDC, which is  
373 compiled using all the available daily streamflows, and the monthly threshold level is the  
374 monthly-varying  $Q_{70}$ s of each month's FDC. The daily variable threshold is the  $Q_{70}$  of the  
375 FDC obtained from the antecedent 365 daily streamflows. The desired yield threshold  
376 determined by the central government consists of domestic, industrial, and agricultural water  
377 uses as well as environmental instreamflow. As a result, the desired yield threshold can  
378 identify streamflow droughts using the total water deficit to the hydrological and  
379 socioeconomic targets, while the fixed, monthly, and daily streamflow thresholds derive the  
380 deficiencies or anomalies from the average of historical streamflow.

381 Streamflow drought SDF curves developed in this study can potentially be exploited to  
382 quantify the water deficit for natural streams as well as reservoirs. In addition, these will be  
383 extended to allow for regional frequency analyses, which can estimate streamflow drought

384 severity at ungauged sites. Therefore, it can be an effective tool to identify any streamflow  
385 droughts using severity, duration, and frequency.

386

387

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390

391

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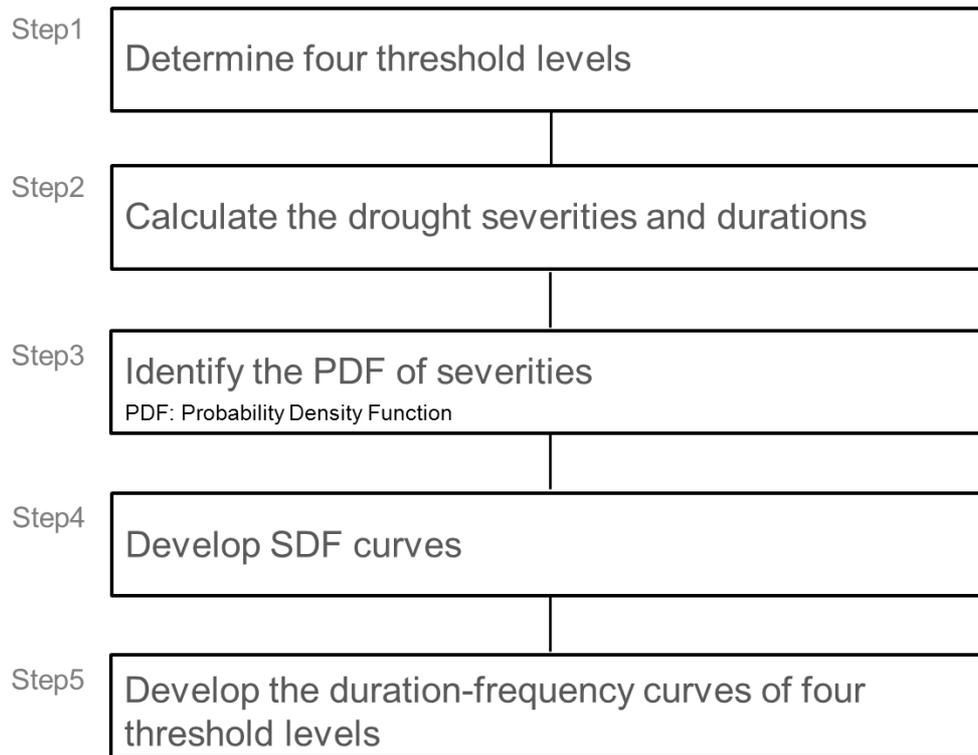
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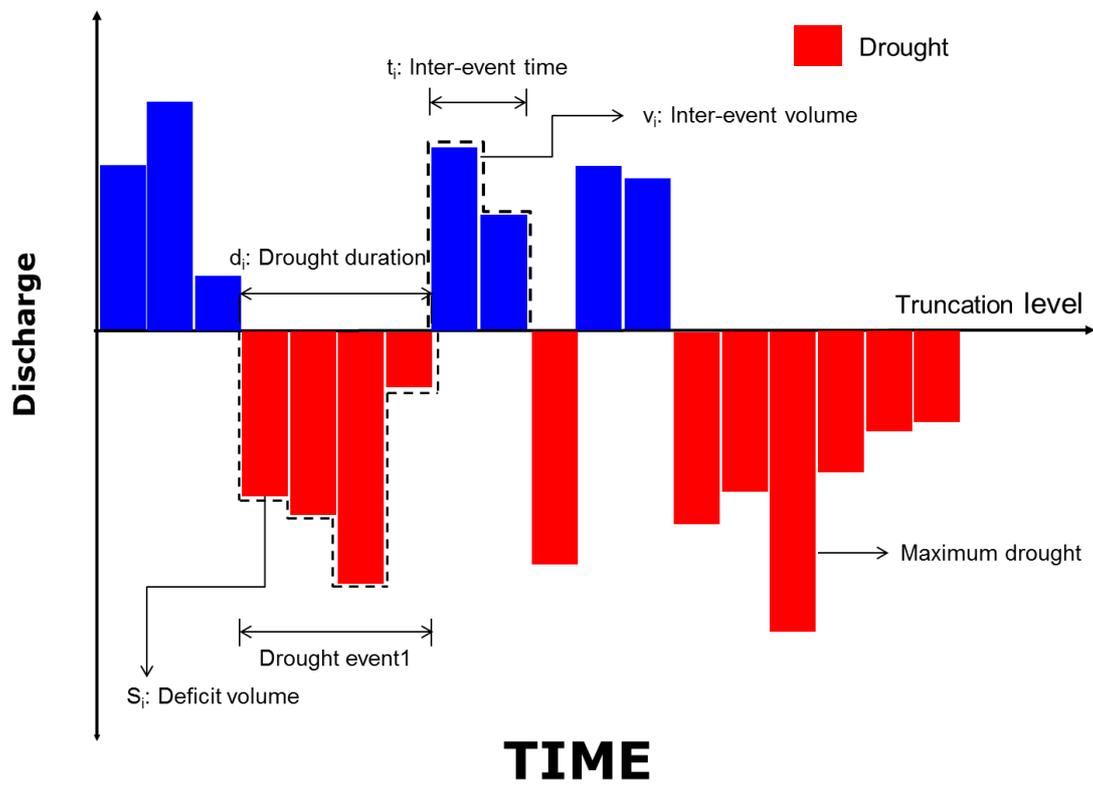


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505 Fig. 1. Procedure used in this study.

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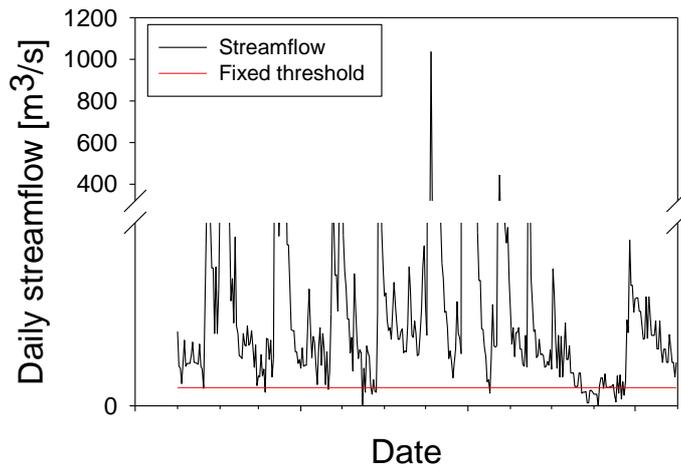


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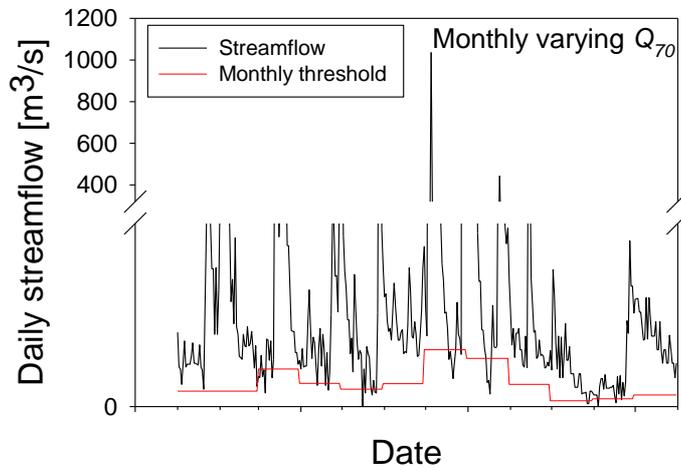
509 Fig. 2. A definition sketch of a general drought event.

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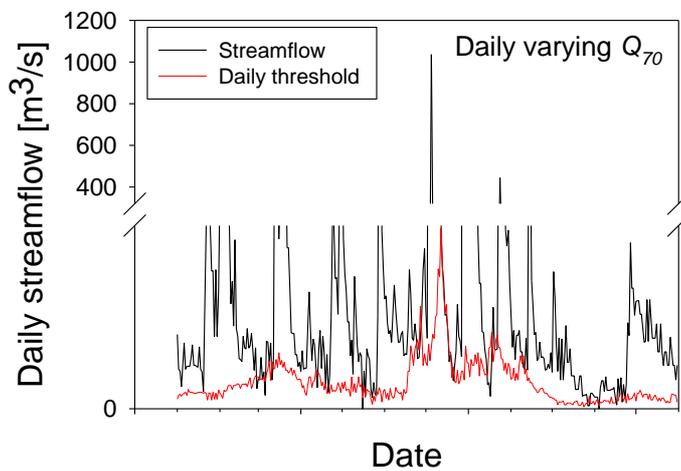
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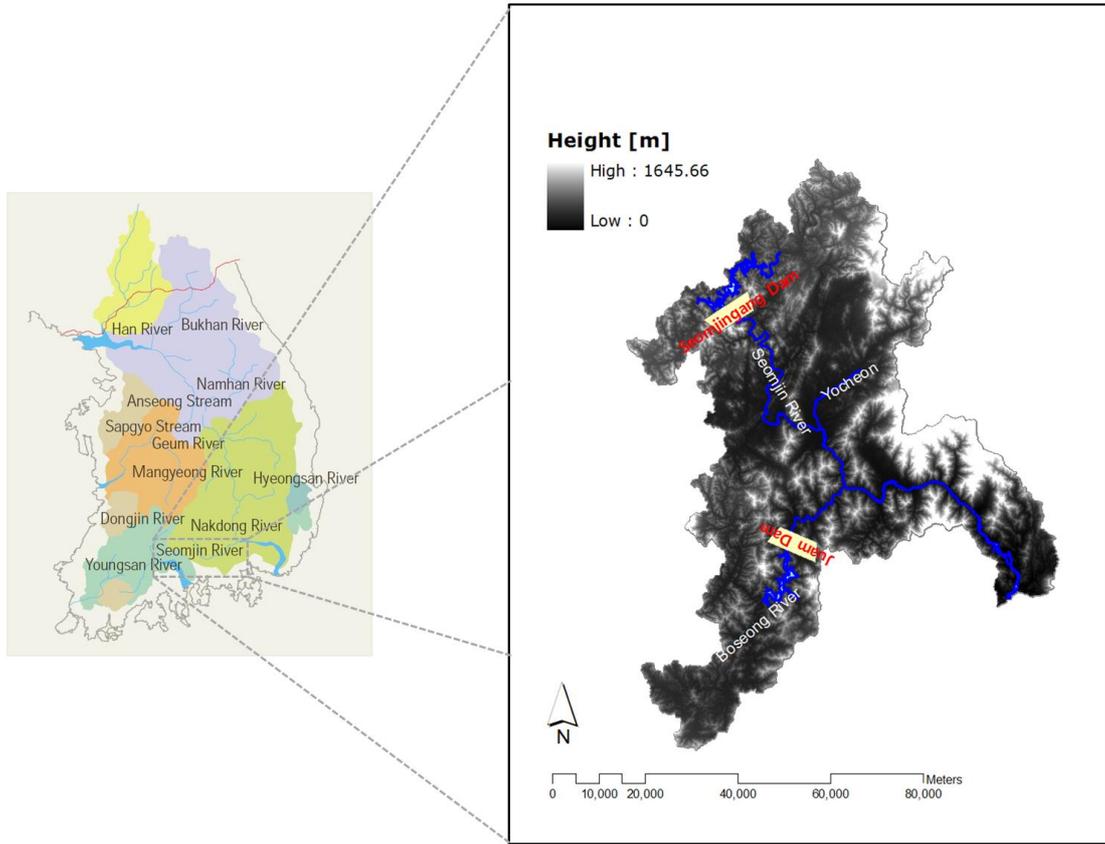


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515 Fig. 3. Examples of threshold levels: fixed (top), monthly varying (middle), and daily varying

516 (bottom).

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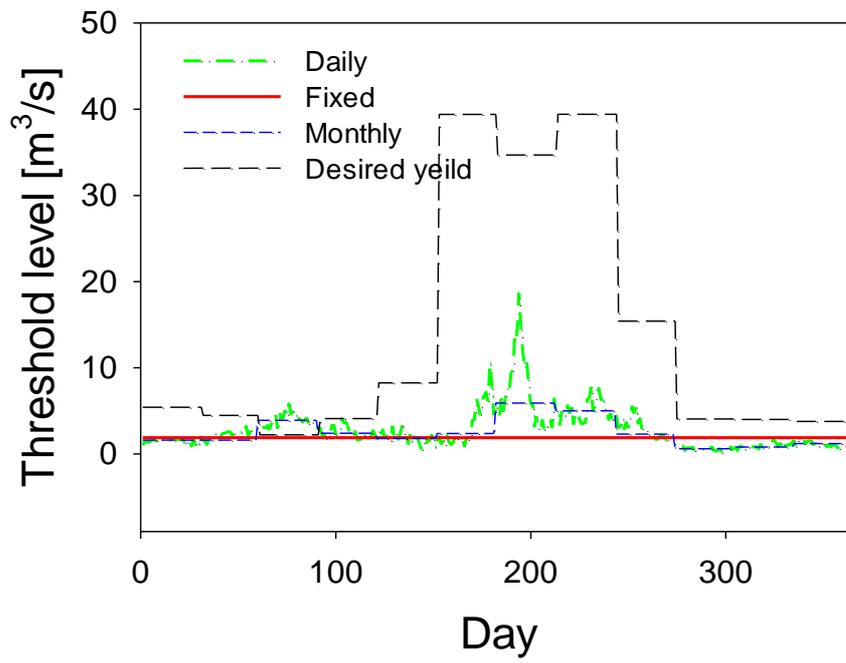


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519 Fig. 4. Location of the selected river basin, including elevation and rivers.

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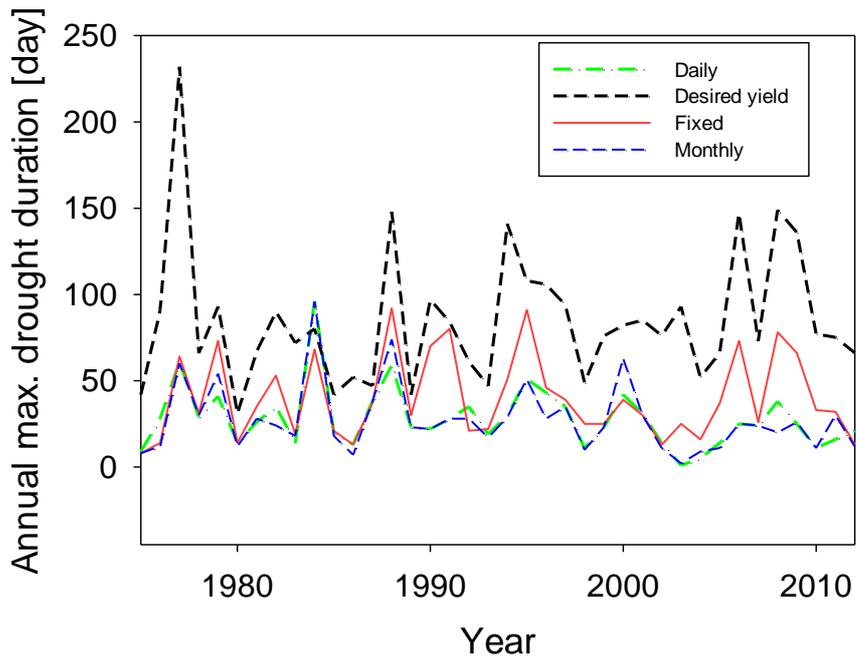


522

523 Fig. 5. Comparison of the four threshold levels used in this study.

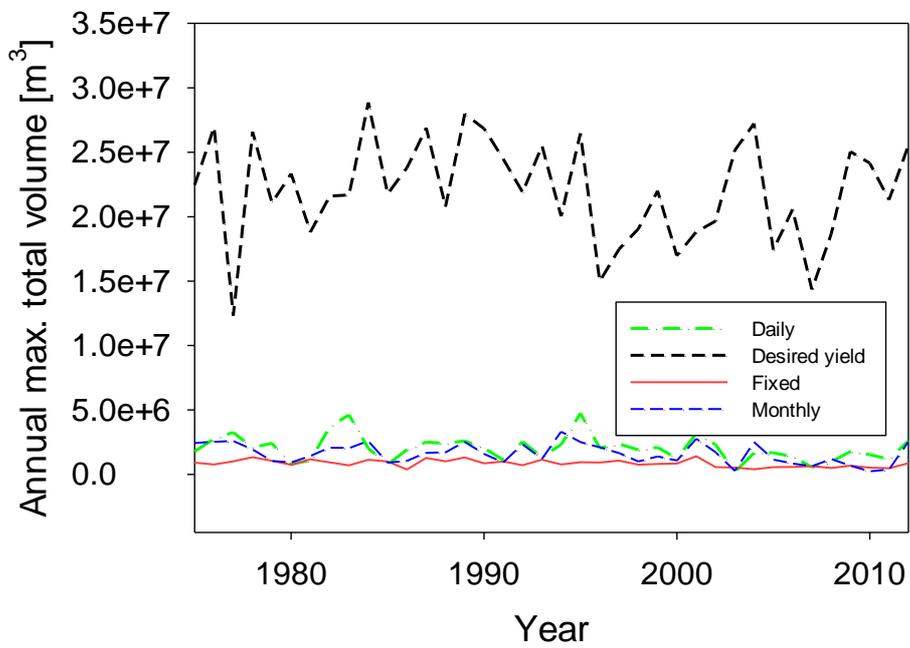
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526

527 (a) Drought duration.

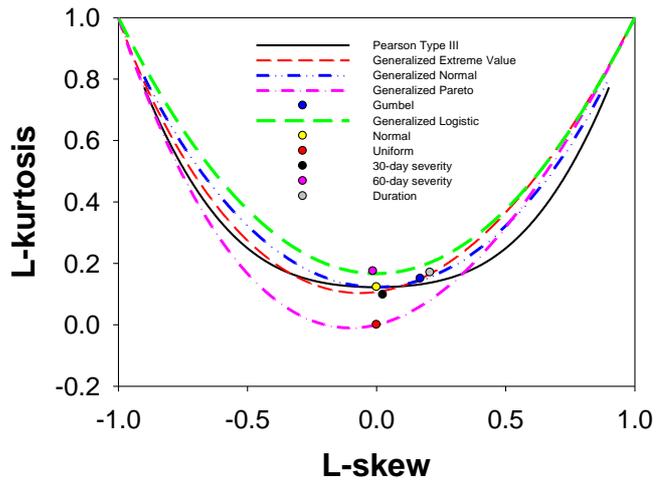


528

529 (b) Total water deficit volume (drought severity).

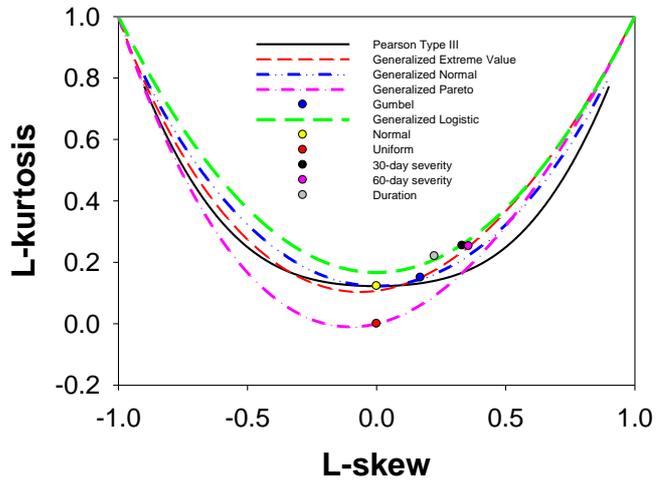
530 Fig. 6. Time series of the annual maxima values of duration and severity.

531



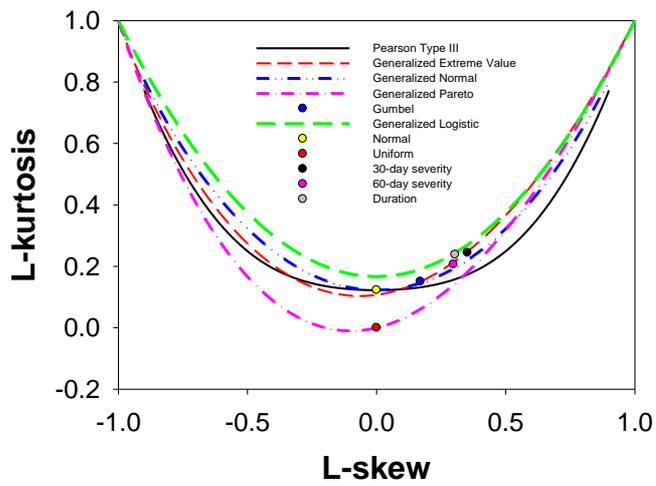
532

533 (a) Fixed.



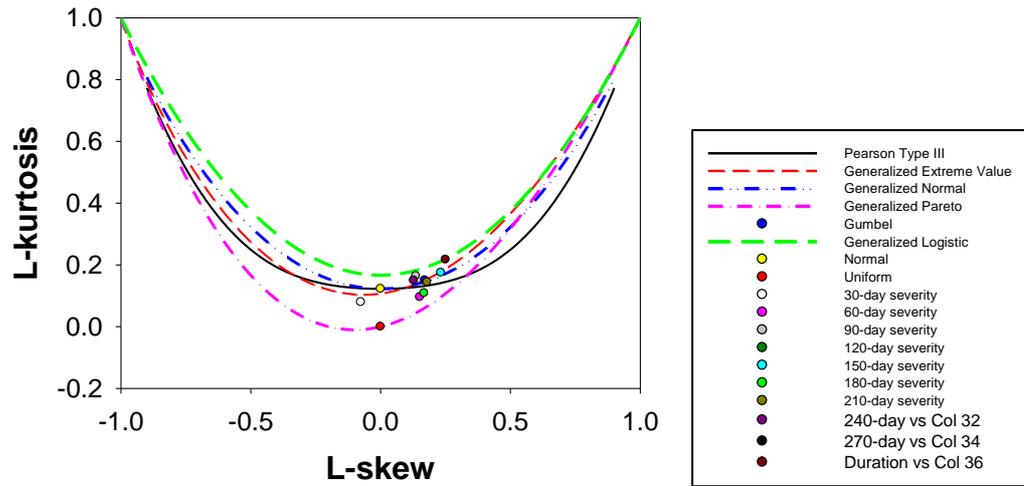
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535 (b) Daily.



536

537 (c) Monthly.

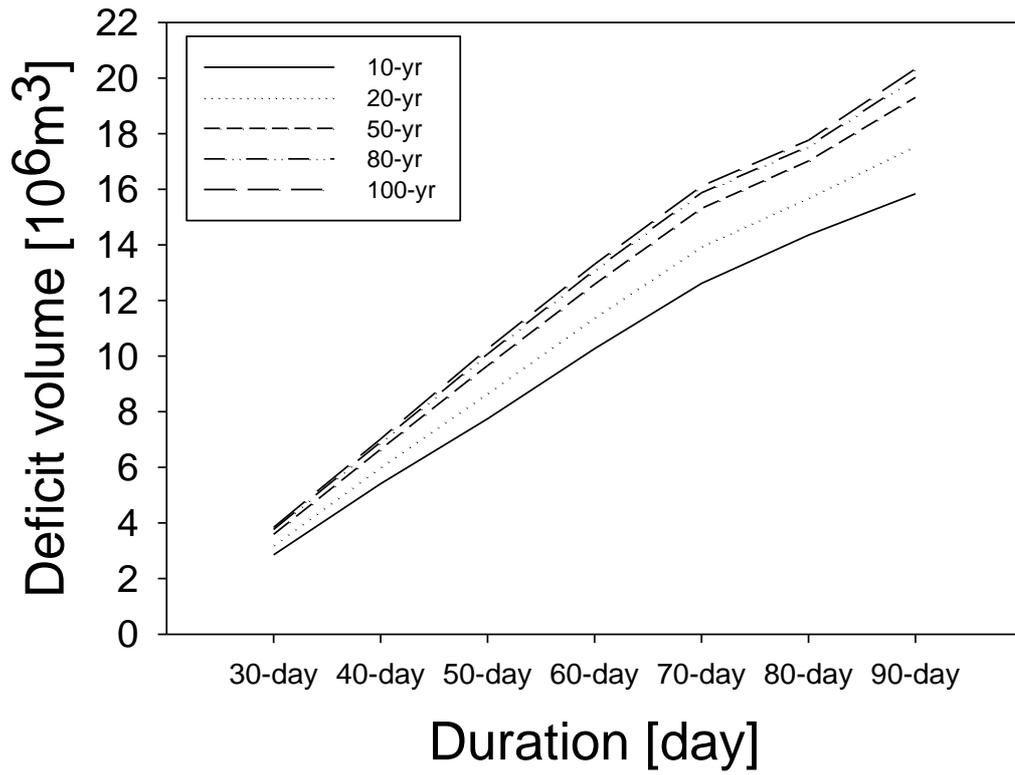


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539 (d) Desired yield.

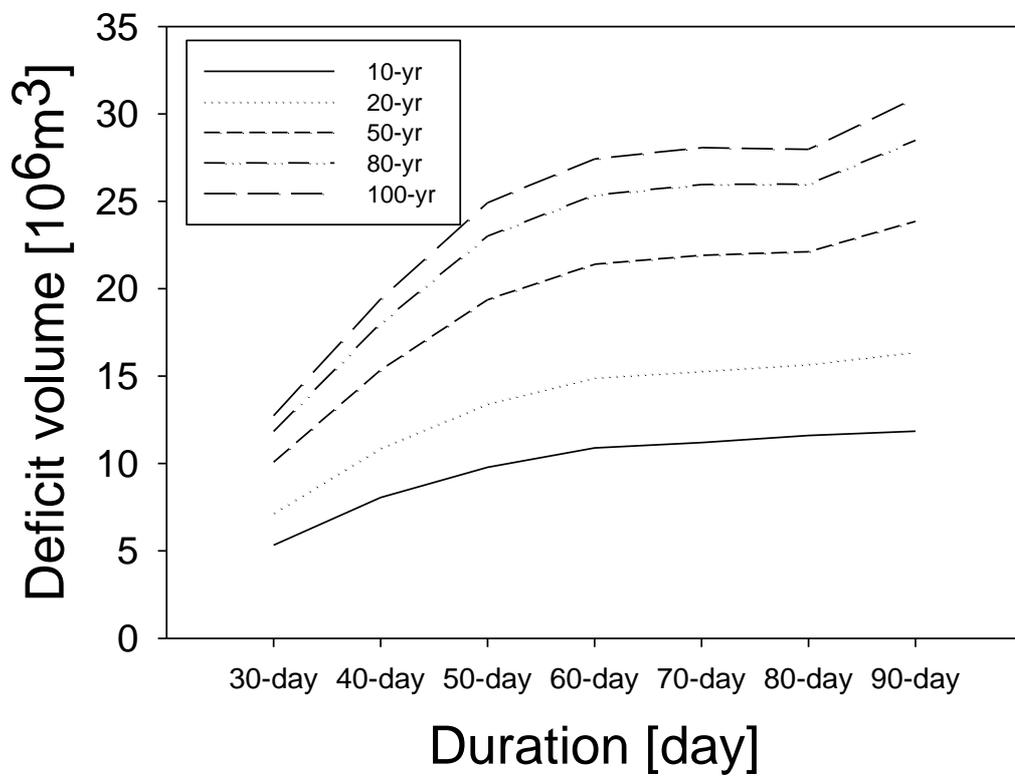
540 Fig. 7. L-moment diagram for probability distribution identification.

541



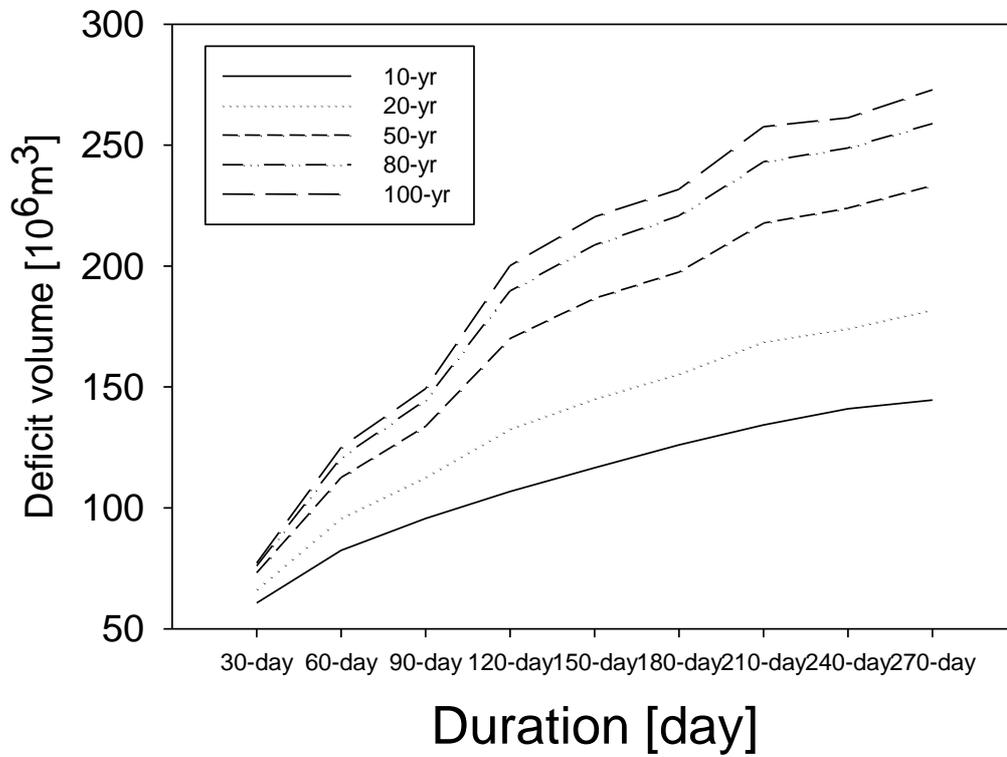
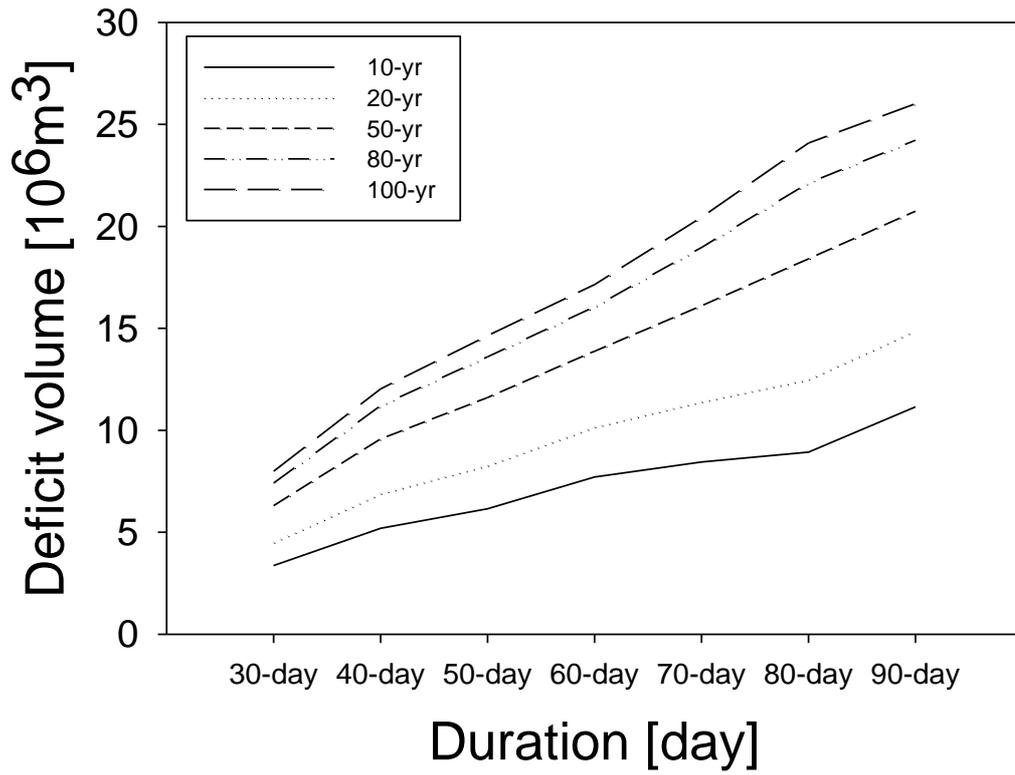
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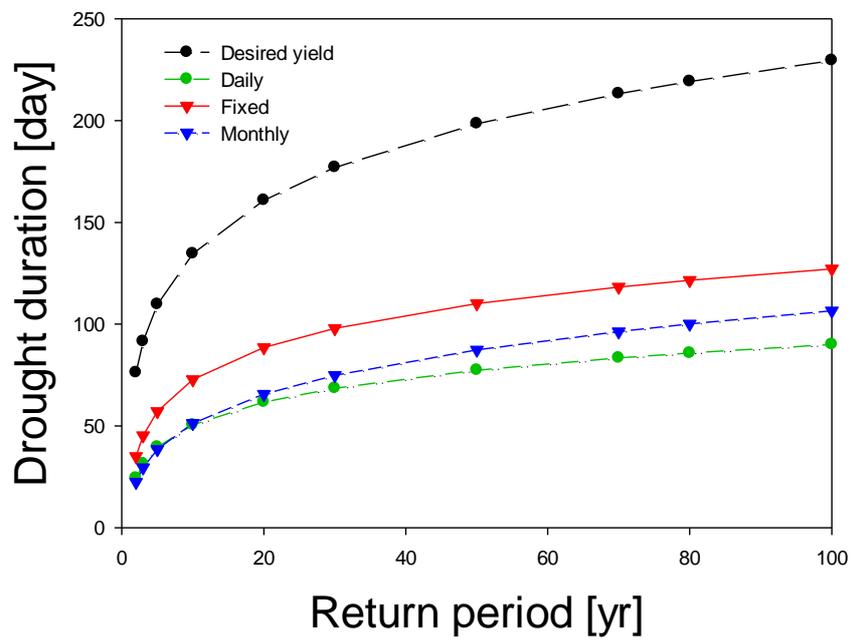
543 (a) Fixed.



544

545 (b) Daily.





553

554 Fig. 9. Duration-frequency curves of the four threshold level approaches in the Seomjin river

555 basin.

556

557

558 Table 1. Monthly average of the four threshold levels.

	Threshold level [m <sup>3</sup> s <sup>-1</sup> ]			
	Fixed	Monthly	Daily	Desired yield
Jan	1.9	1.6	1.5	5.4
Feb	1.9	1.6	2.4	4.5
Mar	1.9	3.9	3.9	2.2
Apr	1.9	2.4	2.5	4.1
May	1.9	1.8	1.9	8.2
Jun	1.9	2.4	3.4	39.4
Jul	1.9	5.9	7.1	34.7
Aug	1.9	5.0	5.1	39.4
Sep	1.9	2.3	2.9	15.4
Oct	1.9	0.6	0.7	4.0
Nov	1.9	0.8	0.9	4.0
Dec	1.9	1.2	1.2	3.8

559

560 Table 2. Summary of the four threshold approaches.

Threshold level method	Maximum Duration (days)	Maximum Severity (m <sup>3</sup> )
Fixed	92	9,304,762
Monthly	96	10,774,642
Daily	96	18,457,943
Desired yield	232	285,854,400

561

562 Table 3. Correlations between durations and severities for the four threshold levels.

Duration				
	Fixed	Monthly	Daily	Desired yield
Fixed	1			
Monthly	0.632	1		
Daily	0.632	0.923	1	
Desired yield	0.677	0.420	0.475	1
Severity				
	Fixed	Monthly	Daily	Desired yield
Fixed	1			
Monthly	0.441	1		
Daily	0.414	0.853	1	
Desired yield	0.281	0.551	0.599	1

563

564

565 Table 4. Severity-duration-frequency of the desired yield in the Seomjin river basin.

Duration [day]	Return period [yr]				
	10	20	50	80	100
30	60.7	66.4	73.1	75.9	77.2
60	82.4	95.9	112.5	120.8	124.9
90	95.6	112.8	133.7	144.6	149.3
120	106.8	132.7	170.0	189.7	200.1
150	116.6	145.2	186.6	208.7	220.3
180	126.0	155.5	197.5	220.8	231.7
210	134.3	168.7	217.7	243.1	257.6
240	141.0	174.2	223.9	248.8	261.3
270	144.6	182.0	233.3	258.9	272.9

566