

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 70th 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 (τ_3), and L-
 187 kurtosis (τ_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 τ_3 and τ_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, ξ is a location, α is a scale, and κ 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 τ_3 and τ_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 km² 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⁻¹ (in the north region) to > 1,600 mmyr⁻¹ (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 km²), forest land (3,400.61 km²), urban area
250 (67.12 km²), and other land uses (567.86 km²). Additionally, 69.2% of the entire basin area
251 (4,911.89 km²) 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 m³s⁻¹ 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 70th 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

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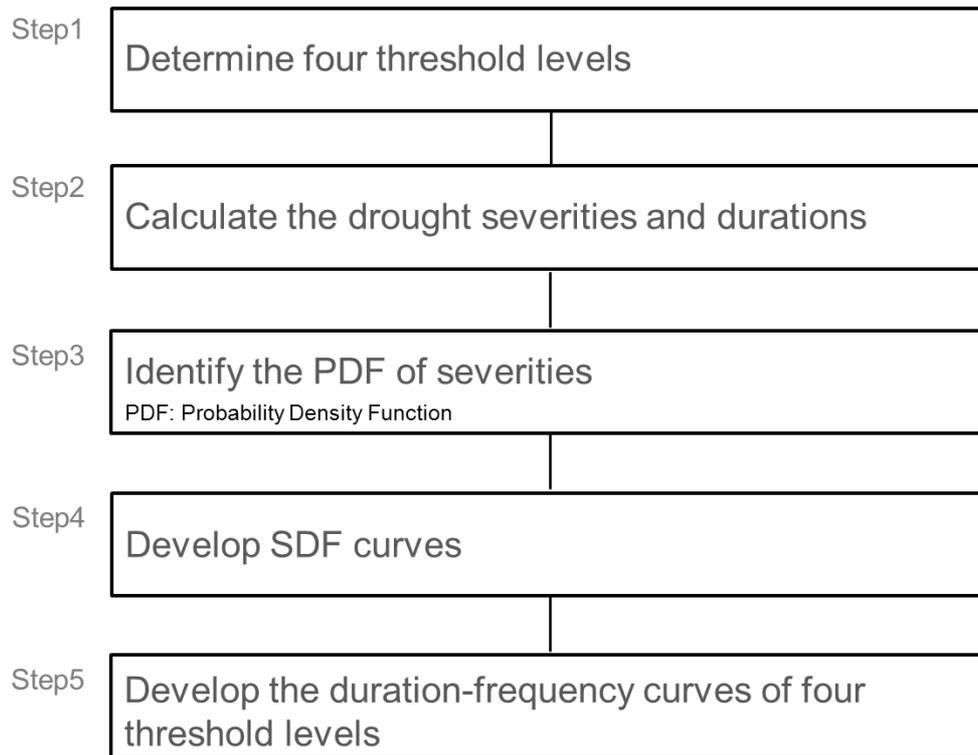
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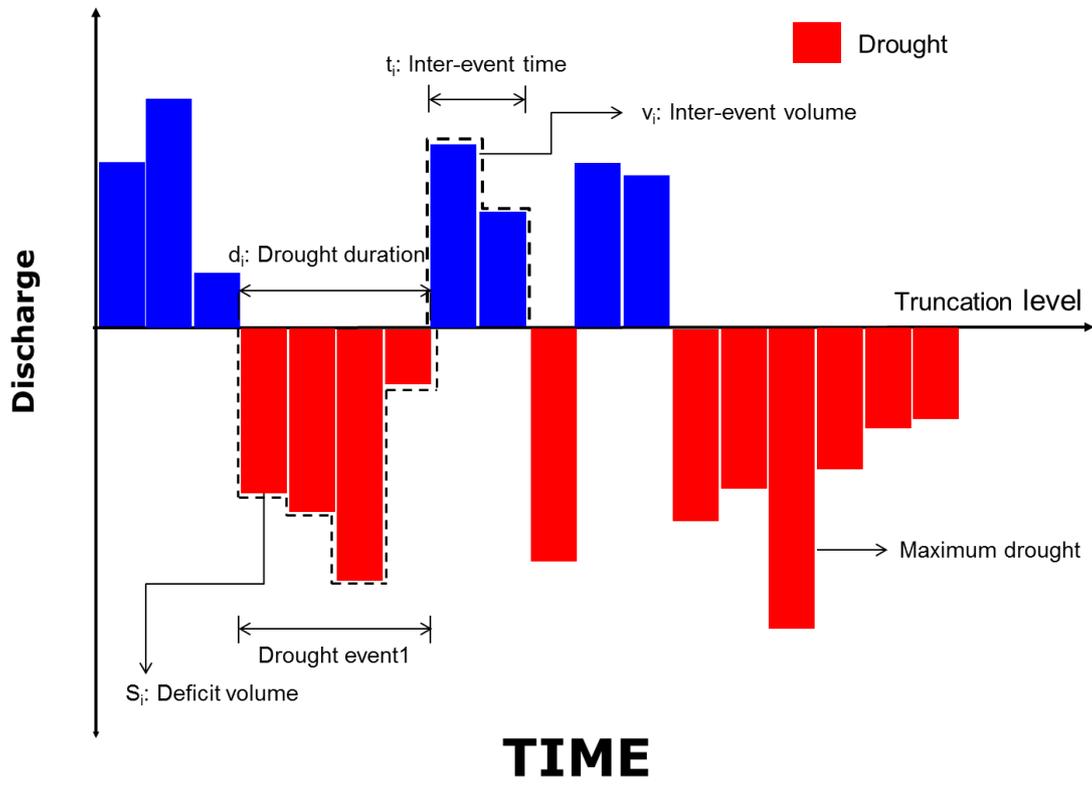


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505 Fig. 1. Procedure used in this study.

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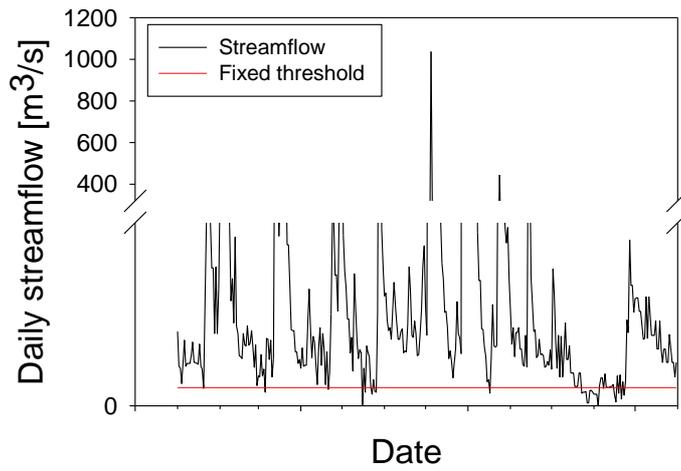


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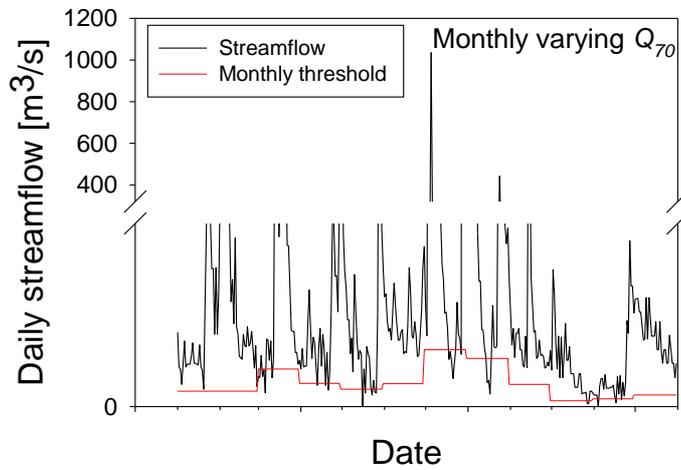
509 Fig. 2. A definition sketch of a general drought event.

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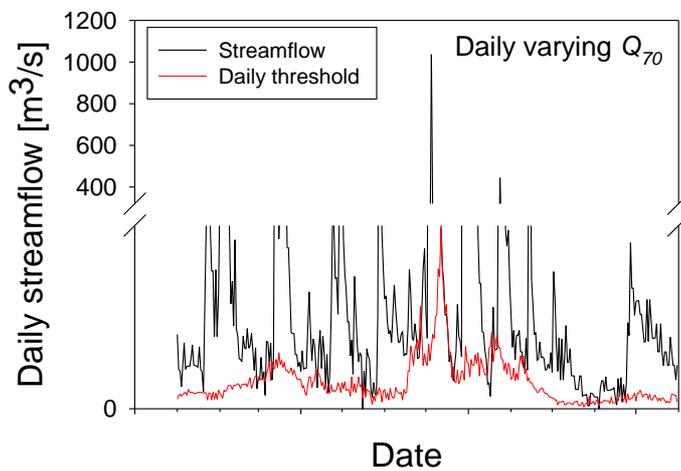
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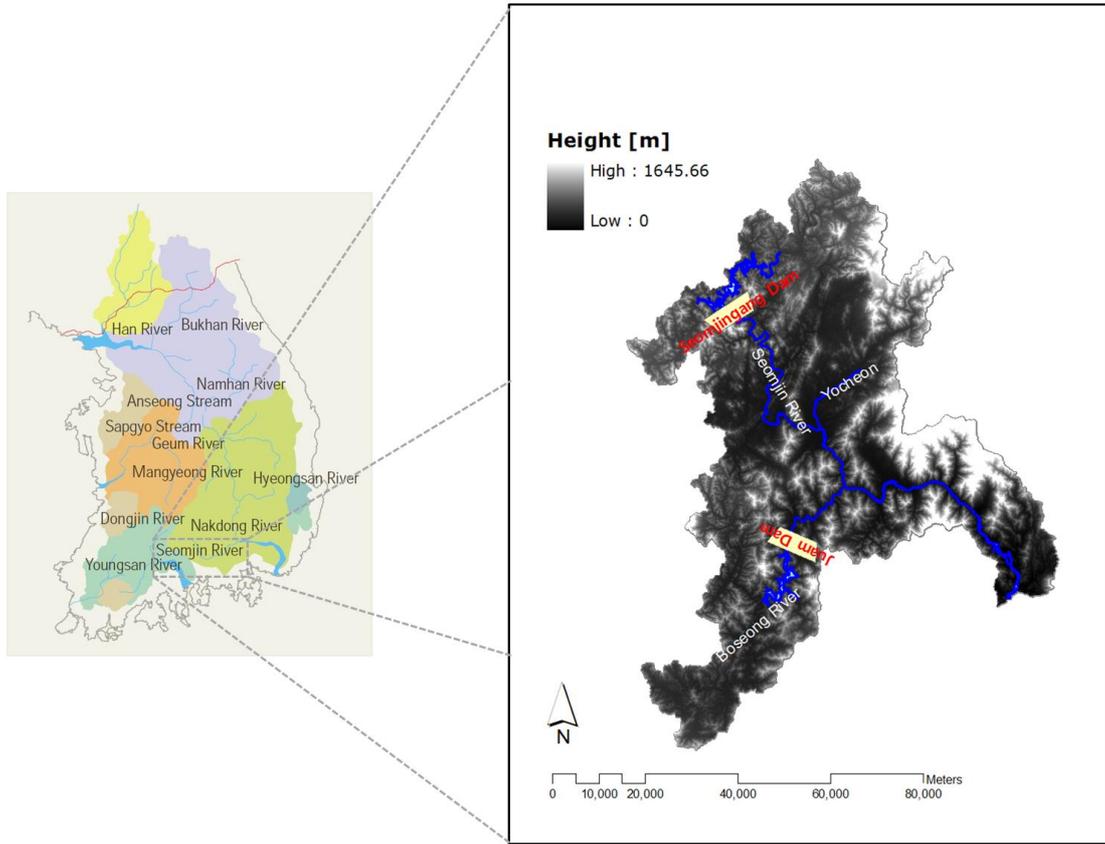


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515 Fig. 3. Examples of threshold levels: fixed (top), monthly varying (middle), and daily varying

516 (bottom).

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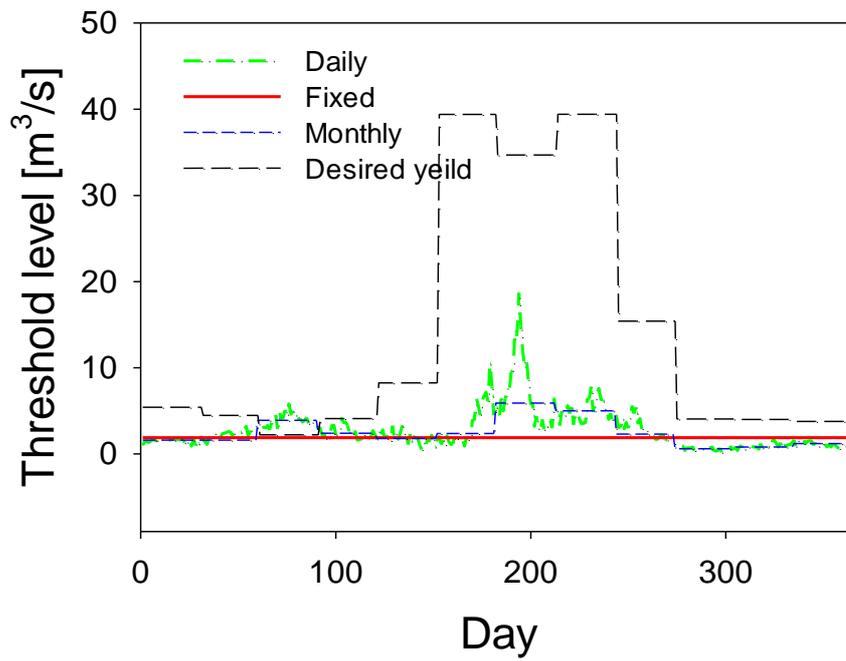


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519 Fig. 4. Location of the selected river basin, including elevation and rivers.

520

521

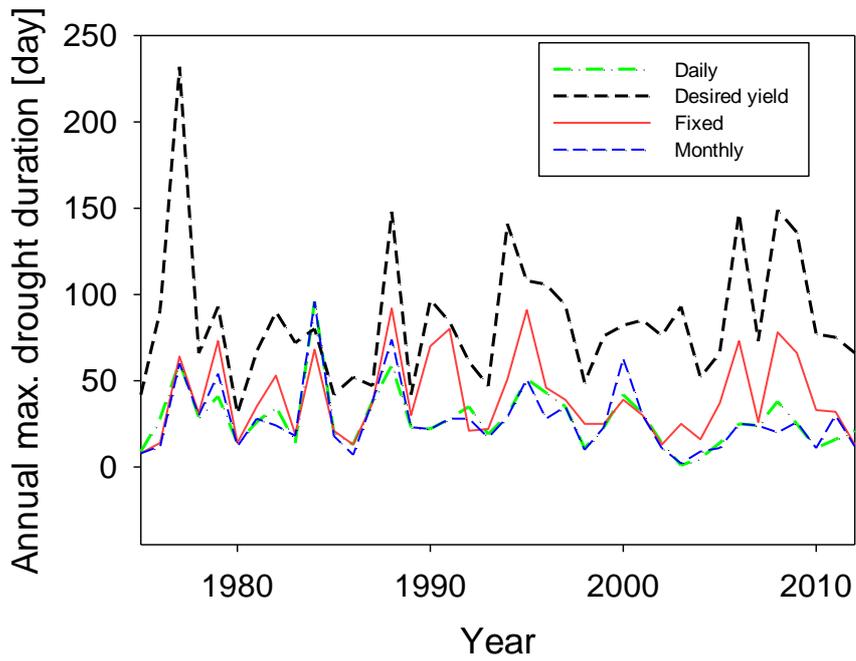


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523 Fig. 5. Comparison of the four threshold levels used in this study.

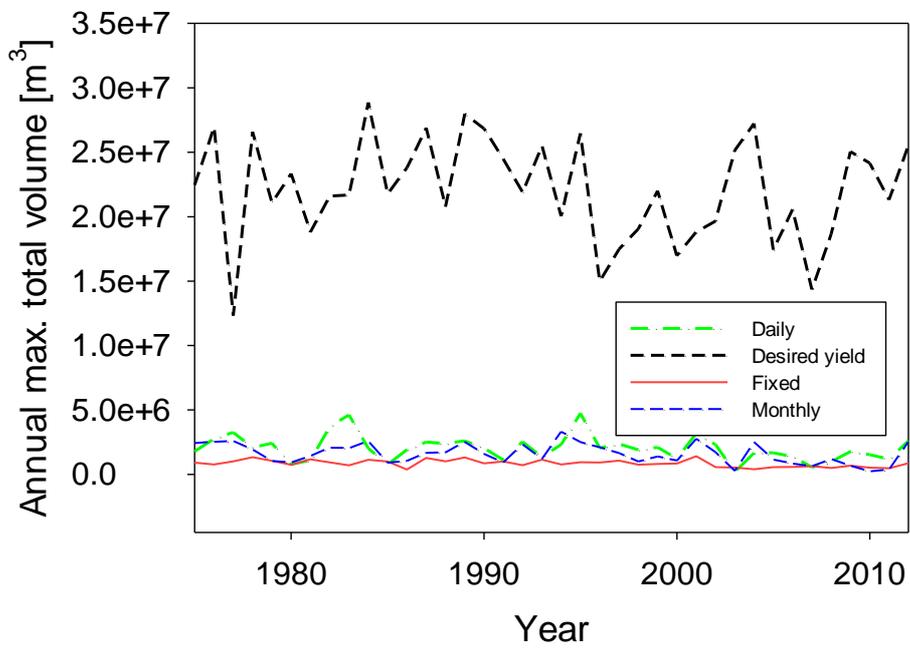
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527 (a) Drought duration.

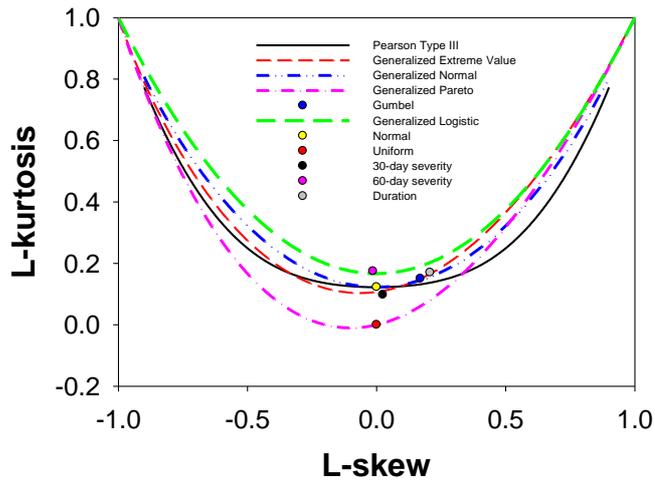


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529 (b) Total water deficit volume (drought severity).

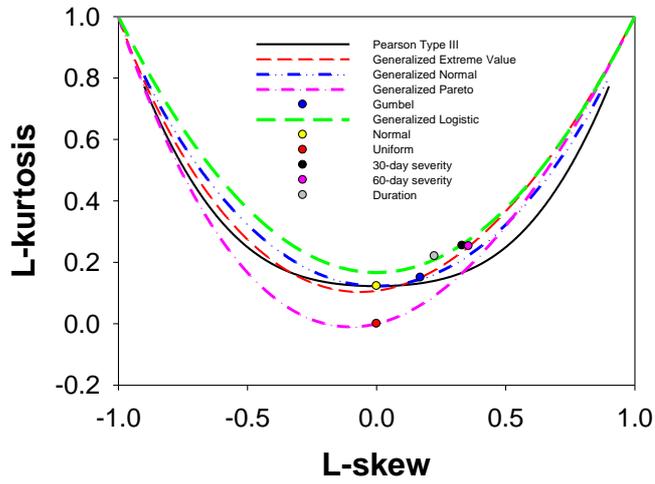
530 Fig. 6. Time series of the annual maxima values of duration and severity.

531



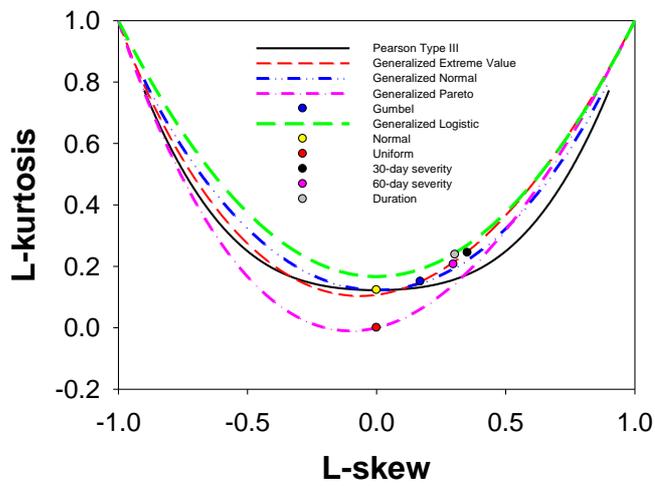
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533 (a) Fixed.



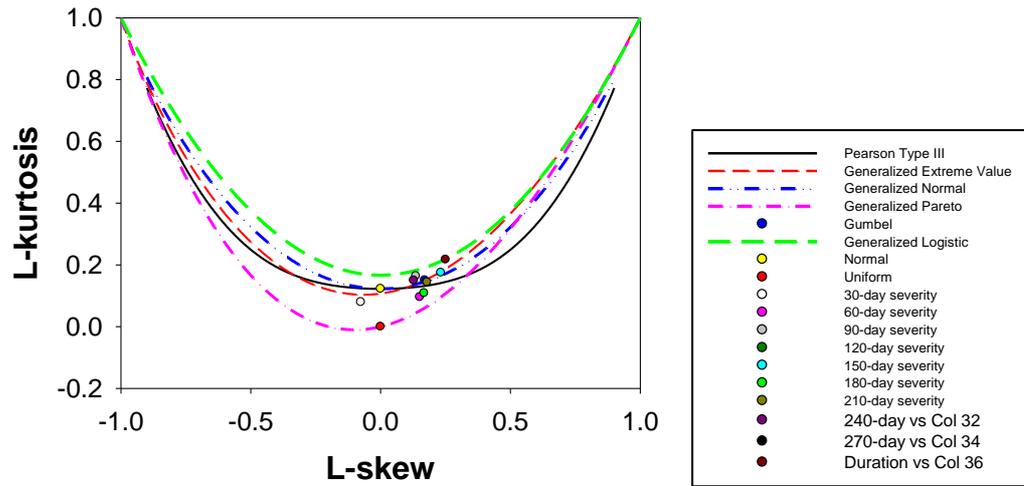
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535 (b) Daily.



536

537 (c) Monthly.

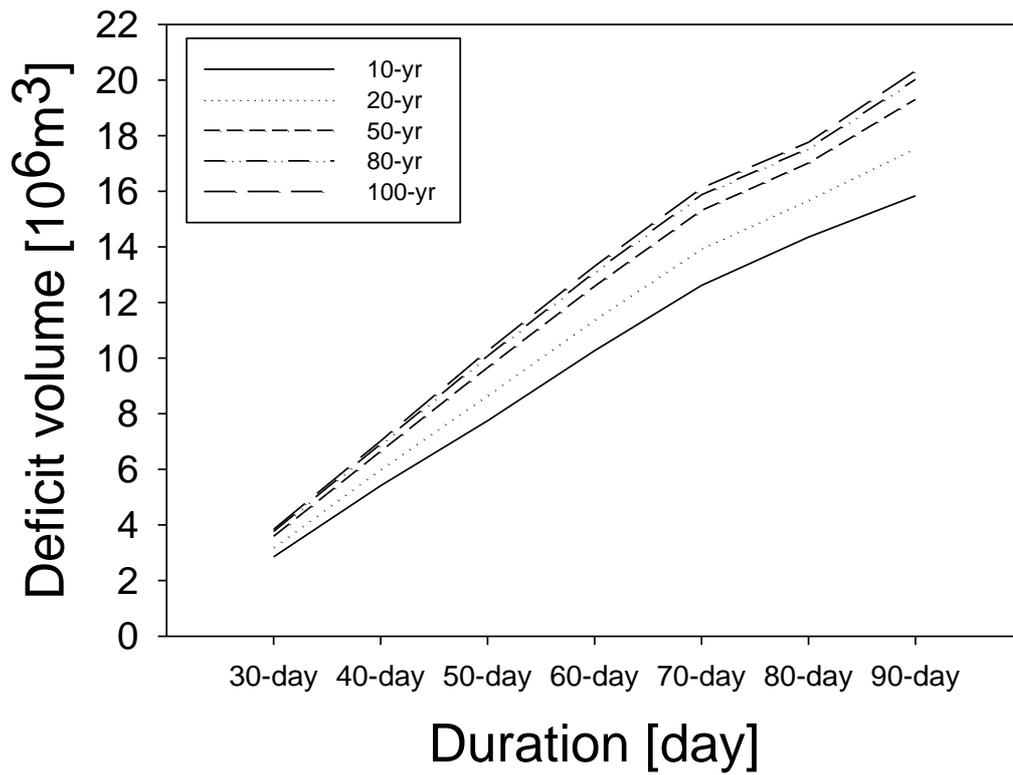


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539 (d) Desired yield.

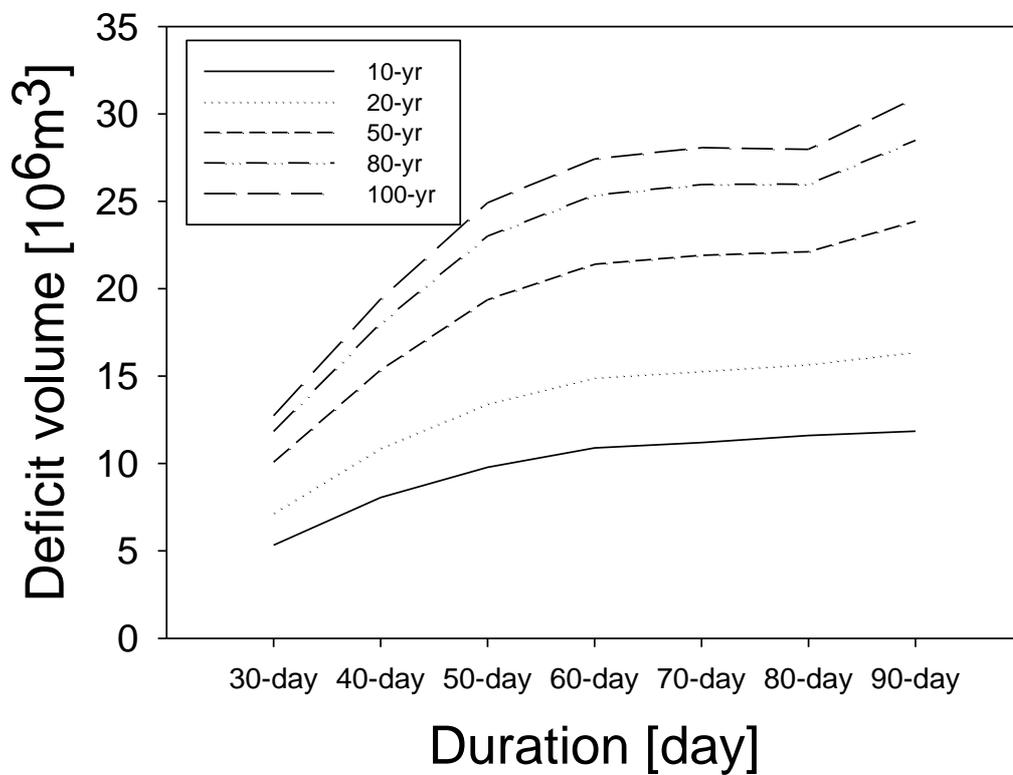
540 Fig. 7. L-moment diagram for probability distribution identification.

541



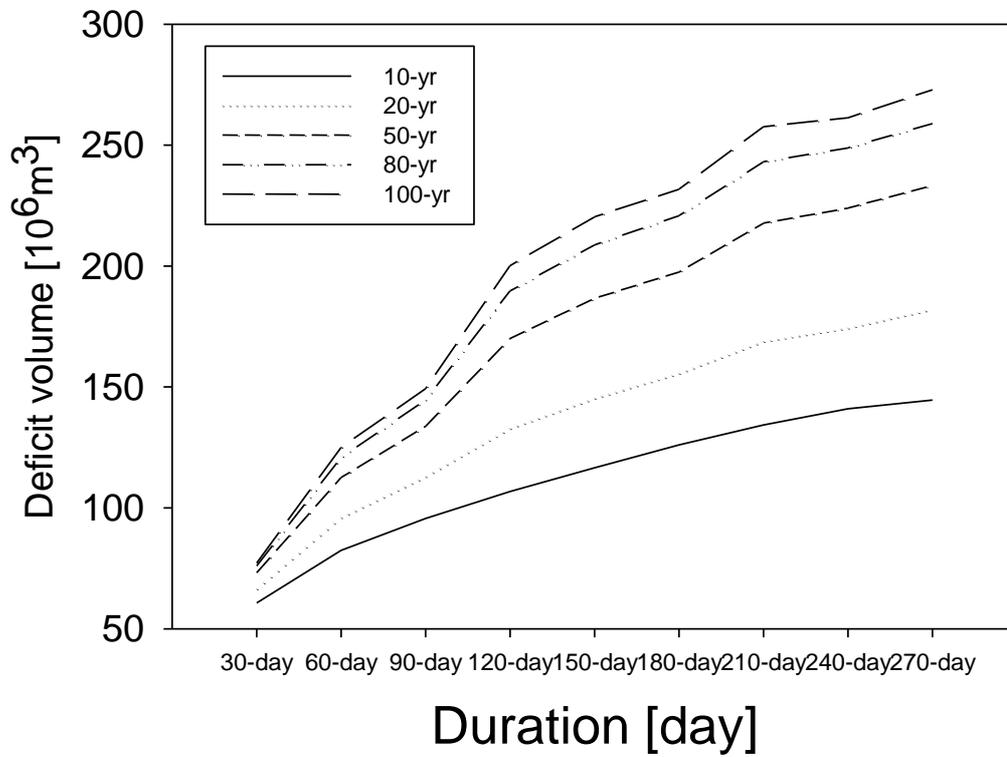
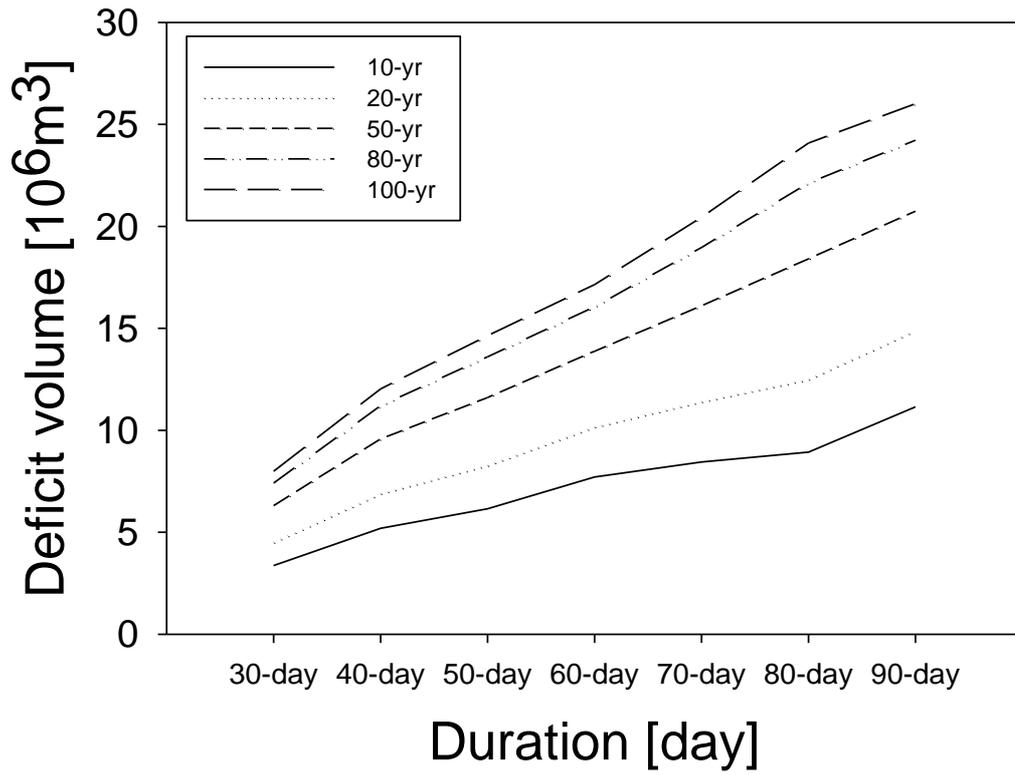
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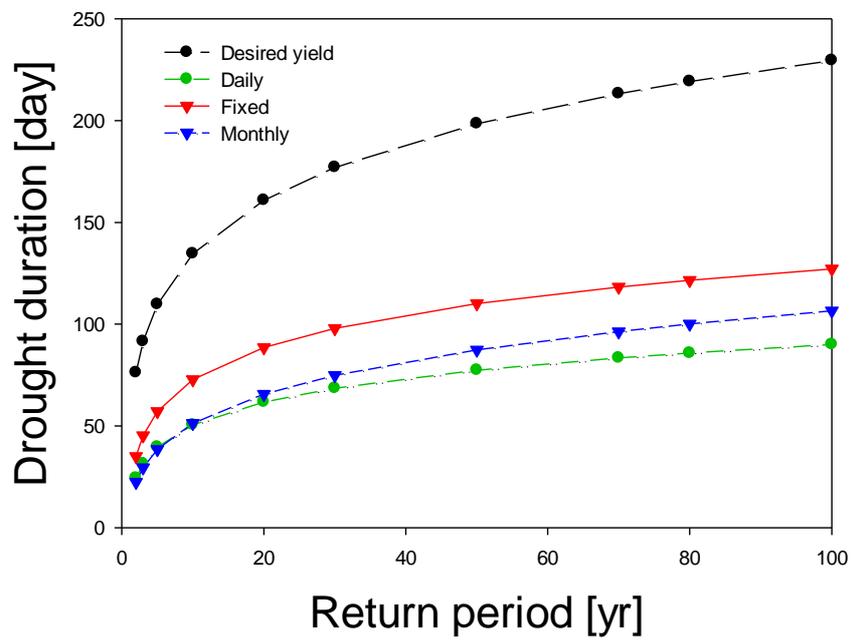
543 (a) Fixed.



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 ³ s ⁻¹]			
	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 ³)
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