

Development of Streamflow Drought Severity-Duration-Frequency Curve Using Threshold Level Method

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Abstract

This study developed a streamflow drought severity-duration-frequency (SDF) curve which is analogous to the well-known intensity-duration-frequency curve used for rainfall. Severity was defined as the total water deficit volume to target threshold for the specific drought duration. The fixed and variable threshold level methods were introduced to set the target instream flow requirement, which can significantly affect the streamflow drought severity. The four threshold levels utilized were fixed, monthly, daily, and desired yield for water use. The fixed in this study is the 70-percentile value (Q_{70}) of flow duration curve (FDC) which resulted from all available daily streamflows and the monthly is monthly-variable Q_{70} s of each month's FDC. The daily variable threshold is Q_{70} of FDC obtained from the antecedent 365 daily streamflows. The desired yield threshold determined by central government consists of domestic, industrial and agricultural water uses and environmental instreamflow. As a result, the desired yield threshold can identify the streamflow drought using total water deficit to the hydrological and socioeconomic targets while the fixed, monthly and daily derived the streamflow deficiencies or anomalies because of the just usage of streamflow data. Based on individual frequency analysis, SDF curves for four thresholds were developed to quantify the relationship among severities, durations and frequencies. For more specification, the drought duration-frequency curve was developed. It can be an effective tool to identify any streamflow droughts using severities, durations and frequencies.

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

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35 **1. Introduction**

36 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 streamflow intervene with the normal agricultural and economic
39 activities of a region, leading to a decrease in agriculture production and in turn affecting
40 everyday life. Dracup et al. (1980) defined drought as follows: 1) nature of water deficit (e.g.,
41 precipitation, soil moisture, or streamflow); 2) basic time unit of data (e.g., month, season, or
42 year); 3) threshold for distinguishing low flows from high flows while considering the mean,
43 median, mode, or any other derived thresholds; and 4) regionalization and/or standardization.
44 Based on these definitions, various indices have been proposed over the years to identify
45 drought. That is, recent studies have focused on such multi-faceted drought characteristics
46 using various indices (Palmer, 1965; Rossi et al., 1992; McKee et al., 1993; Byun and Wilhite,
47 1999; Tsakiris et al., 2007; Pandey et al., 2008a; 2008b; 2010; Nalbantis and Tsakiris, 2009;
48 Wang et al., 2011; Tabari et al, 2013; Tsakiris et al., 2013).

49 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 resulted from the absence or
52 reduction of precipitation and short-term dryness results in an agricultural drought that
53 severely reduces crop yields. Precipitation deficits over a prolonged period reducing
54 streamflow, groundwater, reservoir and lake levels, will result in a hydrological drought. If
55 hydrological drought continues until the supply and demand of some economic goods is
56 damaged, a socioeconomic drought happens (Heim, 2002).

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

72 For more specifications, 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 have been
76 widely applied for drought analyses (Yevjevich, 1967; Sen, 1980; Dracup et al., 1980;
77 Dalezios et al., 2000; Kjeldsen et al., 2000; American Meteorological Society, 2002; Hisdal
78 and Tallaksen, 2003; Wu et al. 2007; Pandey et al., 2008a; Yoo et al., 2008; Tigkas et al.,
79 2012; van Huijgevoort, 2012). These approaches provide an analytical interpretation of the
80 expected availability of river flow; a drought occurs when the streamflow falls below the
81 threshold level. This level is frequently taken as a certain percentile flow for specific duration
82 and is assumed to be steady during the considered month, season, or year. Kjeldsen et al.
83 (2000) extended the steady threshold concept to the variable method, employing seasonal,
84 monthly and daily streamflows.

85 Based on the typical drought characteristics (water deficit and duration) and threshold
86 levels, this study developed a comprehensive concept to quantify the streamflow drought
87 severity which is closely related to hydrological and socioeconomic drought, using fixed,
88 monthly, daily and desired yield threshold levels. Furthermore, this study proposed the
89 streamflow drought severity-duration-frequency (SDF) curve using traditional frequency
90 analyses. This methodology was applied to the Seomjin River basin in South Korea.

92 **2. Methodology**

93 **2.1 Procedure**

94 This study consists of five steps, as shown in Fig. 1. Step 1 is to determine the
95 threshold levels for fixed, monthly, daily, and desired yield for water use. Step 2 is to
96 calculate the severities (total water deficits) and durations for all drought events at the four
97 threshold levels. Step 3 is to derive the annual maxima of severity and duration and to identify
98 the best-fitted probability distribution functions using L-moment ratio diagrams (Hosking and
99 Wallis, 1997). Step 4 is to calculate the streamflow drought severities using the selected
100 probability distribution with best-fitted parameters and to develop SDF curves. Step 5 is to
101 develop the duration-frequency curves of four threshold levels using appropriate probability
102 distribution.

104 **Fig. 1**

106 **2.2. Streamflow drought severity**

107 In temperate regions where the runoff values are typically larger than zero, the most

108 widely used method to estimate a hydrological drought is the threshold level approach
 109 (Yevjevich, 1967; Fleig et al., 2006; Tallaksen et al., 2009; Van Loon and Van Lanen, 2012).
 110 The streamflow drought severity with threshold level method has the following advantages
 111 over SPI (Standardized Precipitation Index) for meteorology (Yoo et al., 2008) and PDSI
 112 (Palmer Drought Severity Index) for meteorology and agriculture (Dalezios et al., 2000): 1)
 113 no a priori knowledge of probability distributions is required and 2) drought characteristics,
 114 such as frequency, duration, and severity, are directly produced if the threshold is set by
 115 sectors impacted by the drought.

116 A sequence of drought events can be obtained using the streamflow and threshold
 117 levels. Each drought event is characterized by its duration, D_i , deficit volume (or severity),
 118 S_i , and time of occurrence, T_i , as shown by the definition sketch in Fig. 2. With a prolonged
 119 dry period, the long drought spell is divided into a number of minor drought events. Because
 120 these droughts are mutually dependent, Tallaksen et al. (1997) proposed that the independent
 121 sequence of drought events must be described using some types of pooling, as described
 122 below.

123

124 **Fig. 2**

125

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

130

$$d_{pool} = d_i + d_{i+1} + t_c,$$

131

$$s_{pool} = s_i + s_{i+1} - z_c \tag{1}$$

132 This study assumed $t_c = 3$ days and $z_c = 10\%$ of d_i or d_{i+1} for simplicity. This numbers
 133 should be studied later in more detail.

134

135 2.3 Threshold selection

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

142 entire record period. The daily varying threshold can be derived using the antecedent 365
143 daily streamflow.

144 The threshold choice is influenced by the study objective and region and available
145 data. In general, a percentile from FDC can be used as the threshold. Relatively low
146 thresholds in the range of Q_{70} to Q_{95} are often used for perennial rivers (Kjeldsen et al.,
147 2000). This study selected Q_{70} for the fixed threshold considering the Korean hydrologic
148 condition, namely, a monsoon climate. Q_{70} means a 70% flow of FDC. That is, 70% is the
149 percentage of time that the streamflow, Q_{70} , is exceeded. However, the threshold selection
150 should be more studied, because Q_{70} doesn't have any obvious evidences to be a
151 representative threshold for the monsoon climate rivers.

152 The time resolution, where to apply series of annual, monthly or daily streamflow,
153 depends on the hydrologic regime under study area. In the temperate zone a given year might
154 include both severe droughts (seasonal droughts) and months with abundant streamflow,
155 meaning that annual data wouldn't often reveal severe droughts. Dry regions are more likely
156 to experience droughts lasting for several years, multi-year droughts, which supports the use
157 of a monthly or an annual time step. Hence, different time resolutions might lead to difference
158 results regarding the drought event selection. This study used daily streamflow data and the
159 time resolutions were selected from 30 days to 270 days because droughts of the study area
160 have never studied.

161

162 **Fig. 3**

163

164 The variable threshold approach is adapted to detect streamflow deviations for both
165 high and low flow seasons. Lower than normal flows during high flow seasons might be
166 important for later drought development. However, periods with relatively low flow either
167 during the high flow season, for example, due to a delayed onset of a snow-melt flood, isn't
168 commonly considered a drought. Therefore, the events defined with the varying threshold
169 should be called streamflow deficiency or streamflow anomalies rather than streamflow
170 drought (Hisdal et al., 2004). On the other hand, the desired yield for sufficient water supply
171 and environmental instreamflow can be an effective way to identify the streamflow drought
172 with the consideration of hydrological and socioeconomic demands since environmental
173 instreamflow becomes important in recent years.

174

175 **2.4 Probability distribution function**

176 L-moment diagram among various goodness-of-fit techniques was used to evaluate

177 the best probability distribution function for datasets in several recent studies (Hosking, 1990;
 178 Chowdhury et al., 1991; Vogel and Fennessey, 1993; Hosking and Wallis, 1997). The L-
 179 moment ratio diagram is a graph where the sample L-moment ratios, L-skewness (τ_3), and L-
 180 kurtosis (τ_4) are plotted as a scatterplot and compared with the theoretical L-moment ratio
 181 curves of candidate distributions. L-moment ratio diagrams have been suggested as a useful
 182 graphical tool for discriminating amongst candidate distributions for a dataset (Hosking and
 183 Wallis, 1997). Two representations used to assist in the selection of statistical distributions are
 184 the sample average and line of best fit, which can be plotted on the same graph to facilitate
 185 selecting the best-fit distribution.

186 When plotting an L-moment ratio diagram, the relationship between the parameters
 187 and L-moment ratios τ_3 and τ_4 for several distributions are required. In the case of a GEV
 188 distribution, the three-parameter GEV distribution described by Stedinger et al. (1993) has the
 189 following probability density function (PDF) and cumulative distribution function (CDF):

$$190 \quad f(x) = \frac{1}{\alpha} \left\{ 1 - \frac{\kappa}{\alpha} (x - \xi) \right\}^{1/\kappa - 1} \cdot \exp \left[- \left\{ 1 - \frac{\kappa}{\alpha} (x - \xi) \right\}^{1/\kappa} \right] \quad \kappa \neq 0 \quad , \quad (3a)$$

$$191 \quad f(x) = \frac{1}{\alpha} \exp \left\{ - \frac{x - \xi}{\alpha} - \exp \left(- \frac{x - \xi}{\alpha} \right) \right\} \quad \kappa = 0 \quad , \quad (3b)$$

$$192 \quad F(x) = \exp \left[- \left\{ 1 - \frac{\kappa}{\alpha} (x - \xi) \right\}^{1/\kappa} \right] \quad \kappa \neq 0 \quad , \quad (4a)$$

$$193 \quad F(x) = \exp \left\{ - \exp \left(- \frac{x - \xi}{\alpha} \right) \right\} \quad \kappa = 0 \quad , \quad (4b)$$

194 where $\xi + \alpha/\kappa \leq x \leq \infty$ for $\kappa < 0$; $-\infty \leq x \leq \infty$ for $\kappa = 0$; and $-\infty \leq x \leq \xi + \alpha/\kappa$ for
 195 $\kappa > 0$. Here, ξ is a location, α is a scale, and κ is a shape parameter. For $\kappa = 0$, the GEV
 196 distribution reduces to the classic Gumbel (EV1) distribution with $\tau_3 = 0.17$. Hosking and
 197 Wallis (1997) provided more detailed information regarding the GEV distribution. The
 198 relationship between the parameters and τ_3 and τ_4 for the shape parameter's GEV
 199 distribution can be obtained as follows (Hosking and Wallis, 1997):

$$200 \quad \tau_3 = \frac{2(1 - 3^{-\kappa})}{(1 - 2^{-\kappa})} - 3, \quad (5a)$$

$$201 \quad \tau_4 = \frac{5(1 - 4^{-\kappa}) - 10(1 - 3^{-\kappa}) + 6(1 - 2^{-\kappa})}{(1 - 2^{-\kappa})}. \quad (5b)$$

202

203 3. Study region

204 The Seomjin River basin is located in southwestern Korea (Fig. 4). The area and total
 205 length of the Seomjin River are approximately 4,911.9 km² and 212.3 km, respectively. The

206 altitude range is rather large spanning from approximately 0 to 1,646 m (Fig. 4). The climate
207 of South Korea is characterized by extreme seasonal variations. Winter is cold and dry under
208 the dominant influence of the Siberian air mass, whereas the summer is hot and humid with
209 frequent heavy rainfall associated with the East Asian monsoon. In the Seomjin River basin,
210 the measured precipitation is mainly concentrated in summer, and the measured mean annual
211 precipitation varies from $< 1,350 \text{ mm/yr}^{-1}$ (in the north region) to $> 1,600 \text{ mm/yr}^{-1}$ (in the
212 southeastern region) during the 1975-2012 observation period. In general, approximately 60%
213 of the annual precipitation occurs during the wet season (July through September) in South
214 Korea. This extreme seasonality in the precipitation causes periodic shortages of water during
215 the dry season (October through March) and flood damage during the wet season.

216

217 **Fig. 4.**

218

219 The administrative districts where the basin is located cover three provinces, four
220 cities, and 11 counties (Namwon City, Jinan County, Imsil County, and Sunchang County in
221 the Northern Jeolla Province; Suncheon City, Gwangyang City, Damyang County, Gokseong
222 County, Gurye County, Hwasun County, Boseong County, and Jangheung County in the
223 Southern Jeolla Province; and Handing County in the Southern Gyeongsang Province). Influx
224 rates into the basin by province are 47% (Southern Jeolla Province), 44% (the Northern Jeolla
225 Province), and 9% (Southern Gyeongsang Province), and a total of 129,322 households and
226 321,104 residents live in these areas.

227 The land use consists of arable land (876.29 km^2), forest land ($3,400.61 \text{ km}^2$), urban
228 area (67.12 km^2), and other land uses (567.86 km^2). Additionally, 69.2% of the entire basin
229 area ($4,911.89 \text{ km}^2$) is forest land. Major droughts occurred in the Southern Jeolla Province
230 from 1967 to 1968 and from 1994 to 1995. The Seomjin River basin had $< 1,000 \text{ mm}$ of
231 precipitation on average in 1977, 1988, 1994, and 2008. Among these years, the annual
232 precipitation in 1988 was only 782.7 mm (56.5%) of the annual average of 1,385.5 mm from
233 1967 to 2008, representing a serious drought. According to the “River Survey Report (K-
234 water, 1992)”, a drought in Seomjin river basin occurs approximately every 10 years.

235

236 **4. Results**

237 **4.1. Determination of the threshold levels**

238 This study used four threshold levels. The fixed is Q_{70} of FDC which resulted from
239 37-year daily streamflows. The monthly thresholds are twelve Q_{70} s of monthly FDCs which
240 resulted from all daily streamflows of January, February ... and December for the past 37
241 years, respectively. The daily threshold is Q_{70} of FDCs which resulted from the antecedent

242 365 daily streamflows. Thus, the daily changes smoothly every day. The desired yield
243 threshold for sufficient water supply and environmental instreamflow was determined by
244 Korean central government. That is, it is related to social and economic droughts since it
245 associates the supply and demand of some economic goods and environmental safety. The
246 desired yield threshold differed considerably from the other levels and represented more
247 realistic conditions because the desired yield is equivalent to the planned water supply.

248 The calculated thresholds are presented in Fig. 5, and the specific monthly-averaged
249 values are listed in Table 1. The average levels were 1.9, 2.5, 2.8, and 13.8 m³s⁻¹ for the fixed,
250 monthly, daily, and desired yields, respectively. The daily threshold levels were highly
251 fluctuating because of the natural streamflow variations for the antecedent 365 days and were
252 the largest of the four threshold levels because a summer period (June, July, and August) was
253 considered. The desired yield level was larger than the fixed, monthly, and daily thresholds.
254 This phenomenon occurred during the winter in Korea, and as a result, both the water demand
255 and natural runoff during the winter (December, January, and February) were quite small.
256 However, the thresholds levels for the daily, monthly, and desired yields during the summer
257 were much higher than those during the other seasons. The threshold levels for the desired
258 yield during May and June were much larger than the levels for the other thresholds because
259 the agricultural water demand was the highest in this season

260

261 **Fig. 5**

262

263 **Table 1**

264

265 **4.2 Calculations of streamflow drought severity and duration**

266 The durations and severities for all streamflow drought events were calculated based
267 on the streamflow drought concept and threshold levels. The annual maxima values of
268 duration and severity are shown in Fig. 6 and the summarized values are listed in Table 2. The
269 maximum durations from the desired yield threshold approach were considerably higher than
270 those from the other thresholds because the desired yields were highest during June and July
271 due to agricultural water use. Similar to the results for drought duration, the severities showed
272 much higher values.

273

274 **Table 2**

275

276 **Fig. 6**

277

278 To compare the differences from four threshold levels, the correlation coefficients were
279 calculated as shown in Table 3. The similar trend was observed in the monthly and daily
280 threshold levels. However, the durations and severity from the desired yield threshold level
281 were completely different from those for fixed, monthly and daily levels. That is, it can be
282 guessed that the drought identification techniques based on general threshold levels cannot
283 reflect the socioeconomic drought in terms of water supply and demand. Therefore, two-way
284 approaches which are anomaly type (fixed, monthly and daily) for hydrological drought and
285 desired yield threshold for socioeconomic drought should be separately included for specific
286 drought characteristics identification.

287

288 **Table 3**

289

290

291 **4.3 Identification of the probability distribution function**

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

296 The L-moment ratio diagrams were derived for the four threshold approaches and are
297 displayed in Fig. 7. Of the distribution models tested, 3 parameter distributions such as the
298 Pearson Type 3(PT3), Generalized Normal (GNO), and Generalized Extreme Value (GEV)
299 distributions appeared consistent with their datasets. In the frequency analysis dealing with
300 extreme values, the distributions which have more than 3 parameters are required for
301 expression of upper tail. PT3, GNO, and GEV distribution can be applied in this study. As
302 shown in Fig. 7, this study selected GEV distribution for a representative probability
303 distribution because most observations are appropriate for the GEV. It corresponds to
304 Dalezios et al. (2000) for PDSI and Yoo et al. (2008) for SPI.

305

306 **Fig. 7**

307

308 **4.4 Development of SDF curves**

309 Streamflow drought SDF curves were developed using the derived probability
310 distribution functions, as shown in Figs. 8. For these plots, 10-, 20-, 50-, 80- and 100-year-
311 frequency severities were calculated at 30-, 60-, 90-, 120-, 150-, 180-, 210-, and 270-day
312 durations. Because the record length of available data is only 37 years, we calculated up to
313 100-year frequency. However, SDF curves from fixed, daily, and monthly thresholds were

314 derived using comparatively short durations since the annual maximum durations varies from
315 30 to 96 days. SDF described streamflow drought severities with respect to any durations and
316 frequencies. The severity increases with increasing frequencies and durations. For specific
317 description, Table 4 compares all of the severities to specific frequencies and durations for the
318 desired yield threshold. As the duration becomes larger, the difference ratio between return
319 periods becomes much bigger. Therefore, because the streamflow drought severity should be
320 more crucial when the drought continues for long period, the frequency to long-drought
321 should be approached with caution.

322

323 **Fig. 8**

324

325 **Table 4**

326

327 **4.5 Development of duration-frequency curve**

328 The drought can be characterized by deficit volume using threshold levels. However,
329 using the deficit volume isn't sufficient to explain the extreme droughts. Thus, analyzing
330 streamflow drought durations can be another useful tool to identify the drought event. So,
331 occurrence probabilities of various duration events were also estimated using general
332 frequency analysis. As a result, using GEV distribution, already shown in Fig. 7, duration-
333 frequency curves for four threshold levels were developed as shown in Fig. 9. For these plots,
334 2-, 3-, 5-, 10-, 20-, 30-, 50-, 70-, 80-, and 100-year-frequency severities were calculated.
335 Similar to SDF curves, the durations for desired yield showed much higher than those for the
336 other three thresholds.

337

338 **Fig. 9**

339

340 **5. Conclusions**

341 This study developed a useful concept to describe the characteristics of streamflow
342 droughts using frequency analyses. SDF curves for streamflow drought were developed to
343 quantify a specific volume according to a specific duration and frequency. Also duration-
344 frequency curves were used to derive the relationship between drought duration and
345 frequency. This study used severity which represented the total water deficit for the specific
346 durations. Using the L-moment diagram method, the GEV was selected for the best-fit
347 probability distribution. As a result, SDF curve were derived to identify the relationship
348 among streamflow drought severity, duration and frequency. The severities increased with
349 increasing durations and frequencies. However, these values were quite different because the

350 four threshold level approaches defined the streamflow drought differently.

351 Streamflow drought SDF curve developed in this study can be potentially exploited to
352 quantify the water deficit for the natural streams as well as reservoirs. In addition, these will
353 be extended to conduct regional frequency analyses, which can estimate streamflow drought
354 severity at ungagged sites. Therefore, it can be an effective tool to identify any streamflow
355 droughts using severity, duration and frequency.

356

357

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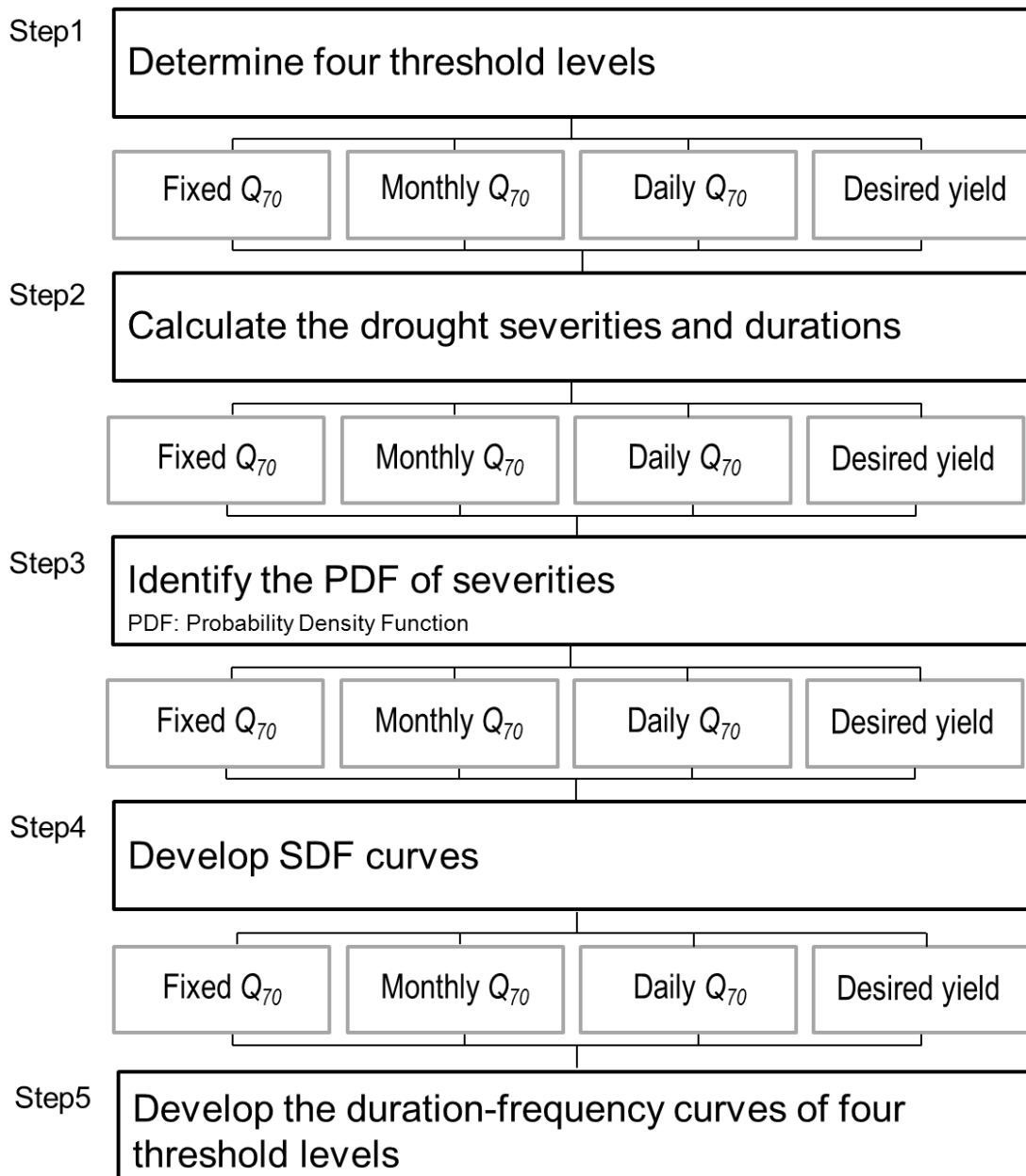
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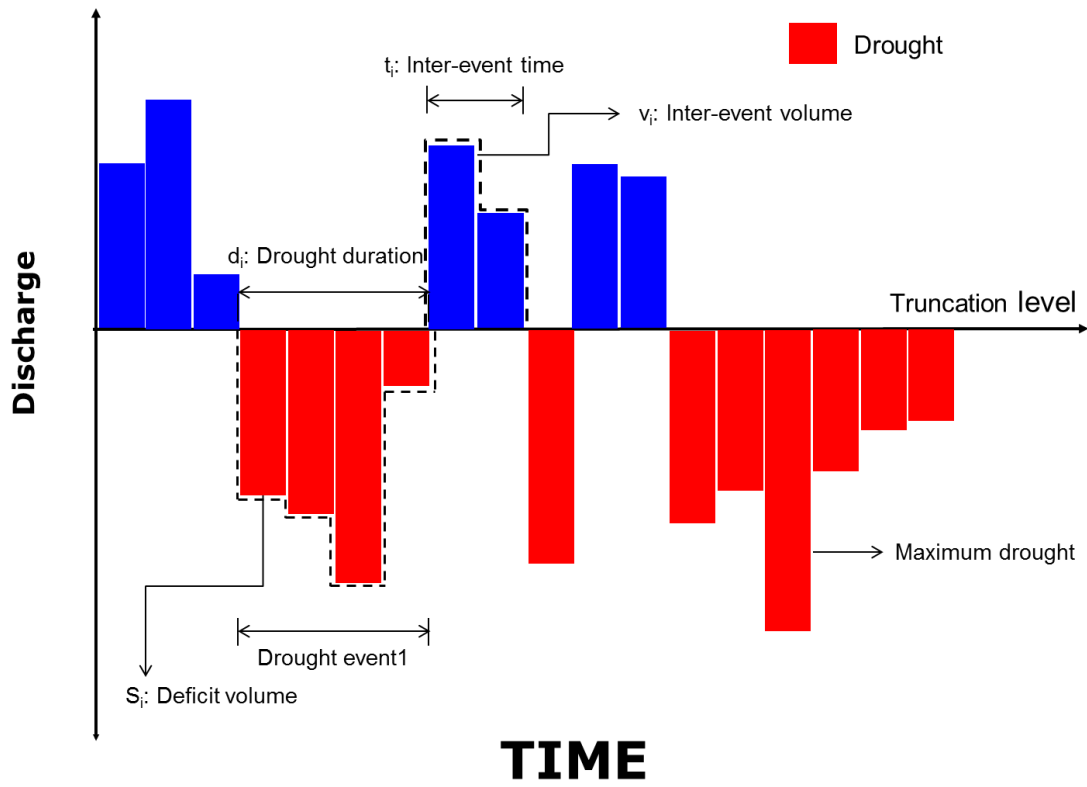


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458 Fig. 1. Procedure of this study

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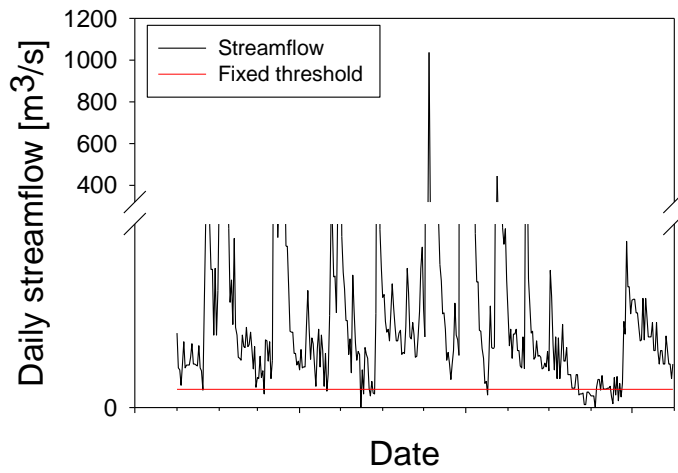


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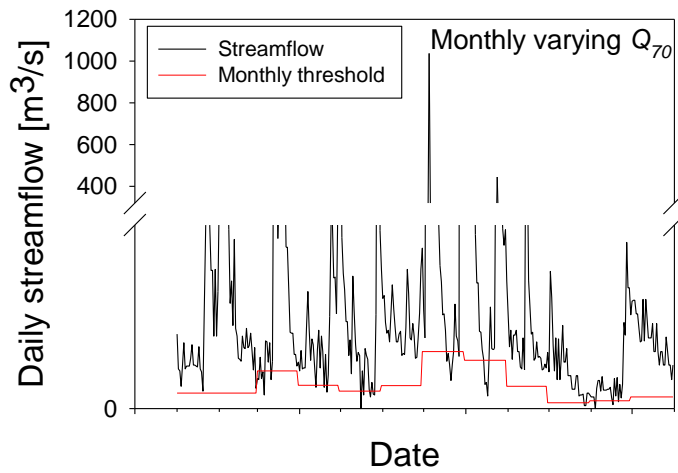
462 Fig. 2. A definition sketch of general drought events

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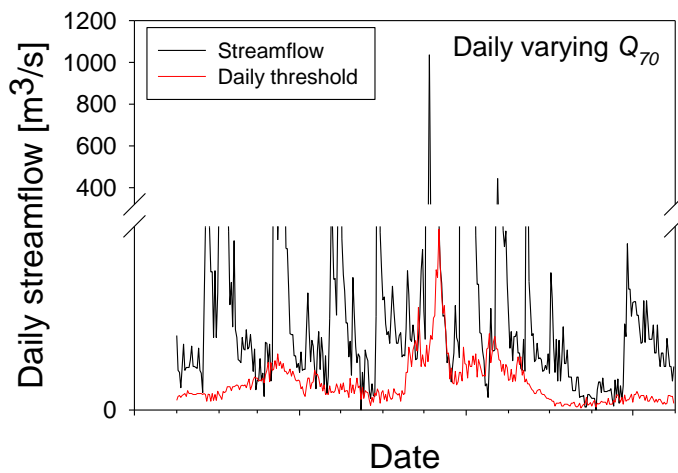
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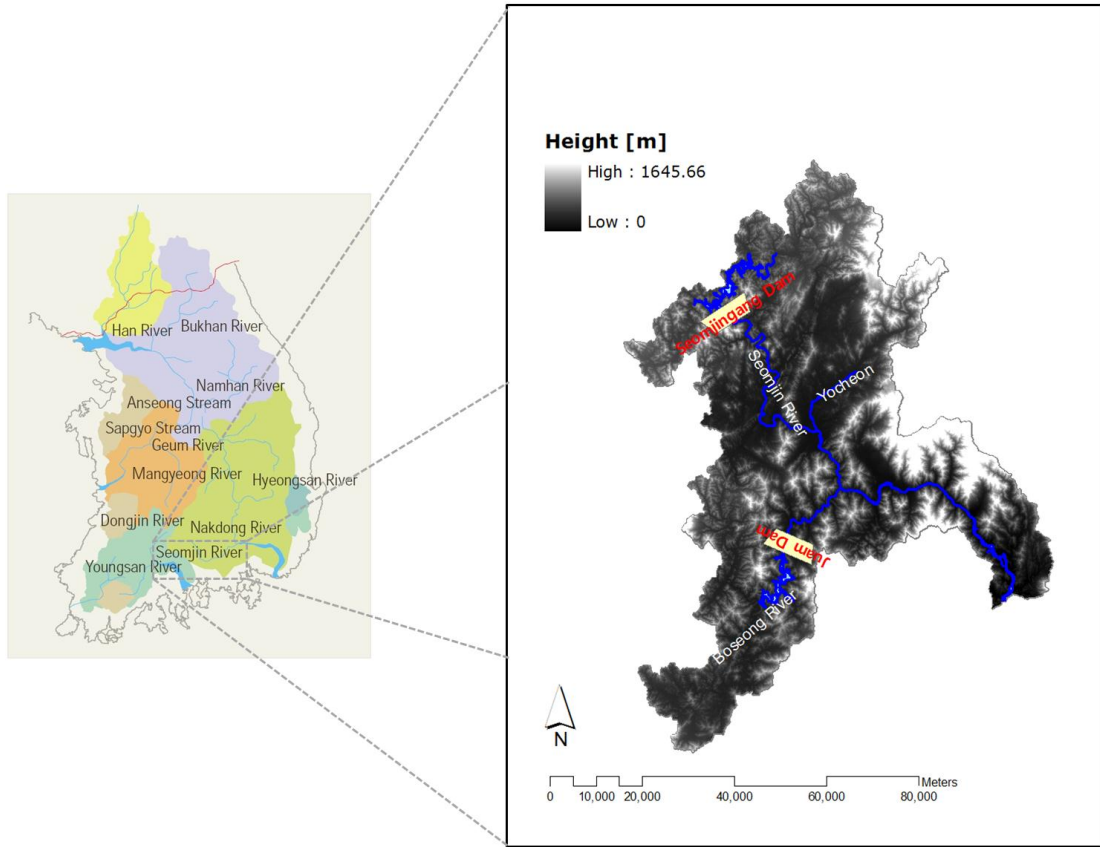


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468 Fig. 3. Examples of threshold levels: Fixed (top); monthly varying (middle); daily varying

469 (bottom)

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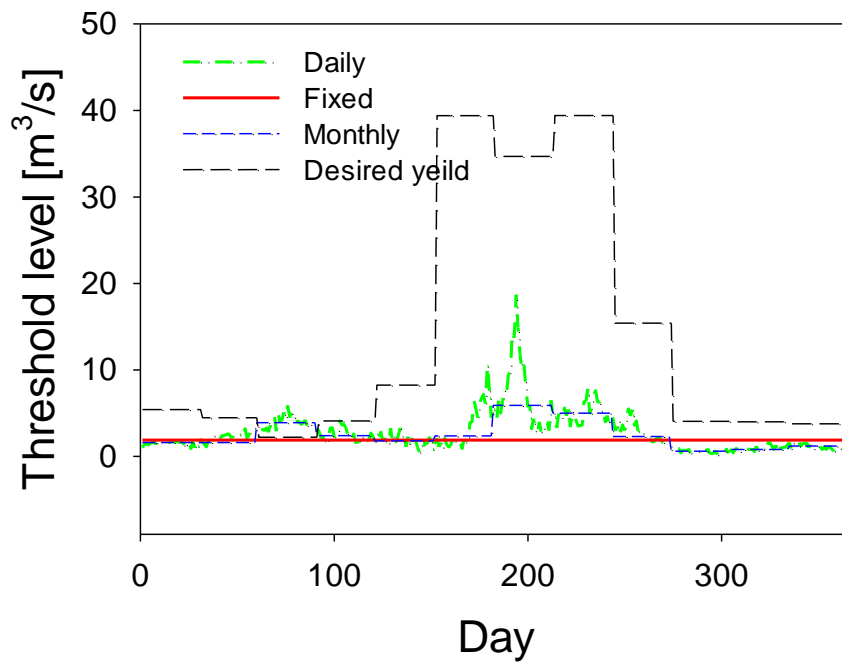


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472 Fig. 4. Location of the selected river basin, including elevation and river

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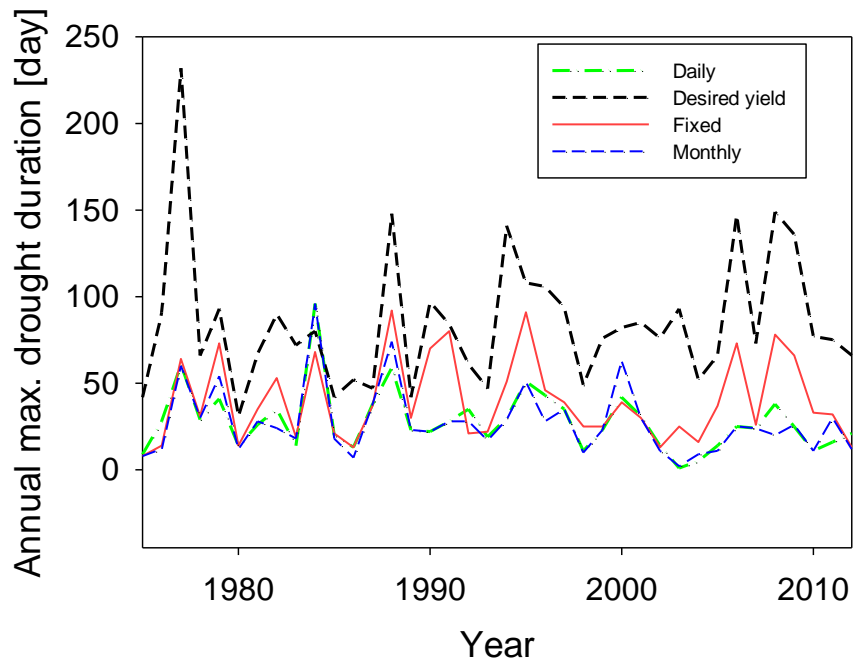


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476 Fig. 5 Comparative four threshold levels used in this study

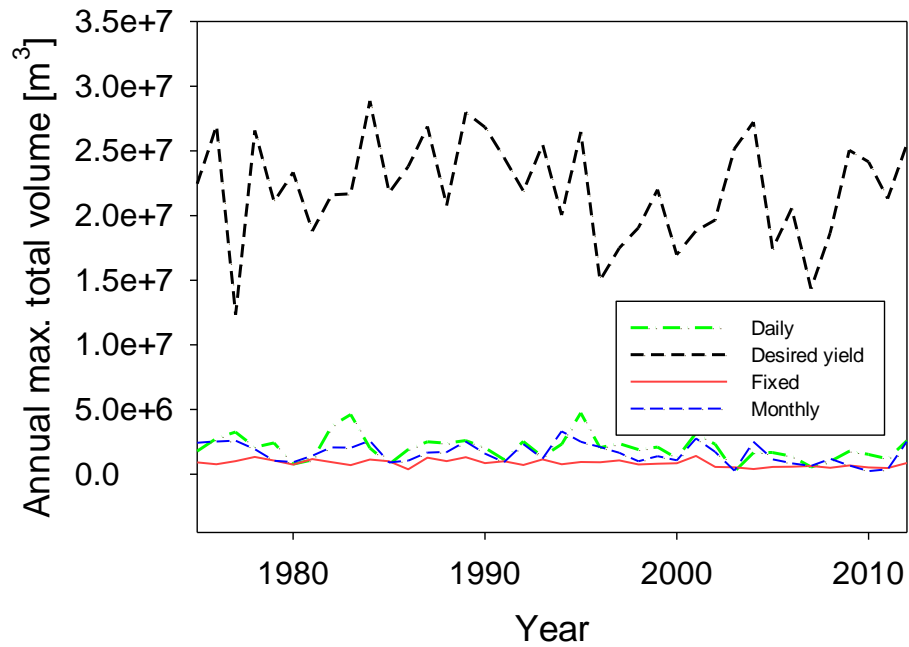
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480 (a) Drought duration

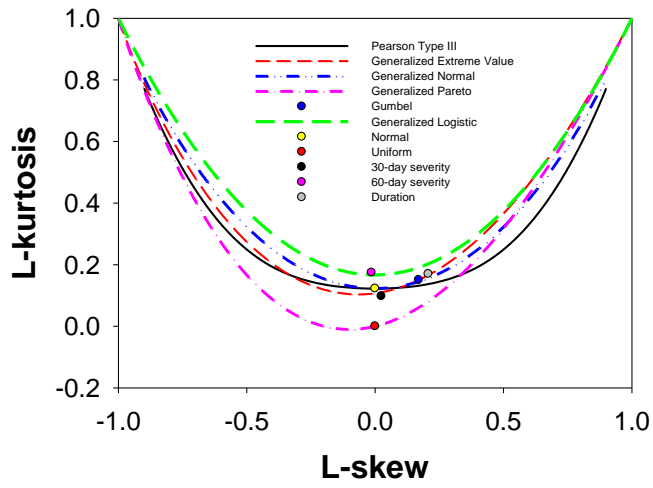


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482 (b) Total water deficit volume (drought severity)

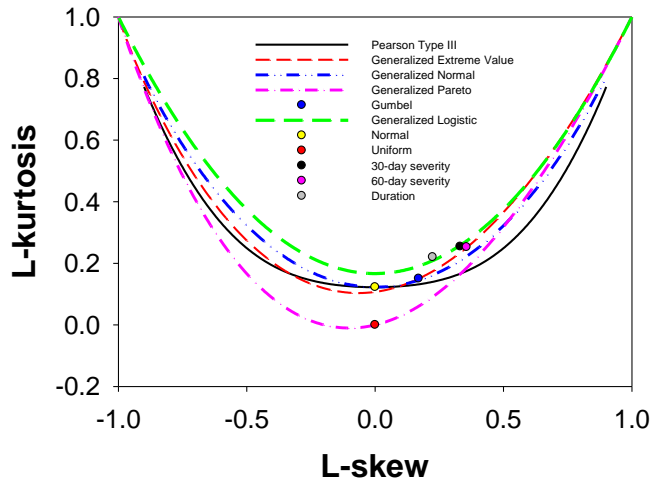
483 Fig. 6. Time series of annual maxima values of duration, severity

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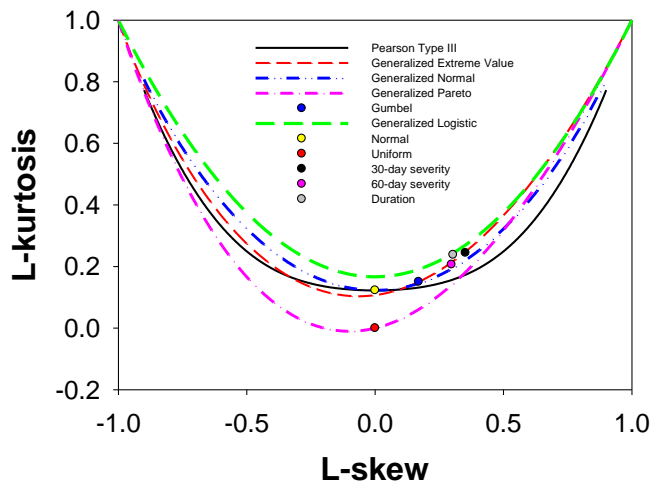
486 (a) Fixed



487

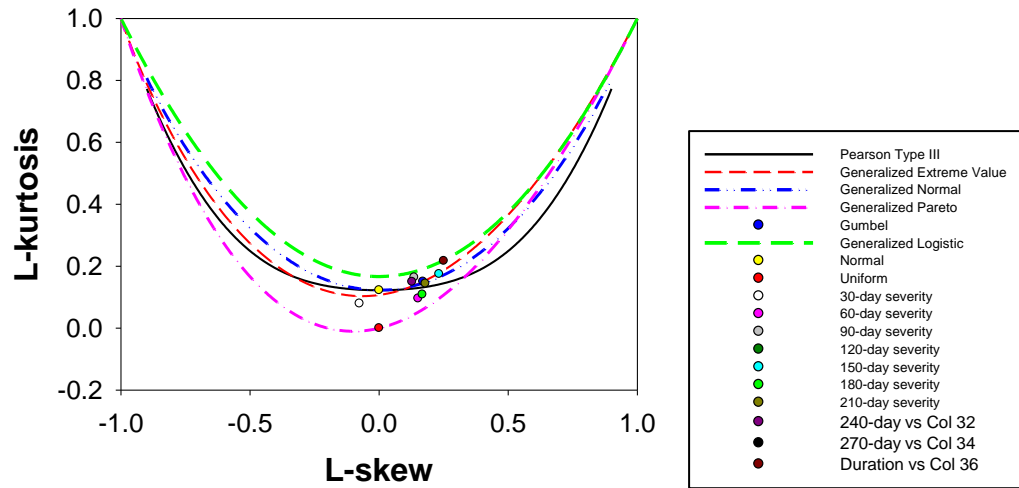
488 (b) Daily

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491 (c) Monthly

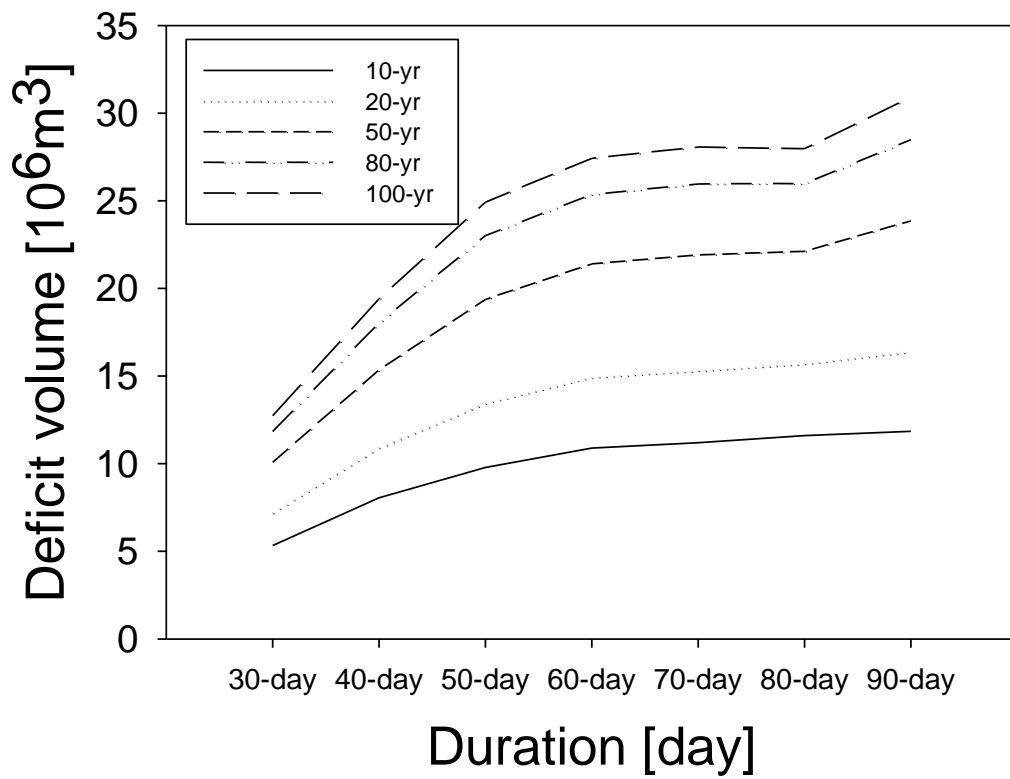
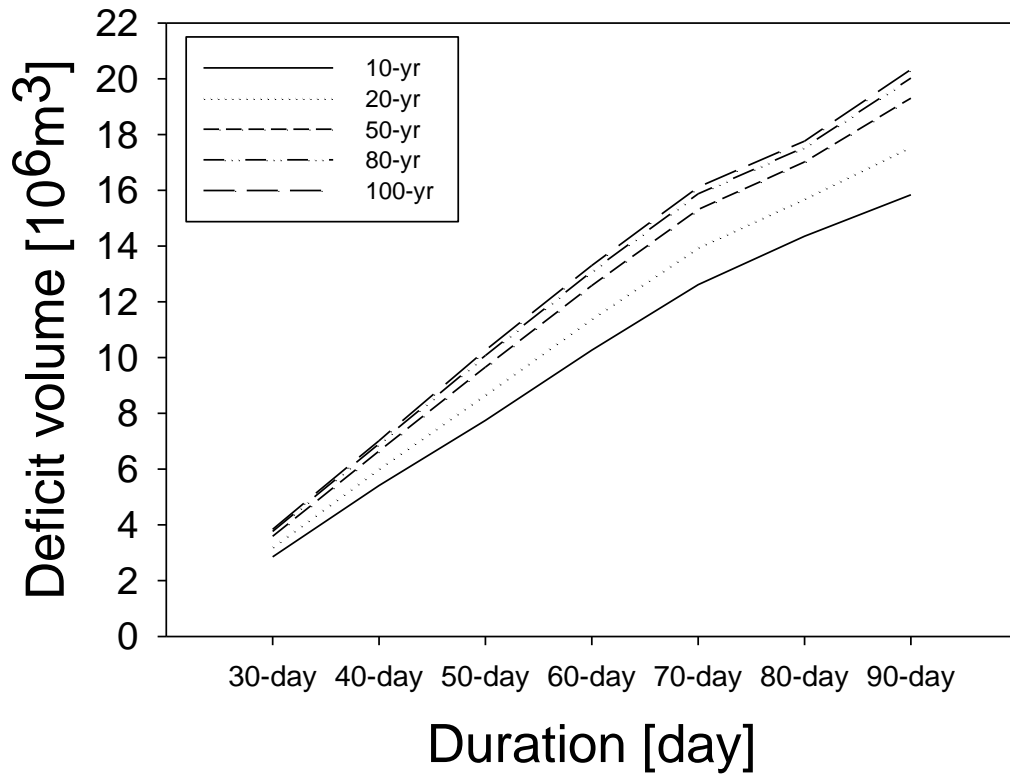


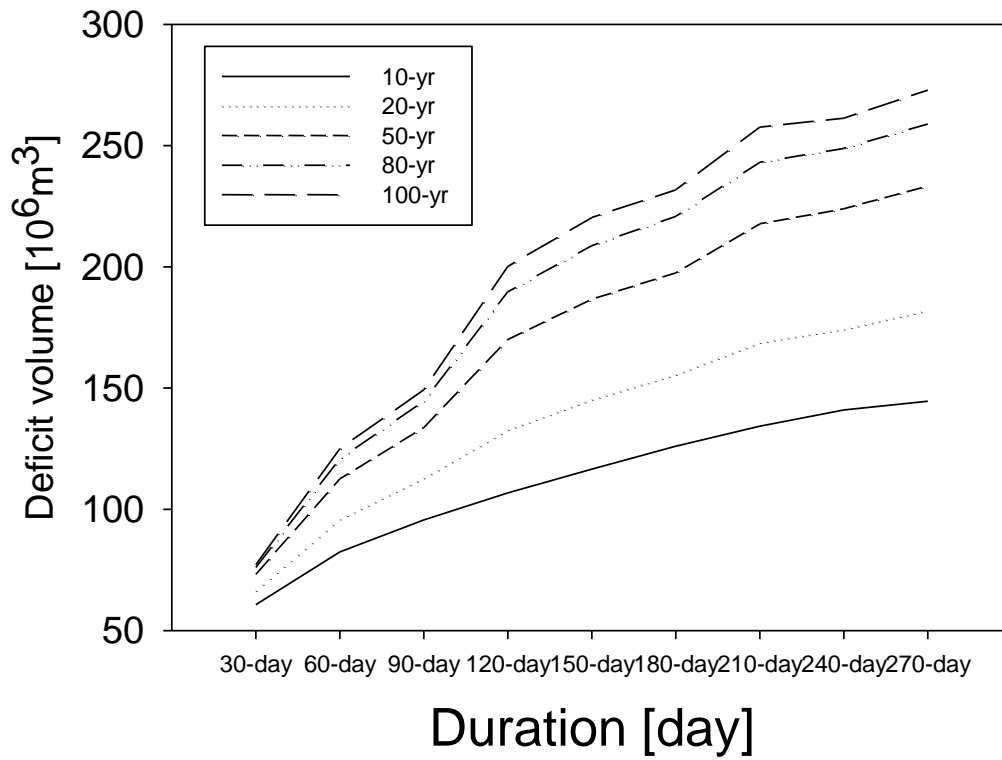
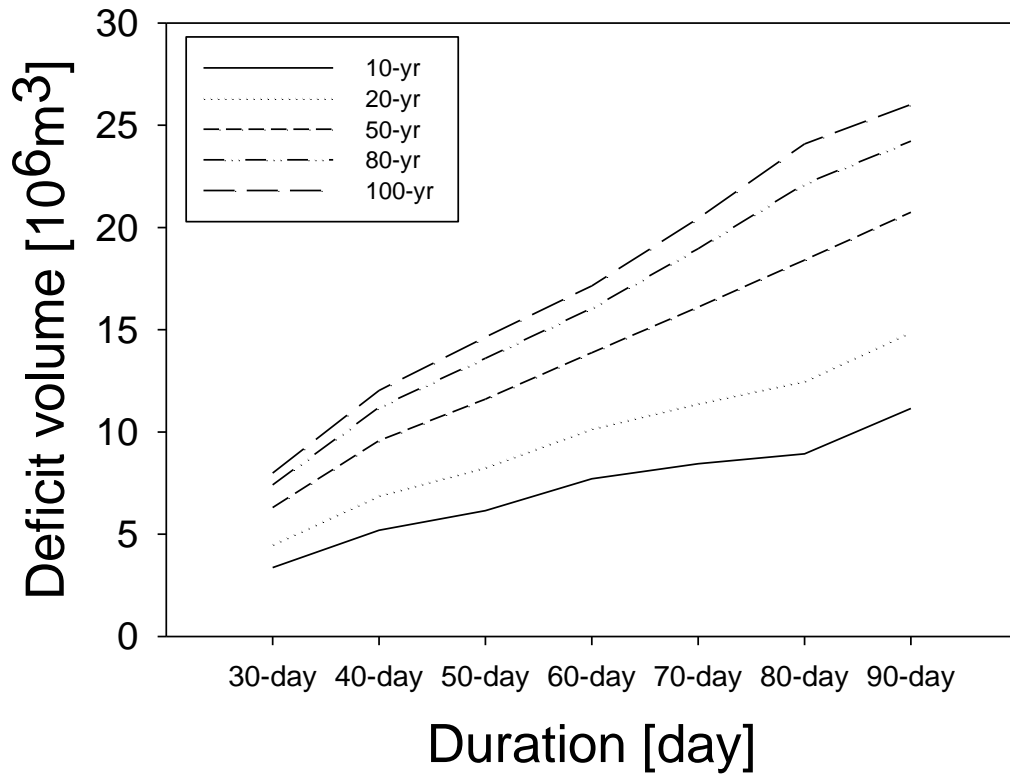
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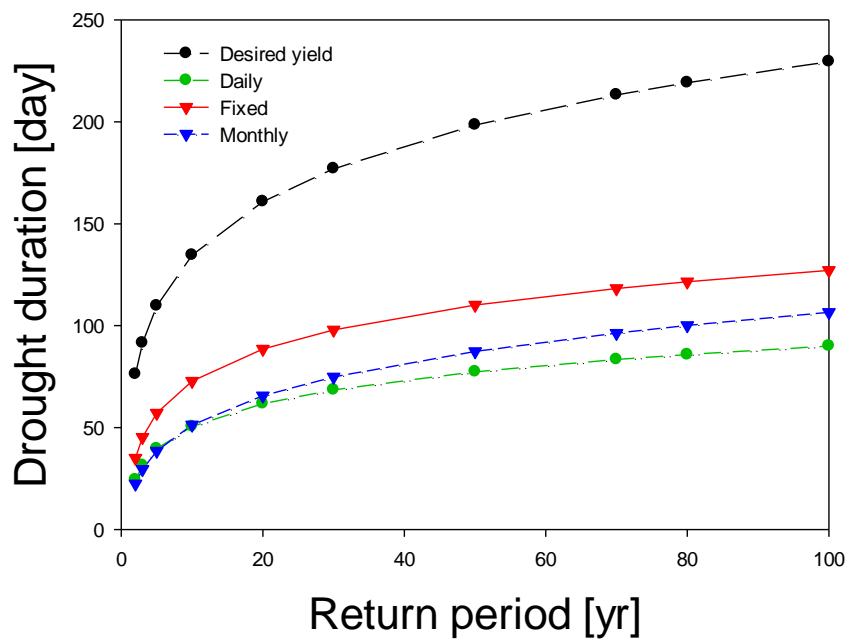
493 (d) Desired yield

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

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508 Fig. 9. Duration frequency curves of four threshold level approaches in Seomjin river basin

509

510 Table 1. Monthly averaged of 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

511

512 Table 2. Summary of 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

513

514

515 Table 3. Correlations between durations and severities from 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

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Table 4. Severity-duration-frequency of desired yield in 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

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