

1 Closer Look at the Baseflow Correlation Method

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3
4 **Abstract:** In 2003, Reilly and Kroll examined the baseflow correlation method at river sites throughout the United States. The current
5 study reexamines Reilly and Kroll's baseflow correlation experiment by investigating the use of different performance metrics, experi-
6 mental parameters, and model assumptions that were not investigated by Reilly and Kroll. The goal of this study is to provide additional
7 guidance on how to implement the baseflow correlation method in practice. The results confirm that baseflow measurements should be
8 obtained during low flow seasons and as far as possible from runoff events. When one has only five baseflow measurements at the
9 low-flow partial-record site, the correlation coefficient between baseflows at gauged and low-flow partial-record sites should be at least
10 0.9; when the number of baseflow measurements is 10 or more, the method performs adequately if the correlation coefficient is greater
11 than 0.6. The performance of the baseflow correlation method improves as the number of baseflow measurements increases, but levels off
12 dramatically when one has more than 10 measurements.

13 **DOI:** XXXX

14 **CE Database subject headings:** Low flow; Base flow; Predictions; Stochastic models.
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17 Introduction

18 The estimation of low-flow statistics at low-flow partial-record
19 sites is a common problem faced by hydrologists and engineers.
20 Low-flow statistics are widely used in water quality management
21 and water supply planning (Smakhtin 2001). The most widely
22 employed low-flow quantile in the United States is the 7-day,
23 10-year low flow ($Q_{7,10}$), which by definition is the tenth percen-
24 tile of the distribution of annual minimum 7-day average flows
25 (Riggs 1980).

26 Regional regression is a common method used for estimating
27 low-flow statistics at ungauged river sites (Smakhtin 2001; Kroll
28 et al. 2004). However, this method often performs poorly in prac-
29 tice to estimate low flows (Thomas and Benson 1970; Barnes
30 1985; Hammett 1985; Arihood and Glatfelter 1986; Vogel and
31 Kroll 1992; Ries 1994; Kroll et al. 2004). Riggs (1965, 1972)
32 proposed correlating baseflow measurements obtained at the low-
33 flow partial-record site with concurrent daily flows at a nearby
34 gauged site as an alternative. Stedinger and Thomas (1985) pro-
35 posed an improved d -day, T -year low-flow estimator and devel-
36 oped a first-order estimator of its variance. This technique is often
37 referred to as the baseflow correlation method (Reilly and Kroll
38 2003).

39 Stedinger and Thomas (1985) examined the performance of

the baseflow correlation method with 20 pairs of stream sites. 40
Reilly and Kroll (2003) expanded this experiment to investigate 41
its performance to estimate the $Q_{7,10}$ at more than 1,300 river sites 42
across the United States. In the experiment by Reilly and Kroll, 43
they employed streamflow sites from the USGS's Hydro-Climatic 44
Data Network (HCDN) (Slack and Landwehr 1992); these 45
streamflow sites are also employed in the current study. Reilly 46
and Kroll concluded that the baseflow correlation method per- 47
forms well in the United States when baseflow measurements are 48
obtained by randomly choosing one baseflow measurement from 49
consecutive recessions (referred to as the "recession" method) 50
and the potential candidate gauged sites are restricted within 51
200 km. They also suggested using at least ten baseflow measure- 52
ments. Their experiment supported the suggestion by Stedinger 53
and Thomas (1985) that the correlation coefficient of concurrent 54
baseflows between the gauged and low-flow partial-record sites 55
should be at least 0.70. 56

The experiment performed by Reilly and Kroll (2003) required 57
a number of experimental parameters. The current study revisits 58
this baseflow correlation experiment and investigates the follow- 59
ing experimental parameters and model assumptions: 60

- The impact of different performance metrics on the results of 61
the experiment; 62
- When to take baseflow measurements at low-flow partial- 63
record sites; 64
- How to designate streamflows as under baseflow conditions; 65
- The impact of the correlation coefficient between concurrent 66
baseflows at the gauged and low-flow partial-record sites (ρ) 67
and segment length; 68
- The assumption that the frequency factor at the low-flow 69
partial-record site is equivalent to the frequency factor at the 70
gauged site; and 71
- The assumption that the log-linear relationship between the 72
annual minimums at the low-flow partial-record site and those 73
at the gauged site is the same as the log-linear relationship 74
between the concurrent baseflows at the two sites. 75

The results provide details as to the tradeoffs between the 76
correlation coefficient and the number of required baseflow mea- 77

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Note. Discussion open until August 1, 2007. Separate discussions
must be submitted for individual papers. To extend the closing date by
one month, a written request must be filed with the ASCE Managing
Editor. The manuscript for this paper was submitted for review and pos-
sible publication on July 1, 2005; approved on May 16, 2006. This paper
is part of the *Journal of Hydrologic Engineering*, Vol. 12, No. 2, March
1, 2007. ©ASCE, ISSN 1084-0699/2007/2-1-XXXX/\$25.00.

78 surements in the baseflow correlation method. The results also
 79 provide additional guidance on how to implement the baseflow
 80 correlation method in practice.

81 Methodology

82 The baseflow correlation method proposed by Stedinger and Tho-
 83 mas (1985) is summarized in the following. This method has
 84 several basic assumptions. The first assumption of the baseflow
 85 correlation method is a linear relationship between y_i , the loga-
 86 rithm of the d -day annual minimum flows at a low-flow partial-
 87 record site, and those at a gauged site, x_i

$$88 \quad y_i = \alpha + \beta x_i + \varepsilon_i, \quad \varepsilon_i \sim N(0, \sigma_\varepsilon^2) \quad (1)$$

89 where α and β =model parameters; and ε_i =independent normal
 90 error terms with a mean of zero and a constant variance, σ_ε^2 .

91 Second, since annual minimum flows are not available for the
 92 low-flow partial-record site, it is assumed that the relationship
 93 between d -day annual minimum flows is similar to the
 94 relationship between instantaneous baseflows. This assumption is
 95 examined in this paper. Thus the linear relationship between the
 96 logarithm of baseflow measurements at the low-flow partial-
 97 record site, \bar{y}_i , and the logarithm of corresponding baseflows at
 98 the gauged site, \bar{x}_i , is given by

$$99 \quad \bar{y}_i = \alpha + \beta \bar{x}_i + \varepsilon_i, \quad \varepsilon_i \sim N(0, \sigma_\varepsilon^2) \quad (2)$$

100 The third assumption is that the annual minimum streamflows
 101 are described by a log Pearson Type 3 (LP3) distribution. The
 102 LP3 distribution has been used by USGS for describing annual
 103 minimum streamflow series in the United States (Rumenik and
 104 Grubbs 1996; Wandle and Randall 1993; Barnes 1985). By this
 105 assumption, the logarithm of $Q_{7,10}$ at the low-flow partial-record
 106 site can be estimated by (Stedinger and Thomas 1985)

$$107 \quad \ln(\hat{Q}_{7,10}) = \hat{\mu}_y + K_y \hat{\sigma}_y \quad (3)$$

108 where $\hat{\mu}_y$ =estimator of the log-space mean; $\hat{\sigma}_y$ =estimator of the
 109 log-space variance; and K_y =associated frequency factor for the
 110 LP3 distribution. The frequency factor is a function of the log-
 111 space skew of the 7-day annual minimum flows and the percentile
 112 of interest. In the baseflow correlation method, the frequency fac-
 113 tor for the low-flow partial-record site, K_y , is assumed equal to the
 114 frequency factor for the gauged site, K_x , an assumption that is
 115 explored in this paper. Thus only estimators of $\hat{\mu}_y$ and $\hat{\sigma}_y$ are
 116 required; Stedinger and Thomas (1985) suggested the unbiased
 117 estimators

$$118 \quad \hat{\mu}_y = a + b m_x \quad (4)$$

$$119 \quad \hat{\sigma}_y^2 = b^2 s_x^2 + s_e^2 \left(1 - \frac{s_x^2}{(L-1)s_x^2} \right) \quad (5)$$

120 where m_x and s_x^2 =log-space mean and variance of the 7-day an-
 121 nual minimum flows at the gauged site, respectively; s_x^2 =sample
 122 variance of the logarithms of the baseflows at the gauged site;
 123 L =number of concurrent baseflow measurements; and a , b , and
 124 s_e^2 =ordinary least-squares estimators of the parameters α , β , and
 125 σ_ε^2 (Draper and Smith 1966) estimated using concurrent baseflow
 126 measurements. Stedinger and Thomas (1985) derived the variance
 127 of the $Q_{7,10}$ estimator as

$$\text{Var}[\ln(\hat{Q}_{7,10})] \cong \frac{s_e^2}{L} + \frac{(m_x - m_{\bar{x}})^2 s_e^2}{(L-1)s_x^2} + \frac{b^2 s_x^2}{n} + \frac{K_y^2}{4\hat{\sigma}_y^2} \left(\frac{4b^2 s_x^4 s_e^2}{L s_x^2} \right. \\ \left. + \frac{2b^4 s_x^4}{n} + \frac{2s_e^4}{L} \right) + \frac{2b s_x^2 (m_x - m_{\bar{x}}) K_y s_e^2}{L \hat{\sigma}_y s_x^2} \quad (6)$$

where n =number of years record at the gauged site. 130

Jackknife Simulation Experiment 131

The jackknife simulation experiment performed by Reilly and 132
 Kroll (2003) is outlined in the following. To assess the perfor- 133
 mance of the baseflow correlation method for estimating $Q_{7,10}$ at 134
 low-flow partial-record sites, this experiment is performed at 135
 gauged sites using the at-site $Q_{7,10}$ estimators derived from the 136
 historic record as “true” values. Each gauged site is sequentially 137
 selected as the site where the $Q_{7,10}$ is estimated using the baseflow 138
 correlation method. This site is referred to as the low-flow partial- 139
 record site hereafter. The sites with “true” $Q_{7,10}$ of zero (i.e., 10% 140
 or more historic 7-day annual minimums are zero) are excluded 141
 from this experiment. After D days of continuously decreasing 142
 streamflow, the streamflow at the low-flow partial-record site is 143
 designated as under baseflow conditions. Reilly and Kroll used D 144
 of 5 days; D of 3, 5, and 7 days is investigated in this study. A 145
 baseflow segment, which is a series of streamflows measured during 146
 baseflow conditions, is constructed at the low-flow partial- 147
 record river site. Reilly and Kroll employed streamflows during 148
 typical low flow months (July through October); here streamflows 149
 across a whole year are also considered. The number of stream- 150
 flow measurements in the series is called the segment length. 151
 Segment lengths of 5, 10, 15, and 20 days were employed by 152
 Reilly and Kroll. Reilly and Kroll employed three methods for 153
 constructing baseflow segments: the “consecutive” method, which 154
 uses consecutive baseflow days; the “random” method using ran- 155
 domly selected baseflow days from consecutive years; and the 156
 “recession” method that employs one random baseflow day from 157
 consecutive baseflow recessions. 158

Potential candidate gauged sites are searched to find those that 159
 have baseflow conditions on the same days that comprise the 160
 baseflow segment. To investigate how to select a gauged site 161
 based on a simple criterion, i.e., the distance between the gauged 162
 site and low-flow partial-record site, three potential candidate 163
 gauged site selection methods are investigated: (1) sites within the 164
 same USGS water resources region; (2) sites within 100 km; and 165
 (3) sites within 200 km. The three methods are necessitated by the 166
 low density of HCDN sites in some regions; in practice one might 167
 find gauged sites much closer than 100 km. The number of candi- 168
 date gauged sites for each segment is denoted as N . If the num- 169
 ber of candidate gauged sites for the segment is greater than or 170
 equal to a specified value, the segment is designated as a valid 171
 segment. The baseflow correlation method is then used to esti- 172
 mate the $Q_{7,10}$ and its variance using each candidate gauged site. 173
 The estimator with the smallest variance is chosen as the best 174
 $Q_{7,10}$ estimate for the baseflow segment and compared to the at- 175
 site $Q_{7,10}$ estimate at the low-flow partial-record site. This process 176
 is repeated for all baseflow segments at the low-flow partial- 177
 record site. All other gauged sites are then sequentially designated 178
 as low-flow partial-record sites. The low-flow partial-record sites 179
 with at least one valid segment are designated as valid low-flow 180
 partial-record sites, which represent the subset of all sites used to 181
 assess the baseflow correlation method. 182

183 Performance Metrics

184 The performance metrics examined by Reilly and Kroll (2003)
 185 were average relative absolute difference (ARAD), relative bias,
 186 and root of relative mean-square error. The ARAD was primarily
 187 used in their analysis since it is easily interpretable as the average
 188 percent deviation from the “true value,” where the true value is
 189 the at-site $Q_{7,10}$ estimate obtained using the historic record. The
 190 ARAD is calculated as

$$191 \text{ ARAD} = \frac{\sum_{i=1}^M \left(\frac{|\hat{Q}_{7,10i} - Q_{7,10}|}{Q_{7,10}} \right)}{M} \quad (7)$$

192 where $\hat{Q}_{7,10i}$ = i th baseflow correlation estimate of the $Q_{7,10}$ at the
 193 low-flow partial-record site; $Q_{7,10}$ =at-site $Q_{7,10}$ estimate obtained
 194 via a LP3 frequency analysis using the historic record; and
 195 M =number of valid segments. All of the metrics used by Reilly
 196 and Kroll consider relative errors, which reduce the influence of
 197 sites with large $Q_{7,10}$ values on the performance metrics. The
 198 problem with these metrics is that they can instead amplify the
 199 impact of sites with small $Q_{7,10}$ values at the expense of sites with
 200 large $Q_{7,10}$ values. To investigate the impact of the metric, results
 201 are stratified based on the magnitude of the at-site $Q_{7,10}$ estimates.
 202 The results of ARAD across the United States, with a segment
 203 length of 10 days, minimum N of 3, $\rho \geq 0.7$, $D=5$, low-flow
 204 months, and sites within 200 km, are shown in Fig. 1(a). It can be
 205 seen that the ARAD increases dramatically as the at-site $Q_{7,10}$
 206 decreases, causing sites with small $Q_{7,10}$ values to dominate the
 207 results. The relative bias and root of relative mean square error
 208 have similar patterns as ARAD.

209 To create a more consistent performance metric that is less
 210 influenced by the magnitude of the observations, a number of
 211 additional performance metrics are investigated. This includes the
 212 absolute difference (AD) and unit area absolute difference
 213 (UAAD), which are computed at each low-flow partial-record site
 214 as

$$215 \text{ AD} = \frac{\sum_{i=1}^M |\hat{Q}_{7,10i} - Q_{7,10}|}{M} \quad (8)$$

$$216 \text{ UAAD} = \frac{\sum_{i=1}^M \left(\frac{|\hat{Q}_{7,10i} - Q_{7,10}|}{A} \right)}{M} \quad (9)$$

217 where A =drainage area of the low-flow partial-record site. These
 218 metrics are then aggregated across the conterminous United
 219 States as was explained in Reilly and Kroll (2003).

220 The results of AD across the United States, with a segment
 221 length of 10 days, minimum N of 3, $\rho \geq 0.7$, $D=5$, low-flow
 222 months, and sites within 200 km, are shown in Fig. 1(b). As ex-
 223 pected, the sites with large $Q_{7,10}$ values dominate the AD results.
 224 If UAAD is instead used as the performance metric, Fig. 1(c) is
 225 obtained. While UAAD increases as the at-site $Q_{7,10}$ values in-
 226 creases, the trend is much less dramatic than when using AD.
 227 Though a unique best performance metric is not identified here, it
 228 is informative to investigate how conclusions vary as perfor-
 229 mance metrics vary. One should be particularly sensitive to per-
 230 formance metrics when a large range of true values is present, or
 231 the values are bounded (e.g., Kroll and Stedinger 1996). Here, the
 232 UAAD appears more consistent compared to the other two met-

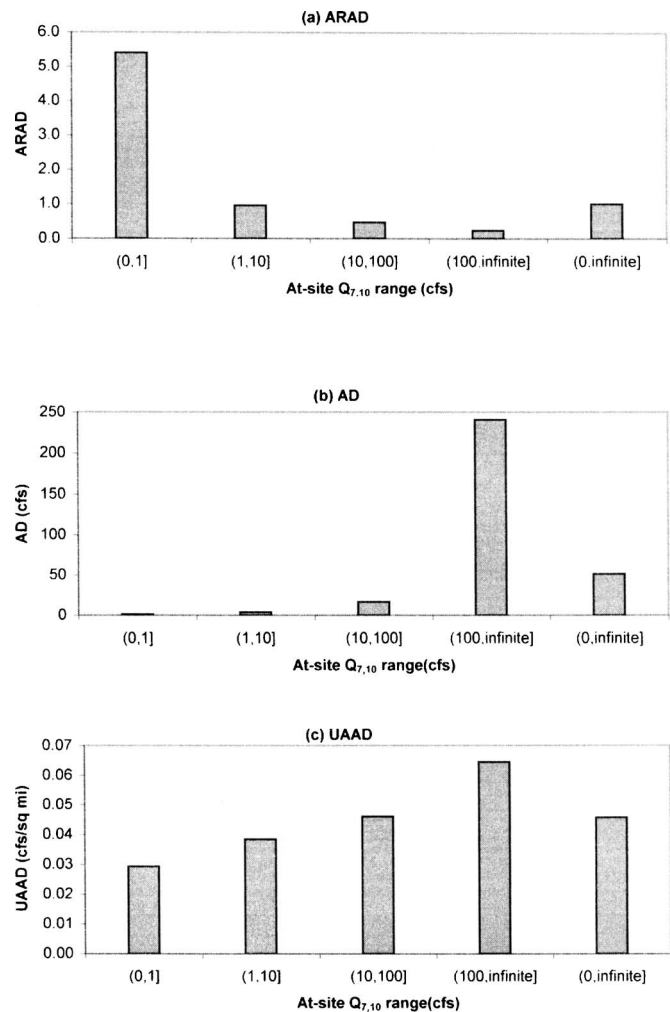


Fig. 1. The performance metrics for the United States with different $Q_{7,10}$ ranges for segment length of 10 days, minimum N of 3, $\rho \geq 0.7$, $D=5$, and sites within 200 km for: (a) ARAD; (b) AD; and (c) UAAD. The x-axis is the ranges of $Q_{7,10}$, i.e., $Q_{7,10}$ from 0 to 1 cfs, from 1 to 10 cfs, etc.

rics examined, since it is not as dramatically impacted by the
 magnitude of $Q_{7,10}$ values and is thus better to assess method
 performance at river sites with a large range of streamflow dis-
 charges.

237 New Results for the United States

In this section, the original results of Reilly and Kroll (2003) are
 presented employing the UAAD. Of interest is whether conclu-
 sions change using this new metric. Though not presented here,
 the results indicate that the random and recession methods for
 compiling a baseflow segment perform much better than the con-
 secutive method. These results, which are the same as those found
 by Reilly and Kroll, indicate the need for nearly independent
 baseflow measurements. Since one would typically not have the
 time to employ the random method, all results shown here are for
 the recession method, which one could employ in practice. Reilly
 and Kroll found that the gauged site selection methods had minimal
 impact on the method performance, and that sites within 200 km
 of each other generally performed as well as sites within

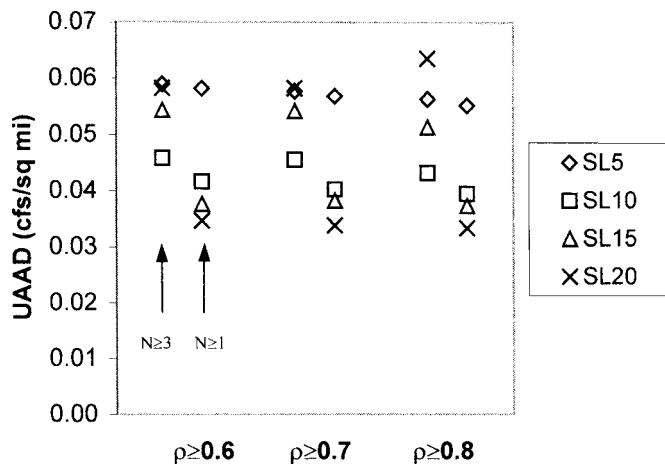


Fig. 2. UAAD for the United States with the recession method, $D=5$, and sites within 200 km, for $\rho \geq 0.6$, $\rho \geq 0.7$, and $\rho \geq 0.8$ and segment lengths of 5, 10, 15, and 20 days. The first, third, and fifth columns are for $N \geq 3$, whereas the second, fourth, and sixth columns are for $N \geq 1$.

251 100 km. We also found this result, and thus only results for sites
 252 within 200 km are presented.
 253 The results for the recession method are shown in Fig. 2. The
 254 first, third, and fifth columns in Fig. 2 are the UAAD for $N \geq 3$,
 255 $D=5$, and $\rho \geq 0.6$, 0.7 , and 0.8 , respectively. Surprisingly, the
 256 conclusion that longer segment lengths (SLs) improve perfor-
 257 mance is not reached for SLs of 15 and 20 days when $N \geq 3$. In
 258 addition, contrary to Reilly and Kroll (2003), the results do not
 259 clearly show the pattern that the performance is improved by
 260 increasing the correlation coefficient when $N \geq 3$. The reason for
 261 these results may be due to the fact that the number of valid
 262 segments is very small for long segment lengths when the mini-
 263 mum N is equal to 3. Table 1 contains the percentage of valid
 264 sites and average valid segments per site for gauged sites within
 265 200 km, $\rho \geq 0.6$, $D=5$, and $N \geq 3$. It can be seen that there are
 266 only 20 valid sites for a segment length of 20 days with 1.3 valid
 267 segments per site when $N \geq 3$.
 268 To address this issue, the minimum N was instead set to 1.
 269 Thus only one candidate gauged site was necessary in order to
 270 have a valid baseflow segment. Table 2 contains the percentage of
 271 valid sites and average valid segments per site for gauged sites
 272 within 200 km, $\rho \geq 0.6$, $D=5$, and $N \geq 1$. For a segment length of
 273 20 days, there are now 363 valid sites, and the average number of
 274 valid baseflow segments per site is 3.7. Using a minimum N of 1
 275 greatly increases the number of valid segments employed in the
 276 jackknife experiment.

Table 1. Impact of Segment Lengths on Valid Sites Used for the Analysis of the Recession Method, Using Gauged Sites within 200 km, $\rho \geq 0.6$, $D=5$, and $N \geq 3$

Segment length (days)	Valid sites	Percentage of valid sites	Average valid segments per site
5	1,249	90.2	53.7
10	689	49.7	7.3
15	213	15.4	2.5
20	20	1.4	1.3

Table 2. Impact of Segment Lengths on Valid Sites Used for the Analysis of the Recession Method, Using Gauged Sites within 200 km, $\rho \geq 0.6$, $D=5$, and $N \geq 1$

Segment length (days)	Valid sites	Percentage of valid sites	Average valid segments per site
5	1,338	96.6	95.7
10	1,198	86.5	19.6
15	823	59.4	7.6
20	363	26.2	3.7

The second, fourth, and sixth columns in Fig. 2 are the UAAD
 across the entire United States for $N \geq 1$, $D=5$, $\rho \geq 0.6$, 0.7 , and
 0.8 , respectively. The surprising results observed with $N \geq 3$ are
 no longer present. For instance, as the segment length increases,
 the performance of the baseflow correlation method improves for
 all ρ . In addition, as the ρ increases, the method performance also
 increases, as one would expect. The results show that the mini-
 mum N value had a large impact on the performance of the base-
 flow correlation method for large segment lengths as the number
 of valid segments is increased considerably by using a minimum
 N of 1. Therefore, the minimum N is taken as 1 for subsequent
 results in this paper.

Based on the new results using UAAD, it appears that having
 $\rho \geq 0.6$ performs nearly as well as when $\rho \geq 0.7$ and 0.8 , regard-
 less of the segment length. One reason for this may be due to the
 fact that sites with $\rho \geq 0.7$ and 0.8 are included in the results for
 $\rho \geq 0.6$, and thus there is duplicate information in these results.
 The impact of ρ on the baseflow correlation method is further
 investigated in this paper in the section entitled “The Impact of
 Correlation Coefficient (ρ).”

When to Take Baseflow Measurements

Reilly and Kroll (2003) only employed flow records during the
 typical low flow months of July through October as annual low
 flow occur in most regions of the United States during this peri-
 od. As most streamflow sites annually have only 5–10 stream-
 flow recessions of adequate length during the low flow months, it
 will take a number of years to gather enough baseflow measure-
 ments to implement the baseflow correlation method. One way to
 reduce this time is to take baseflow measurements throughout the
 entire year, an issue not investigated by Reilly and Kroll. The
 impact of using all the flow records across the entire year is
 investigated in this section. The results are shown in Fig. 3. It is

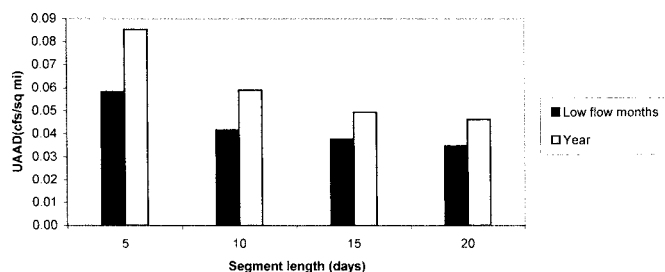


Fig. 3. The impact of using flow record of an entire year (“Year”) and low-flow months for the recession method, gauged sites within 200 km, $\rho \geq 0.6$, $D=5$, and $N \geq 1$

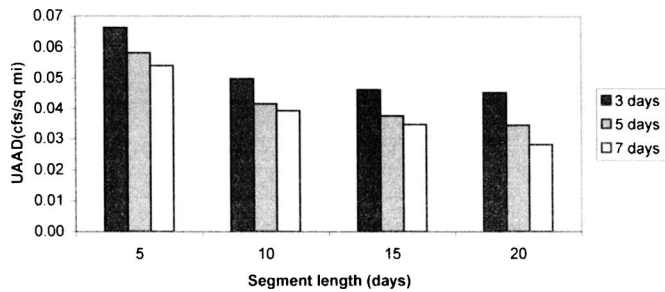


Fig. 4. The impact on UAAD of D , the number of days of continuously decreasing flows to designate baseflow conditions, for the recession method, using gauged sites within 200 km, $\rho \geq 0.6$, and $N \geq 1$

309 easy to see that the baseflow correlation method performs much
 310 worse when the flow records across an entire year are used than
 311 when the flow records during low flow months are employed.
 312 These results confirm that the baseflow measurements should be
 313 taken during the low flow months, from July through October, in
 314 the United States. All subsequent results in this paper are for the
 315 low flow months.

316 How to Designate Baseflows

317 Reilly and Kroll (2003) designated streamflows at a low-flow
 318 partial-record site as baseflows after 5 days of continuously de-
 319 creasing streamflow ($D=5$). The impact of the number of days of
 320 continuous decreasing streamflow (D) is investigated in this sec-
 321 tion. The experiment is repeated for D of 3, 5, and 7. The results
 322 are shown in Fig. 4. The method performance is improved as the
 323 number of days of continuous decreasing streamflow increases.
 324 This result indicates that the method performs better as the base-
 325 flow measurements are obtained further away from runoff events,
 326 though results are more similar when D is 5 and 7 than when D is
 327 3 and 5.

328 Impact of Correlation Coefficient (ρ) 329 and Segment Length

330 To examine the impact of ρ , the UAAD is computed based on
 331 stratifying ρ . The results are shown in Fig. 5, where UAAD has
 332 ranges of ρ of 0.6–0.7, 0.7–0.8, 0.8–0.9, and 0.9–1.0. This result
 333 shows the pattern that the UAAD decreases as ρ increases. Fig. 5

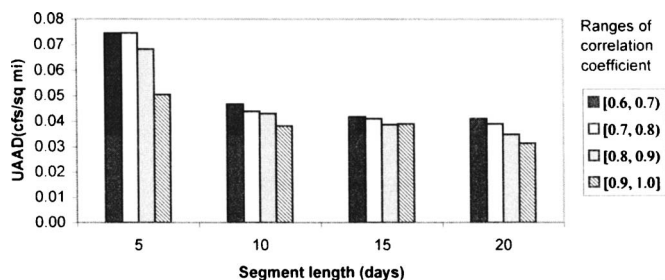


Fig. 5. The impact on UAAD of magnitude of correlation coefficient (ρ) and segment length for recession method, using gauged sites within 200 km, $D=5$, and $N \geq 1$.

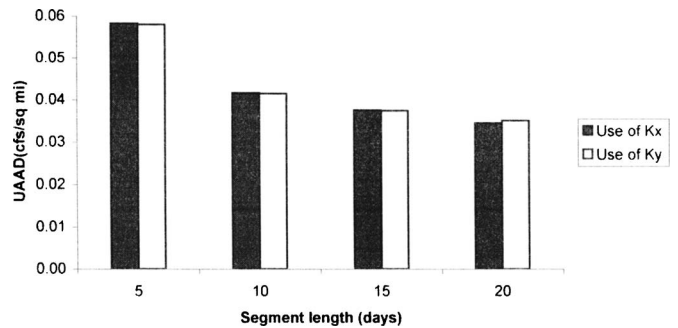


Fig. 6. The impact on UAAD of frequency factor for the recession method, using gauged sites within 200 km, $\rho \geq 0.6$, $D=5$, and $N \geq 1$

allows one to examine the trade-off between segment length and ρ . For instance, if a segment length is 5 days and $\rho \geq 0.9$, the baseflow correlation estimator is nearly as good as when the segment length is 10 days and ρ is between 0.6 and 0.7. These results indicate a large improvement when the segment length increases from 5 to 10 days, but only minimal improvement when the segment length is greater than 10 days. It also shows that the method performance is more sensitive to ρ when the segment length is small. Although having a large ρ produces a better baseflow correlation estimator, these results indicate that only a slight increase in performance is observed when ρ is greater than 0.6 for a segment length of 10 days or more.

Assumption of Equivalent Frequency Factor

One assumption of the baseflow correlation method is that the frequency factor for the low-flow partial-record site, K_y , is equal to the frequency factor for the gauged site, K_x . To investigate this assumption, the jackknife experiment is repeated, but instead of using the frequency factor from the gauged site in Eq. (3), the at-site frequency factor from the low-flow partial-record site (K_y) is employed. Although K_y is not available in practice, this experiment allows us to examine the impact of employing K_x in the baseflow correlation method. The results are shown in Fig. 6, which shows that the assumption of equivalent frequency factors for the gauged site and low-flow partial-record site turns out to be a good one. This result may be due to the log-skew being relatively constant at HCDN sites that are within 200 km of each other.

Assumption of Same Log-Linear Relationships

An additional assumption investigated here is that the log-linear relationship between the 7 day annual minimum flows of the low-flow partial-record site and the gauged site is the same as the log-linear relationship between the baseflow measurements at the low-flow partial-record site and the concurrent baseflows at the gauged site. To investigate this assumption, the estimators of α and β in Eqs. (1) and (2) (a and b) are estimated using all annual minimums and all concurrent baseflow measurements at pairs of sites across the United States. Only sites with at least 40 years of concurrent record within 200 km of each other are employed. The results are shown in Figs. 7 and 8, for a and b , respectively. The x -axis is the estimate using the concurrent baseflow measurements and the y -axis is the estimate using the annual minimums. There is indeed a cloud of points in Figs. 7 and 8. Though some estimates using the annual minimums are different from the esti-

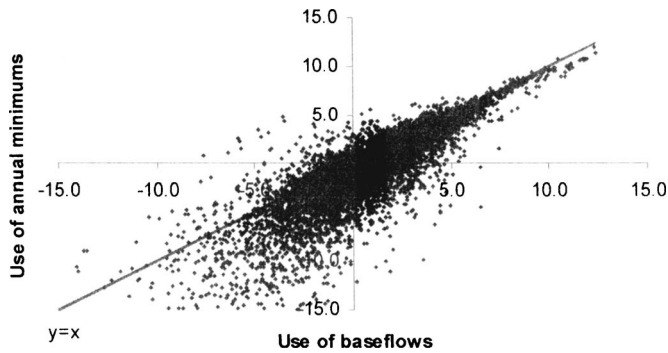


Fig. 7. Estimates of a using annual minimums and concurrent baseflows

377 mates using the concurrent baseflow measurements, most esti-
 378 mates are similar. An investigation (not presented here) shows
 379 that this relationship is not impacted by the strength of correlation
 380 of the baseflows, or by the ratio of watershed characteristics at the
 381 gauged and low-flow partial-record sites, such as the magnitude
 382 of the flows, drainage area, and stream length. In addition, a linear
 383 model was fit in Figs. 7 and 8 by line of organic correlation
 384 (Hirsch and Gilroy 1984). The resulting models were then em-
 385 ployed to estimate a and b for the annual minimums as a function
 386 of a and b from baseflow measurements, and these values were
 387 then used in the baseflow correlation method. Unfortunately, the
 388 baseflow correlation method performed worse using this tech-
 389 nique. This may be due to the fact that the regression line adjusts
 390 the estimates of a too much when the original estimates of a are
 391 close 0, which occurs at many sites. In addition, a and b are in
 392 log-space, and thus the impact of this model adjustment is ampli-
 393 fied when transforming the data back to real space. Further inves-
 394 tigations of this assumption are warranted.

395 Conclusions

396 On the basis of this experiment, the following conclusions of
 397 Reilly and Kroll (2003) are supported:

- 398 • Nearly independent baseflow measurements are needed, as the
 399 consecutive method to obtain baseflow measurements per-
 400 forms much worse than the random and recession methods.
 401 The recession method of picking one baseflow measurement
 402 from consecutive recessions represents a reasonable and prac-
 403 tical way to obtain baseflow measurements.

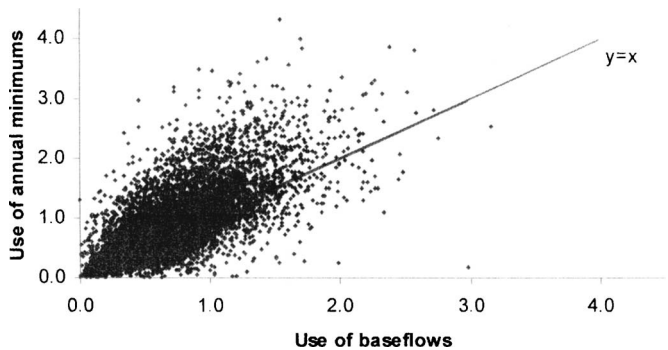


Fig. 8. Estimates of b using annual minimums and concurrent baseflows

- The gauged site selection methods did not have an impact on
 method performance in this experiment. This may be due to
 the sparseness of HCDN gauges in many portions of the
 country. 404
 405
 406
 407
 - The performance of the baseflow correlation method improves
 as the number of baseflow measurements increases. However,
 some leveling off of performance is observed with more than
 ten baseflow measurements. 408
 409
 410
 411
- In addition, this experiment provides guidance on how and when
 412 to take baseflow measurements at low-flow partial-record sites.
 413 The following new conclusions are reached in this investigation. 414
- Unit area performance metrics are less impacted by the mag-
 nitude of $Q_{7,10}$ values compared with metrics measuring rela-
 tive or absolute errors. 415
 416
 417
 - The method performance is most sensitive to the correlation
 coefficient between the baseflows when only five baseflow
 measurements are employed; with more baseflow measure-
 ments the correlation coefficient has less impact on the method
 performance. 418
 419
 420
 421
 422
 - The method performs well if the correlation coefficient is
 greater than or equal to 0.6 for 10 or more baseflow measure-
 ments. With five baseflow measurements, a correlation coeffi-
 cient of at least 0.9 is needed to obtain similar results. 423
 424
 425
 426
 - The baseflow measurements should be obtained during the low
 flow season, defined as July through October in this study. 427
 428
 - Ideally baseflow measurements should be taken as far as possi-
 ble from runoff events, though in general the stream should
 have at least 5 days of continuously decreasing streamflow. 429
 430
 431
 - The assumption of equivalent frequency factors between the
 gauged and low-flow partial-record sites has little impact on
 the baseflow correlation method. 432
 433
 434
 - The assumption of the same log-linear relationship between
 annual minimums and baseflows generally appears reasonable,
 though in some cases wide variations are observed. 435
 436
 437
- One problem with this analysis is that the HCDN river sites are
 438 sparsely distributed across the United States. Further evaluation
 439 of the baseflow correlation method is warranted for regions with a
 440 denser network of river sites. As the metrics used in the current
 441 research are always averaged across the United States and the
 442 baseflow correlation method performance varies in different re-
 443 gions, further regional investigations of the baseflow correlation
 444 method may be valuable. In addition, information from simulta-
 445 neously employing multiple gauged sites may improve the base-
 446 flow correlation method. These issues are being explored by the
 447 writers in further analyses of the baseflow correlation method. 448
 449

Acknowledgments

The writers would like to acknowledge the U.S. Environmental
 450 Protection Agency's Science to Achieve Results (STAR) Program
 451 (Grant No. R825888), the U.S. Geological Survey State Water
 452 Resources Research Institute (WRI) Program (Grant No.
 453 2003NY33G), and the USDA Cooperative State Research, Edu-
 454 cation, and Extension Service (CSREES) Program (Grant No.
 455 NYR-2005-03897) which provided financial assistance to this re-
 456 search. This research has not been subjected to any EPA, USGS,
 457 or USDA review, and therefore does not necessarily reflect the
 458 views of those agencies, and no official endorsement should be
 459 inferred. The writers would also like to thank the two anonymous
 460 reviewers who provided comments that improved this manuscript. 461

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