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A framework for multivariate data-based at-site flood frequency analysis: Essentiality of the conjugal application of parametric and nonparametric approaches

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SUMMARY

In watershed management, flood frequency analysis (FFA) is performed to quantify the risk of flooding at different spatial locations and also to provide guidelines for determining the design periods of flood control structures. The traditional FFA was extensively performed by considering univariate scenario for both at-site and regional estimation of return periods. However, due to inherent mutual dependence of the flood variables or characteristics [i.e., peak flow (P), flood volume (V) and flood duration (D), which are random in nature], analysis has been further extended to multivariate scenario, with some restrictive assumptions. To overcome the assumption of same family of marginal density function for all flood variables, the concept of copula has been introduced. Although, the advancement from univariate to multivariate analyses drew formidable attention to the FFA research community, the basic limitation was that the analyses were performed with the implementation of only parametric family of distributions. The aim of the current study is to emphasize the importance of nonparametric approaches in the field of multivariate FFA; however, the nonparametric distribution may not always be a good-fit and capable of replacing well-implemented multivariate parametric and multivariate copula-based applications. Nevertheless, the potential of obtaining best-fit using nonparametric distributions might be improved because such distributions reproduce the sample's characteristics, resulting in more accurate estimations of the multivariate return period. Hence, the current study shows the importance of conjugating multivariate nonparametric approach with multivariate parametric and copula-based approaches, thereby results in a comprehensive framework for complete at-site FFA. Although the proposed framework is designed for at-site FFA, this approach can also be applied to regional FFA because regional estimations ideally include at-site estimations. The framework is based on the following steps: (i) comprehensive trend analysis to assess nonstationarity in the observed data; (ii) selection of the best-fit univariate marginal distribution with a comprehensive set of parametric and nonparametric distributions for the flood variables; (iii) multivariate frequency analyses with parametric, copula-based and nonparametric approaches; and (iv) estimation of joint and various conditional return periods. The proposed framework for frequency analysis is demonstrated using 110 years of observed data from Allegheny River at Salamanca, New York, USA. The results show that for both univariate and multivariate cases, the nonparametric Gaussian kernel provides the best estimate. Further, we perform FFA for twenty major rivers over continental USA, which shows for seven rivers, all the flood variables followed nonparametric Gaussian kernel; whereas for other rivers, parametric distributions provide the best-fit either for one or two flood variables. Thus the summary of results shows that the nonparametric method cannot substitute the parametric and copula-based approaches, but should be considered during any at-site FFA to provide the broadest choices for best estimation of the flood return periods.

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1. Introduction

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Flood frequency analysis (FFA) defines the severity of a flood event by summarizing the flood variables/characteristics [i.e., peak (P), volume (V) and duration (D)] and by estimating their mutual







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dependence structure. FFA is often required for the planning (Stedinger and Griffis, 2008), design (Haddad and Rahman, 2012), and operation of hydraulic structures (De Michele et al., 2005) and for spatial mapping of the flood risk together with vulnerability and exposure for urban or riverine flood management (Karmakar et al., 2010). FFA approaches can be broadly classified into the following categories (Renard et al., 2013): at-site FFA, climate/weather-informed at-site FFA, historical and paleoflood analyses, regional FFA.

1.1. A brief review of univariate and multivariate approaches

A large body of literature is available for at-site FFA, beginning with Cunnane (1985) and extending to a recent study by Rahman et al. (2013), as shown in Fig. 1, wherein FFA is broadly categorized into three different classes. Initially, FFA was predominantly performed on peak flood flow intensities or flood peaks (Cunnane, 1985; Ahmad et al., 1988; Seckin et al., 2011), but in hydrologic planning and design, it is also important to determine the characteristics of the flood volume (area above the threshold value) and flood duration (length of the flood event) and their joint probabilistic behavior together with flood peak because these flood variables are correlated (Singh and Singh, 1991; Adamson et al., 1999; Yue, 1999, 2001; Karmakar and Simonovic, 2008). Hence, many researchers have performed multivariate FFA. For example, Yue (2001) selected a five-parameter bivariate gamma distribution to describe the joint probability distributions, conditional distributions, and associated joint return periods of two correlated variables. Additionally, Yue et al. (2001) provided a comprehensive review of the bivariate gamma distributions for hydrological applications. Furthermore, Yue and Wang (2004) provided a comparison of two bivariate extreme value distributions, i.e., Gumbel mixed and Gumbel logistic models, in applications to FFA. In the same study, the researchers concluded that these two distributions provided the same joint return periods and may be useful for representing the joint statistical properties of the two random variables. Yue and Rasmussen (2002) and Sandoval (2007) used the bivariate extreme value distribution to determine the joint probability and return period of the correlated flood variables. Apart from above mentioned, numerous studies were conducted using different bivariate distributions (Fig. 1), with restrictive assumptions (Zhang and Singh, 2006).

In most multivariate FFA, it is assumed that the marginal distribution of P, V, and D originates from the same parametric family of statistical distributions (e.g., exponential, gamma, extreme value, lognormal, log-Pearson, etc.), whose probability density function (PDF) is known; thus, the procedure is mathematically straightforward (Adamowski, 1989). Numerous studies have described parametric multivariate analyses (a few recent examples are shown in Fig. 1). However, all of these methods contain certain limitations: (i) the joint distributions among flood peak-volume (P-V), volume-duration (V-D), and peak-duration (P-D) originate from the same family of parametric distributions; (ii) the mathematical formulation becomes more complicated if the number of variables are increased; and (iii) it is not possible to distinguish between the marginal and joint behavior of the variables (Grimaldi and Serinaldi, 2006). However, for natural phenomena, the best-fitting marginal distribution may not follow the same family of distributions (Zhang and Singh, 2006). Hence, to overcome the limitations of traditional multivariate FFA, the concept of a copula is introduced, which relaxes the requirement to select from the same family of marginal distributions for the flood variables.

Copulas are functions that combine the best-fit marginals from the univariate distributions of the flood variables to form multivariate distributions (Salvadori and De Michele, 2004). Hence, this approach provides greater flexibility in choosing the univariate

Past Efforts on Flood Frequency Analysis (FFA)			
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Univariate Parametric FFA: The method is often used to perform FFA	Multivariate FFA		Miscellaneous Procedures for FFA: Apart from different distribution fitting,
based on <i>Peak</i> data. Peak flow is considered as a representative flood characteristic.	Bivariate Parametric FFA: In FFA, it is important to understand the association among the flood volume (V), duration (D) and flood peak (P). The	Copula-based Mixed Bivariate FFA: Past literature shows that the	recently some of the researchers were concentrated on regression-based FFA, hierarchical Bayesian analysis and
Representative literature •GEV (Karim and Chowdhury, 1995, Overlier and Reitan, 2009, Rahman et al., 2013) •Wakeby (Seckin et al., 2011)	bivariate FFA has been applied to estimate the joint distributions of P-V, V-D and P-D pairs and their return periods respectively. Representative literature	copula functions were used to estimate joint distribution and return periods considering mixed marginals from both	functional data analysis to include several factors, such as, sea surface temperature, predicted GCM precipitation, climate indices, snow pack depth, shape of the
 Gumble/EVI distribution (Haktanir 1992, Opere et al., 2006) Log-normal/Gumbel (Haktanir, 1992) LP-3 (Singh et al., 2005; Stedinger and 	•Bivariate Gamma marginals (Yue, 2001; Yue et al., 2001) •Bivariate Extreme Value distribution (Yue and	parametric and non-parametric families.	hydrograph, sample variability, rating curve imprecision, etc. during FFA.
Griffis, 2008; Griffis and Stedinger,2009; Ibrahim and Ahaneku, 2009, Rahman et al., 2013) •Gamma and Generalized logistic (Ghorbani et al., 2010; Afreen and Muhammad, 2012)	Rasmussen, 2002; Sandoval, 2007) •Bivariate Gumbel (Yue et al., 1999; Yue and Wang, 2004; Shiau, 2003) •Log-normal Marginals (Yue, 2000) •Exponential Marginals (Choulakian et al., 1990) •Bivariate Normal (Yue, 1999)	Representative literature •Karmakar and Simonovic, (2009)	Representative literature: •Yue et al., (2002); Kwon et al., (2008); Overleir and Reitan, (2009); Gaal et al., (2010); Haddad et al., (2012); Chebana et al., (2012).
•Log-logistic (Ahmad et al., 1988) Univariate Non-Parametric FFA: To overcome some of the limitations of	•Meta-Gaussian (Kelly and Krzysztofowicz, 1997) Copula-based FFA: The conventional parametric	Trivariate FFA: Here the associations among all the three variables (P-V-D) were	Nonstationary FFA: In conventional FFA, independence and stationarity are necessary assumptions.
parametric methods, nonparametric estimations have been implemented by many researchers for FFA.	bivariate FFA assumes that the marginals for P, V and D follow a specific distribution from same family. The use of copula relaxes this assumption by selecting the marginals from different best-fit distributions and consequently determines the return period.	estimated using a trivariate joint distribution and consequent prediction of return period were carried out. All past attempts have been made on parametric	However, due to perceivable impact of climate change, land use pattern and urbanization, these assumptions would no longer valid, hence resulting
Representative literature •Kernel density (Adamowski, 1989; Bardsley, 1989; Adamowski and Feluch, 1990; Moon and Lall, 1994; Adamowski et al., 1998; Adamowski 2000)	Representative literature •Archimedean copulas (Genest and MacKay, 1986;	and copula based distributions. Representative literature	nonstationarity in time series. Representative literature:
Adamowski, 2000) •Non-parametric Bayesian (O'Connell, 2005) •Non-parametric orthonormal (Karmakar and Simonovic, 2008)	Favre et al., 2004; Salvadori, 2004; Salvadori and De Michele, 2004; Zhang and Singh, 2006; Salvadori and De Michele, 2007) •Meatelliptical copulas (Genest et al., 2007) •Gussian copula (Renard and Lang, 2007)	•Sandoval and Villasenor, (1994); Sandoval, (1998); Grimaldi and Serinaldi, (2006); Zhang and Singh, (2007); Sandoval and Villasenor, (2008).	•Khaliq et al., (2006); Villarini et al., (2009); Strupczewski et al.,, (2009); Villarini et al., (2011); Ouarda and El-Adlouni, (2011); Gilroy and McCuen, (2012); Seidou et al., (2012); Bender et al., (2014)

Fig. 1. Past efforts on flood frequency analysis (FFA), showing different classes of frequency analyses with representative literature. (See above-mentioned references for further information.)

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