

# Hydrological analysis and investigation of the Aaos (Vjosë) - Voidomatis hydrosystem, Greece

Leontaritis A.D.<sup>1</sup> and Baltas E.<sup>2</sup>

<sup>1</sup> *Metsovion Interdisciplinary Research Center of the National Technical University of Athens, Metsovion 44200, Greece,*

[leontari@mail.ntua.gr](mailto:leontari@mail.ntua.gr)

<sup>2</sup> *School of Civil Engineering, Department of Water Resources and Environment, Laboratory of Hydrology and Water Resources Management, National Technical University of Athens, 5, IroonPolytechniou, 157 73 Athens, Greece, [baltas@chi.civil.ntua.gr](mailto:baltas@chi.civil.ntua.gr)*

## Abstract

The main goal of this work is to create a solid background for the further study of the Aaos-Voidomatis hydrosystem. Aaos is the only upstream transboundary river of Greece, and together with its main tributary, Voidomatis, form a unique, protected ecosystem. Currently, there are no hydrological studies in the literature following the construction of the Aaos Springs dam in 1988, or any work to summarize its hydrological and geographical characteristics. This paper examines the extension and supplementation of available hydrological data in several measuring stations of the hydrosystem between 1950 and 2012. The original historical time-series data set created can be used in future studies as primary data prior to and after the construction of the dam. Furthermore, this study provides a forecast for the monthly water flow rates of the river in the future through the creation of a 50-year long synthetic time-series. As a result, the risk of failure to cover existing water needs was estimated. The methodology developed for the supplementation of the historical time-series, as well as, the creation of the synthetic time-series, includes some simple linear regression models and the autoregressive forecast models AR(1) and AR(2). These models were applied to the available data with statistically significant results. The main conclusion that derived from this study is that current water flow rates can sufficiently cover the existing water needs; mainly for the irrigation of the Konitsa plain. However, it was proven that the hydrosystem is sensible to climate fluctuations and/or water use.

**Keywords:** Aaos; Vjosë; Voidomatis; hydrological analysis; autoregressive models; water sufficiency

## 1. Introduction

Aoos is a major Greek river with a series of interesting characteristics that form the main reasons for its selection as a case study. It is the only upstream transboundary river of Greece (Eleftheriadou and Mylopoulos 2005), and even though there is no regulation between Greece and Albania regarding the water use from the Greek side of the border, its management remains an important issue. Its total catchment area covers 6519 km<sup>2</sup>; 67% or 4365 km<sup>2</sup> belongs to Albania while the remaining 33% or 2154 km<sup>2</sup> belongs to Greece; its discharge is estimated at 2154 hm<sup>3</sup> on the Greek-Albanian borders, and at 5550 hm<sup>3</sup> at its estuary in the Adriatic Sea (Ecologic Institute 2012). As shown in figure 1, Aoos and its main tributary, Voidomatis, are part of a unique, protected ecosystem (North Pindos National Park, Pindos State Park, Vikos-Aoos State Park, several Natura 2000 areas). The relief and the geology of the wider area is diverse with extended mountainous regions, plateaus and lowlands, affecting significantly the microclimate and the respective precipitation in each sub-basin, as well as its hydrological response throughout the year. Moreover, the hydrosystem has been affected by the Aoos Springs reservoir, as it will be analyzed next, while there are plans for further capitalizing from its high hydrodynamic potential. Currently, there is no work summarizing its hydrological and geographical characteristics or any hydrological study in the literature following the construction of the Aoos Springs dam in 1988. Therefore, any studies using the available hydrological data incorporate a significant error: the absence of the effect of the dam on the water flow of the river. Accordingly, the estimation of the water flow reduction in the different sections of the river after 1988 is of great importance, while on the other hand it constitutes a measure for the environmental impact of the dam. This work examines the extension and supplementation of available hydrological data in all the measuring stations of the hydrosystem for the period between 1950 and 2012. Furthermore, a forecast for the monthly water flow rates of the river in the future is provided through the creation of a 50-year long synthetic time-series using the autoregressive models AR(1) and AR(2) and the risk of failure to cover the existing water needs is estimated.

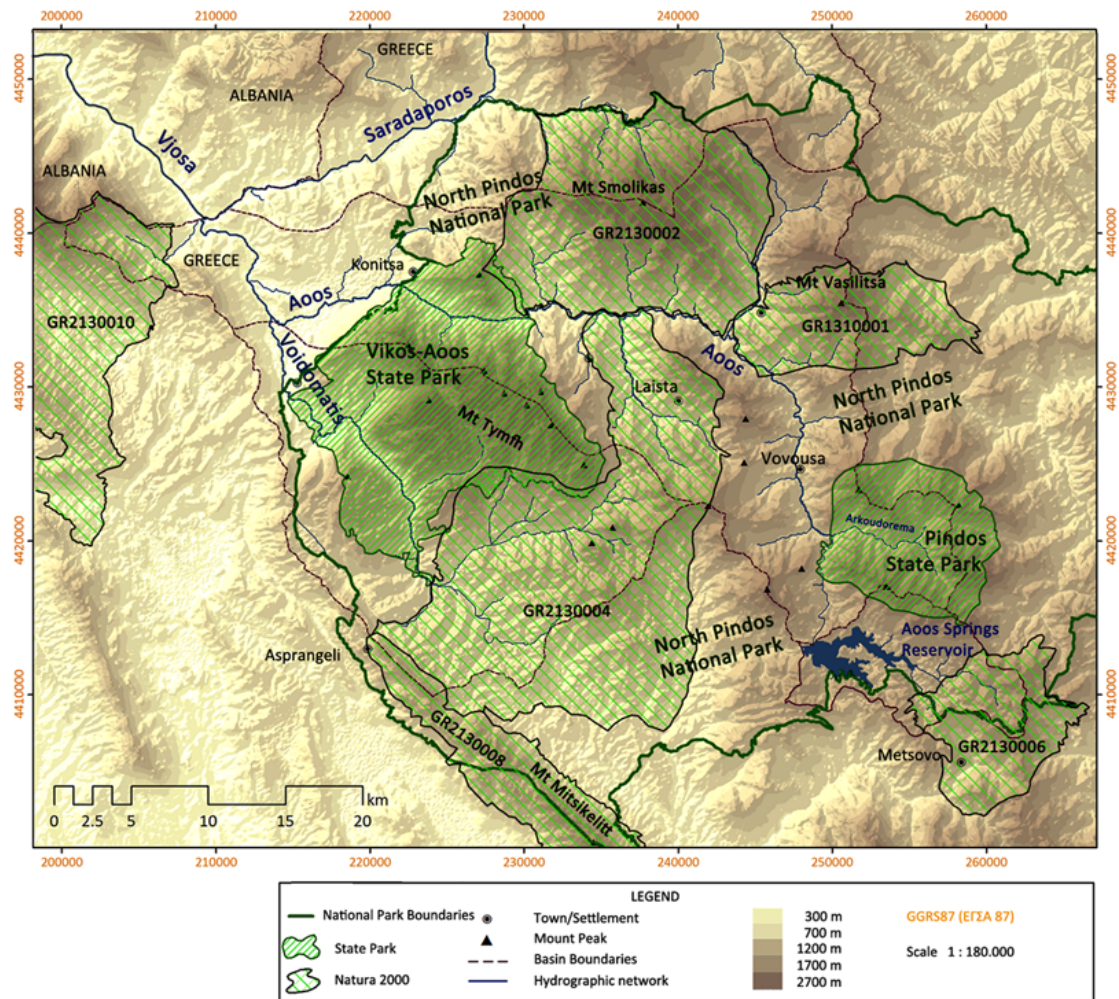


Figure 1. Map of Protected Areas

## 2. Materials and Methods

### 2.1 Data acquisition

The hydrological data used in this study concern monthly water flow time series and were acquired from the archives of the Greek Public Power Corporation. The hydrosystem and the sites of the available hydrometric stations are presented in figure 2. As already discussed, the construction of the Aios Springs dam (hydrological year 1987-88) has significantly affected the hydrosystem as shown in figure 3, which presents the discharge of Aios in Vovousa Bridge, just a few kilometres downstream from the dam. Thus, the hydrological study has been divided into two periods: **A)** from hydrological year 1950-51 until 1986-87 and **B)** from 1988-1989 until 2011-2012. Hydrological year 1987-88 is not included in the study as a

transient year between period A and B. It is noted that the water withheld in the reservoir is entirely diverted through the Aaos Springs Hydroelectric Plant to the basin of Arachthos hydrosystem and since there is no ecological discharge downstream from the dam, the sub-catchment area of the reservoir is abstracted in study period B.

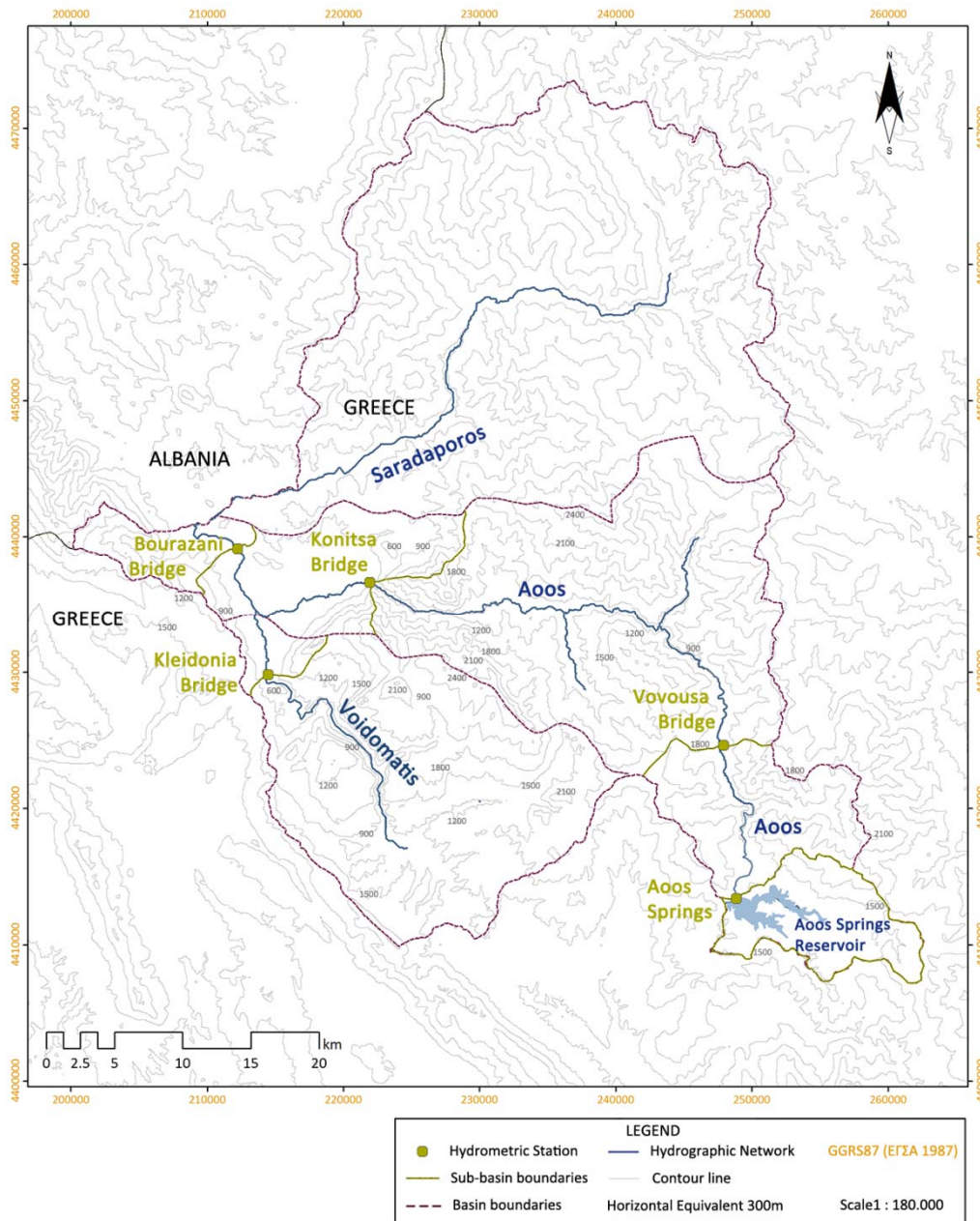


Figure 2. Aaos-Voidomatis hydrosystem and hydrometric stations

Table 1 summarises the catchment areas of all sub-basins for the two periods and table 2 shows the respective average annual discharges according to the available data. At any case, the abstracted catchment area should not affect the qualitative response of the system. This is

confirmed in Figure 3, where as it can be seen, the monthly response of the system follows a similar pattern in the two periods, suggesting a high reliability of the available hydrological data. Additionally, in Figure 3b the infamous drought of the early 90s as well as the wet period of 2003-2006 is clearly visible.

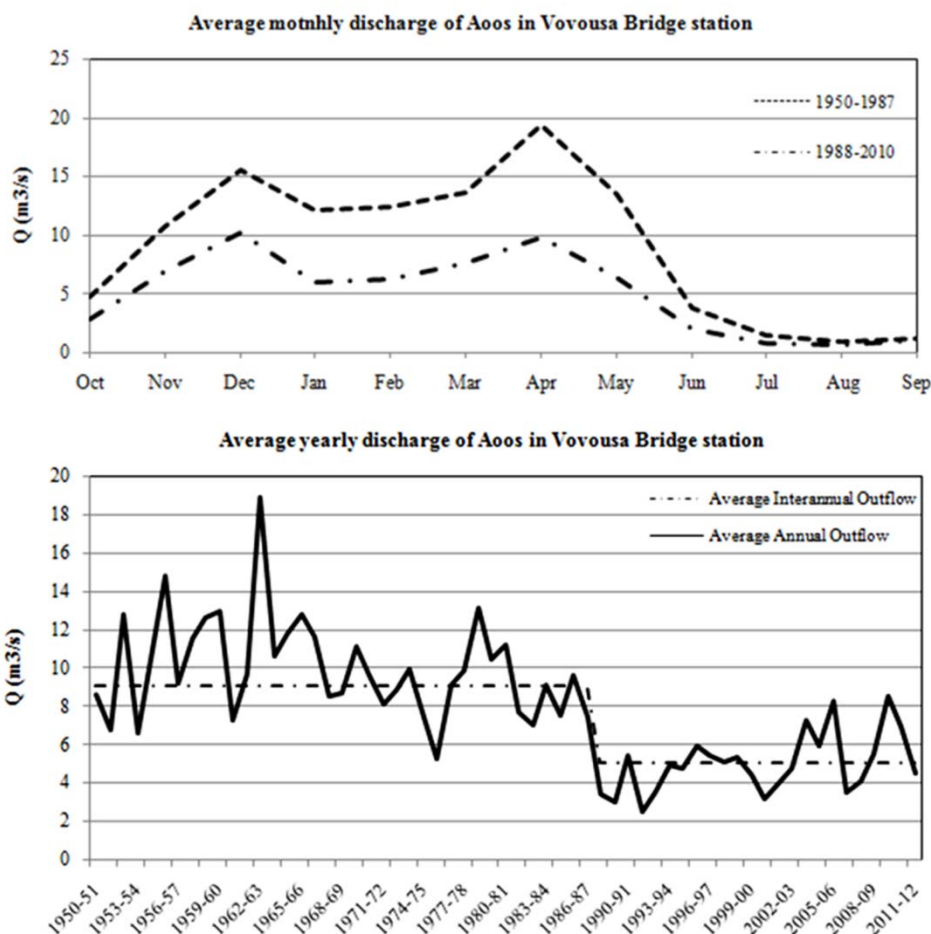


Figure 3. Effect of the Aaos Springs dam (1988) on the a. Monthly discharge and b. Annual discharge of Aaos River in Vovousa Bridge

Table 1. Hydrometric stations and catchment area of their respective sub-basins prior to (A) and (B) after the construction of the Aaos Springs Reservoir

HYDROMETRIC STATIONS (Source: Public Power Corporation)			
Site	Basin	Catchment Area A (km <sup>2</sup> )	Catchment Area B (km <sup>2</sup> )
Aaos Springs	Aaos	79.1	85.3
Vovousa Bridge	Aaos	202	116.7
Konitsa Bridge	Aaos	665	579.7
Kleidonia Bridge	Voidomatis	332	332
Bourazani Bridge	Aaos-Voidomatis	1154.6	1069.3

Table 2. Average annual discharge of the hydrometric station prior to (1967-77) and after (1991-2006) the construction of Aaos Springs Reservoir

Average Annual Discharge		
Hydrometric Station	Q <sub>A</sub> (m <sup>3</sup> /s)	Q <sub>B</sub> (m <sup>3</sup> /s)
Aaos Springs	3.52	3.3
Vovoussa Bridge	9.23	5.03
Konitsa Bridge	22.88	N.A.
Kleidonia Bridge	14.88	N.A.
Bourazani Bridge	46.66	N.A.

Table 3. Available Hydrological Data

Hydr. Year	Aaos Springs	Vovoussa Bridge	Konitsa Brisege	Kleidonia Bridge	Bourazani Bridge
1950-51	•	X	X	X	X
1951-52	•	X	X	X	X
1952-53	•	X	X	X	X
1953-54	•	X	X	X	X
1954-55	•	X	X	X	X
1955-56	•	X	X	X	X
1956-57	•	X	X	X	X
1957-58	•	X	X	X	X
1958-59	•	X	X	X	X
1959-60	•	X	X	X	X
1960-61	•	X	X	X	X
1961-62	•	X	X	X	X
1962-63	•	X	X	X	X
1963-64	•	X	•	X	X
1964-65	•	X	•	X	X
1965-66	•	X	•	X	X
1966-67	•	•	•	X	X
1967-68	•	•	•	•	X
1968-69	•	•	•	•	X
1969-70	•	•	•	•	X
1970-71	•	•	•	•	X
1971-72	•	•	•	•	X
1972-73	•	•	•	•	X
1973-74	•	•	•	•	•
1974-75	•	•	•	•	•
1975-76	•	•	•	•	•
1976-77	•	•	•	•	•
1977-78	X	•	X	X	X
1978-79	X	•	X	X	X
1979-80	X	•	X	X	•
1980-81	X	•	X	X	•

1981-82	X	X	X	X	•
1982-83	X	X	X	X	•
1983-84	X	•	X	X	X
1984-85	X	•	X	X	X
1985-86	X	•	X	X	X
<b>1986-87</b>	X	•	X	X	X
1988-89	X	•	X	X	X
1989-90	X	•	X	X	X
1990-91	X	•	X	X	X
1991-92	•	•	X	X	X
1992-93	•	•	X	X	X
1993-94	•	•	X	X	X
1994-95	•	•	X	X	X
1995-96	•	•	X	X	X
1996-97	•	•	X	X	X
1997-98	•	•	X	X	X
1998-99	•	•	X	X	X
1999-00	•	•	X	X	X
2000-01	•	•	X	X	X
2001-02	•	•	X	X	X
2002-03	•	•	X	X	X
2003-04	•	•	X	X	X
2004-05	•	•	X	X	X
2005-06	•	•	X	X	X
2006-07	•	X	X	X	X
2007-08	•	X	X	X	X
2008-09	•	X	X	X	X
2009-10	•	X	X	X	X
2010-11	•	X	X	X	X
2011-12	•	X	X	X	X

## 2.2 Supplementation of the historic time series

The basic elaboration of the available hydrological data concerned the identification of any outliers and, if necessary, their correction. The availability of measurements in different sections of the river enables the attribution of these outliers to flooding incidents or to errors. Errors can be induced both through the measuring procedure and through the elaboration of the initial hydrometric data (i.e. construction of level-flow curves). Having tracked any error-outlier value, it can be corrected with a simple linear regression model applied on the values of the flow in suitable hydrometric stations.

The supplementation of the missing data is also based on simple linear regression models. The most important aspect is the selection of suitable stations to apply the modelling procedure. First of all, the two stations should have available data for a sufficient common period. The other decisive parameter is the level of correlation between the two time-series for the common period (n), expressed by the correlation factor (r). The two time series are considered as suitable for the application of a simple linear regression model when

$$r \geq \frac{2}{\sqrt{n}}$$

In various occasions, the supplementation of the data of a hydrometric station required different linear regression models with more than one hydrometric station (e.g. for the wet and for the dry period). Figure 4 shows a qualitative approach of the level of correlation between different hydrometric stations. The similarity of the curves of the different stations indicates a high correlation between them. Naturally there are some deviations but overall the available data were sufficiently correlated for the complete supplementation of the historic time-series.

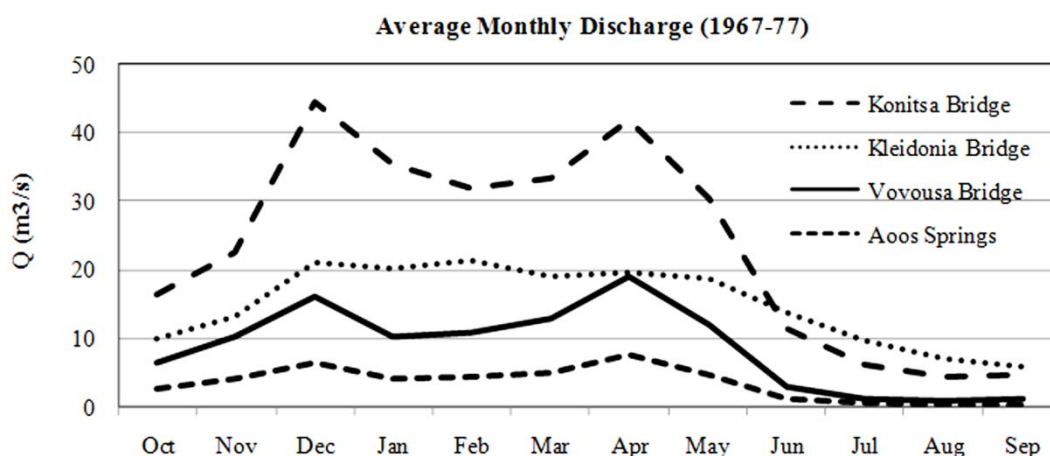


Figure 4. Average monthly discharge in different stations (1967-1977)

Regarding period B, a slightly different approach is needed in order to take into consideration the effect of the dam. As it can be seen in table 3, after 1987 there are available data only for the stations “Aaos Springs” (calculated from the reservoir water flow equilibrium) and “Vovousa Bridge”. Since no measurements have been made in the rest of the stations for this

whole period, no regression model can be directly applied for the supplementation of the missing data. In order to overcome this issue, the concept of the naturalised flow of the river was used. Practically the naturalised flow of the river is the theoretical natural flow of the river without the effect of the dam. Thus, the naturalised flow of the river for “Vovoussa Bridge” station is calculated as the sum of the real flow at this point and the water flow withheld in the reservoir (“Aaos Springs” hydrometric station). Subsequently, these values can be used as input for the models used for period A and the naturalised monthly flow in all stations can be calculated for the whole period; the real flow is calculated by subtracting the respective monthly flow in “Aaos Springs” station. “Kleidonia Bridge” station is an exception as it is located on Voidomatis River and is not affected by the dam.

### **2.3 Creation of synthetic time-series – the autoregressive models AR(1) and AR (2)**

In order to forecast the monthly water flow rates of the river it is necessary to create a synthetic time-series using a mathematical model. Since it is impossible to define all the parameters that influence physical quantities such as the water flow of a river, it is very difficult to describe their evolution with a deterministic model. On the other hand, their strong dependence on non-deterministic parameters (e.g. weather) makes possible their description over time through a stochastic model that can estimate the probability distribution of their values (Fortin et al. 2004). According to Hipel (1985) in some cases of hydrological time series modelling a simple stochastic model may yield better results than a more complex deterministic model (Lohani et al. 2012). Stochastic models are used to generate synthetic hydrologic time series for the planning and management of water resource systems (Loucks et al. 1981; Bras and Rodrigues-Iturbe 1985; Salas 1993; Thyer et al. 2006).

During the last few decades, several types of stochastic models have been developed and proposed (Yevjevich 1972; Salas and Smith 1981; Stedinger and Taylor 1982) for modelling hydrological time series and generating synthetic stream flows (Lohani et al. 2012). These models are called system theoretic transfer function models because they attempt to establish

a linkage between several phenomena without internal description of the physical processes involved (Lohani et al. 2012). Broadly, the stochastic models are classified as Autoregressive Moving Average (ARMA) models (Box and Jenkins 1970), disaggregation models (Valencia and Schaake 1973), and models based on the concept of pattern recognition (Panu and Unny 1980). The forecast of the flow of rivers has been successfully achieved in several cases using linear models, such as AR, ARMAX, and Kalman filter (Burn and McBean 1985; Awwad et al. 1994; El-Fandy et al. 1994; Castellano-Méndez et al. 2004).

.For this study the autoregressive models AR(1) and AR(2) were selected. AR models incorporate solely autoregressive terms for modelling univariate time series and are used for the analysis of systems which have a generally linear input–output relationship (Chua and Wongl 2011). AR models represent a time series  $Z_t$  as a linear function of its previous values ( $Z_{t-1}, Z_{t-2}, \dots$ ) and a random shock series ( $a_t, a_{t-1}, a_{t-2}, \dots$ ), an uncorrelated Gaussian random variable, with zero mean and constant variance (Thyer et al. 2006). The random shock series represent the stochastic error terms of the model and are assumed to be normally distributed random variables with zero mean and constant variance.

The AR(1) and AR(2) models have the following form:

$$Z_t = \Phi_1 \cdot Z_{t-1} + a_t \quad \text{AR(1)}$$

$$Z_t = \Phi_1 \cdot Z_{t-1} + \Phi_2 \cdot Z_{t-2} + a_t \quad \text{AR(2)}$$

where  $\Phi_1$  and  $\Phi_2$  are the lag-one and lag-two autoregressive factors

According to Box and Jenkins (1970) the development of an AR model requires at least a 50 event long time series but for strongly varying physical quantities this number should be even greater. At any case, the 62 year long historic time-series of monthly water flow corresponds to a total of  $62 \cdot 12 = 644$  events which can be considered as sufficiently long. Moreover, AR models can only be applied on stationary time-series and therefore the historic water flow time series need to be transformed accordingly. The conversion of the historic monthly

discharge time-series to stationary is achieved through the following procedure: the monthly mean value is abstracted from each value and then it is divided by the respective standard deviation. This way any systematic variation of the values is abstracted and the stationary time-series contains only the stochastic part of the original historic time series. This procedure is expressed by the following equation:

$$x_1(i,j) = \frac{[X(i,j) - \bar{X}(j)]}{\sigma_z(j)}$$

where

$x_1(i,j)$       *the monthly stationary value of discharge*

$X(i,j)$       *the real historic monthly discharge*

$j$               *the order of the month,*

$i$               *the order of the year,*

$\bar{X}(j)$          *the real historic mean value of discharge for month j*

$\sigma_z(j)$       *the real standard deviation for month j.*

Following the application of the auto regression models AR(1) and AR(2) two distinct 50 year long **stationary** synthetic time series of monthly discharges are created for each model. Therefore, the stationary synthetic time series need to be converted to “real” synthetic time-series following the opposite procedure to the one describe above. The stationary synthetic time series is divided into 12 classes (j) consisting of 50 values each that correspond to a single month. The classification can be done randomly as the time-series is stationary. Subsequently, the value i of each class j is converted using the equation:

$$X(i,j) = x_1(i,j) \cdot \sigma_z(j) + \bar{X}(j)$$

where

$x_1(i,j)$       *the synthetic stationary value of the monthly discharge*

$X(i,j)$       *the “real” synthetic monthly discharge*

$j$	<i>the order of the month (class)</i>
$i$	<i>the order of the year</i>
$\bar{X}(j)$	<i>the real historic mean value of discharge for month <math>j</math></i>
$\sigma_z(j)$	<i>the real standard deviation for month <math>j</math>.</i>

## 2.4 Risk of failure

The N-year long synthetic time-series at a specific location of the river can be used to estimate the risk of non adequacy for covering existing water needs like water supply, irrigation or hydroelectric power generation in the future (Lohani et al. 2012). Thus it is first necessary to exactly determine these needs at a monthly basis and then the risk can be estimated by using the following simple procedure:

The monthly discharge values  $Q(i,j)$  of the synthetic time series, where  $i$  denotes the order of the year, are sorted for each month  $j$  in descending order. The respective total monthly water needs are  $A(j)$ . Assuming for example month  $j$ , the cases for which  $Q(i,j) < A(j)$  are counted and  $m$  is the total number of these events. If  $N$  the length of the synthetic time series in years, then the possibility (risk) of not covering the water needs is calculated as:

$$P\{Q(i,j) \leq A(j)\} = \frac{m}{N + 1}$$

## 3. Results and Discussion

The supplemented historic time-series for the periods prior to and after the construction of the dam is summarized in table 4. It is reminded that the naturalized flow of the river ( $B_n$ ) is defined as the theoretical natural flow of the river without the effect of the dam.

It is also noted that the discharge during period B is reduced compared to period A. This is due to two factors. Firstly, to the dam of Aoos Springs, which is particularly evident at all locations and especially at "Vovoussa Bridge" and "Konitsa Bridge" where as expected the discharge in period B is significantly reduced. Secondly, this reduction is owed to reduced

precipitation. This is confirmed by comparing average annual discharges during the two periods, at the locations that are not affected by the dam ("Aaos Springs", "Kleidonia Bridge"). The same conclusion can be drawn by comparing the average annual discharge for period A with the average naturalized annual discharge for period B, at the locations of "Vovoussa Bridge", "Konitsa Bridge" and "Bourazani Bridge". In the latter station the difference of the discharge between the two periods is greater than in the other stations, which could be attributed to its larger catchment area. However, due to the limited availability of primary data for this station it needs to be investigated.

Table 4. Average monthly discharge prior to (A) and after (B) the construction of the Aaos Springs Reservoir

Average monthly water flow for the period 1950-2012 (m <sup>3</sup> /s)														
Station	Period	OCT	NOV	DEC	JAN	FEB	MAR	APR	MAY	JUN	JUL	AUG	SEP	YEAR
Aaos Springs	A	1.83	3.96	5.78	4.88	5.01	5.32	7.52	5.27	1.51	0.52	0.32	0.43	3.52
	B	2.15	4.96	6.93	4.24	4.54	6.06	6.54	3.03	0.83	0.39	0.34	0.7	3.39
Vovoussa Bridge	A	4.77	10.78	15.54	12.21	12.44	13.63	19.44	13.5	3.84	1.43	0.92	1.2	9.12
	Bn*	4.91	11.84	17.17	10.22	10.77	13.71	16.38	9.48	2.78	1.17	0.94	1.67	8.41
	B	2.75	6.88	10.24	5.98	6.23	7.65	9.83	6.44	1.95	0.79	0.6	0.97	5.02
Konitsa Bridge	A	12.89	22.45	40.69	31.27	35.66	34.98	42.01	33.31	13.74	5.95	4.07	4.49	23.39
	Bn*	13.13	26.46	46.78	28.33	31.29	35.3	35.98	24.59	10.76	4.88	4.33	5.56	22.24
	B	11	21.5	39.85	24.08	26.75	29.24	29.44	21.56	9.93	4.49	3.99	4.86	18.85
Kleidonia Bridge	A	8.27	13.15	20.08	21.29	22.9	20.68	19.77	19.82	15.4	9.01	6.19	5.58	15.13
	B	8.38	15.2	21.91	20.19	21.2	20.84	17.64	16.6	11.52	7.03	6.25	6.62	14.41
Bourazani Bridge	A	22.8	53	64.26	64.48	60.58	68.55	72.43	59.25	30.71	18.54	13.52	12.38	44.95
	Bn*	27.08	53.08	74.97	58.11	54.5	63.78	61.08	44.21	25.81	15.66	14.31	14.16	42.17
	B	24.95	48.12	68.05	53.87	49.96	57.72	54.54	41.17	24.98	15.28	13.97	13.46	38.78

(A=1950 – 1987, B =1988-2012, Bn\*= naturalized flow for period B)

A further analysis of the results in "Bourazani Bridge" not only provides their verification but also indicates the reliability of the used model. The "Bourazani Bridge" is located a few kilometres downstream of the confluence of Aaos and Voidomatis River, while the locations of "Konitsa Bridge" and "Kleidonia Bridge" are located at a short distance upstream from the above confluence, in Aaos and Voidomatis, respectively (Fig. 2). The data verification consists of the following steps; firstly the sum of the annual discharge and basin area of "Kleidonia Bridge" and "Konitsa Bridge" are calculated for both periods A, B. Subsequently,

the respective ratios "Bourazani Bridge"/Sum are calculated. The data and the results of the verification are given in table 5.

*Table 5. Verification of the results at "Bourazani Bridge"*

	<b>Konitsa Bridge</b>	<b>Kleidonia Bridge</b>	Sum	<b>Bourazani Bridge</b>	<b>Ratio (Bourazani /Sum)</b>
Basin A (km <sup>2</sup> )	665	332	997	1154.6	<b>1.158</b>
Basin B (km <sup>2</sup> )	579.7	332	911.7	1069.3	<b>1.173</b>
Discharge A (m <sup>3</sup> /s)	23.39	15.13	38.52	44.95	<b>1.167</b>
Discharge B (m <sup>3</sup> /s)	18.85	14.41	33.26	38.78	<b>1.166</b>

As shown in the table, these two ratios have about the same value for both periods A and B. This confirms the initial hypothesis that the greater discharge differences between the two periods in this location, in relation to other locations, is due to the greater basin area and not to model fault or wrong primary data. The above proportions remain the same for period B, during which there are no primary data at the above locations. This shows the accuracy of the simple linear regression model. It is noted that this procedure could be applied for these locations as the catchment area between the confluence of the two rivers and the location of "Bourazani Bridge" has similar characteristics with the rest catchment area of Aaos upstream from "Konitsa Bridge" and "Kleidonia Bridge": a combination of high mountains up to 2500m and semi-mountainous regions at an altitude ranging from 300 to 700m. However, this is not the case for the remaining locations due to the significantly heterogeneous characteristics of the basins. For example, the basin ratio for the locations of "Konitsa Bridge" and "Vovoussa Bridge" is equal to 5.7, while the average ratio of annual discharge is equal to 3.5. The average annual rainfall in "Konitsa Bridge" (mixed catchment area) is about 1000mm, while in "Vovoussa Bridge" (a purely mountainous catchment area) 1400mm.

### **3.1 Irrigation needs and water availability**

The main water needs that are covered by Aaos and Voidomatis concern the irrigation of the plains of Konitsa (11500 hectares irrigated by Aaos River) and Kleidonia (4500 hectares irrigated by Voidomatis River). The water intake takes place upflow from "Konitsa Bridge"

and "Kleidonia Bridge" respectively (Fig. 5). Other water uses include the water supply of several minor villages through small streams and springs of the wider basin. The overall quantities are practically insignificant and are therefore neglected in this study. The monthly needs in irrigation water for the two plains have been estimated in the study for the construction of the new irrigation system of the greater Konitsa area (Prefecture of Epirus 2008). Table 6 summarizes the irrigation monthly needs (May-September) and the respective average available discharges as well as the standard deviation for the two locations as calculated in the first part of this study. For "Kleidonia Bridge" that is not affected by the dam the discharge refers to the whole study period (1950-2012) while for "Konitsa Bridge" only to period B (1988-2012).



Figure 5. Map of irrigated areas within the Konitsa-Kleidonia Plain

In "Kleidonia Bridge" the water needs are low compared to the discharge of Voidomatis River and thus there is practically no risk of failure and no further studying is needed. However, in "Konitsa Bridge" during July, August and September the irrigation needs are greater, and the

discharge of Aaos combined with the relatively high standard deviation indicate a possibility of failure, which will be further studied next.

*Table 6. Irrigation Needs and Water Availability at "Konitsa Bridge" (1988-2012) and "Kleidonia Bridge" (1950-2012)*

<b>Irrigation Needs and Water Availability</b>						
	(m <sup>3</sup> /s)	May	June	July	August	September
Konitsa Bridge	Irrigation	0.381	0.59	0.7	0.61	0.4
	Discharge	21.56	9.93	4.49	3.99	4.86
	S.D.	7.38	3.26	0.92	1.17	2.97
Kleidonia Bridge	Irrigation	0.149	0.232	0.273	0.238	0.155
	Discharge	18.97	13.8	8.69	6.66	6.23
	S.D	4.34	4.34	2.75	2.03	3.2

### 3.2 Synthetic Time Series

For the reasons explained in the above paragraph the synthetic time-series will be created only for "Konitsa Bridge". However, since the forecast models refer to future discharges of the river the effect of the dam should be taken into consideration and thus the historic time series needs to be modified in order to be entirely used as an input for the models: all discharge values for period A need to be deducted to period B (denaturalised flow). This is achieved simply by abstracting the discharge in "Aaos Springs" from the respective value in "Konitsa Bridge". The result is a hypothetical historic denaturalised monthly discharge time-series in "Konitsa Bridge" (Tab. 7) as if the dam existed since 1950.

*Table 7. Historic denaturalised monthly discharge time-series at "Konitsa Bridge" (1950-2012)*

	OCT	NOV	DEC	JAN	FEB	MAR	APR	MAY	JUN	JUL	AUG	SEP	YEAR
Q (m <sup>3</sup> /s)	11.02	20.14	37.6	25.14	28.52	29.43	31.73	24.51	10.98	4.92	3.88	4.49	<b>19.32</b>
σ (m <sup>3</sup> /s)	9.2	13.13	22.44	9.9	11.5	11.89	9.45	9.25	3.77	1.27	0.97	2.3	<b>4.19</b>

The results of the synthetic-series calculations are summarized in table 8. It is noted, that as expected, the average monthly discharges as well as the respective standard deviations are similar to the values of the historic time series for both models and thus there is no need to use higher class (>2) AR models.

Table 8. Results of the synthetic time-series calculations at "Konitsa Bridge" using the forecast models AR(1) and AR(2)

Synthetic Time-Series												
	OCT	NOV	DEC	JAN	FEB	MAR	APR	MAY	JUN	JUL	AUG	SEP
<b>Historic time-series</b>												
(m <sup>3</sup> /s)	11.02	20.14	37.6	25.14	28.52	29.43	31.73	24.51	10.98	4.92	3.88	4.49
$\sigma$ (m <sup>3</sup> /s)	9.2	13.13	22.44	9.9	11.5	11.89	9.45	9.25	3.77	1.27	0.97	2.3
<b>AR(1)</b>												
(m <sup>3</sup> /s)	10.87	23.16	34.95	25.84	27.54	31.27	34.16	24.69	10.87	4.94	4.2	4.67
$\sigma$ (m <sup>3</sup> /s)	9.12	13.21	21.77	8.58	12.29	13.04	8.32	9.23	3.59	1.21	0.94	2.02
<b>AR(2)</b>												
(m <sup>3</sup> /s)	10.51	21.32	33.67	25.25	27.86	29	31.09	26.61	11.86	4.52	3.77	4.57
$\sigma$ (m <sup>3</sup> /s)	7.55	12.89	21.08	8.64	11.37	12.4	8.6	10.22	4.2	1.34	1.11	1.85

### 3.3 Risk of failure

Following the creation of the synthetic time series of monthly discharges, the risk of non adequacy for covering irrigation needs at "Konitsa Bridge" was calculated using the procedure described in section 3.3. The procedure was applied only for the months of July, August and September for the reasons analyzed above. The results of the procedure for the two synthetic time series are shown in table 9.

A miss is observed in September, once in 50 years, in both synthetic time series. This corresponds to a **1.96% risk**. At this point it is important to emphasize that special care is required when interpreting these results. It is reminded that irrigation water is abstracted through technical work at a location about 1 km upstream of the hydrometric station of "Konitsa Bridge" (Fig. 4). However, there are no quantitative data regarding this irrigation project. Its operation began before 1963, when the first measurements were taken, and therefore its effect on monthly discharges during the irrigation period is incorporated in the available measurements. In any case, the irrigation needs are fully covered by this project to date. Additionally, the construction of a new modern project is scheduled for the irrigation of the entire area of Konitsa from the rivers Aooos and Voidomatis. The water transfer will be

more efficient, reducing the total losses and therefore the water amount that is needed. Thus, the risk should be redefined qualitatively.

Table 9. Results of risk calculation

Procedure of risk calculation (m <sup>3</sup> /s)						
YEAR	AR(1)			AR(2)		
	JUL	AUG	SEPT	JUL	AUG	SEPT
<b>Irrigation</b>	0.7	0.61	0.4	0.7	0.61	0.4
<b>1</b>	7.54	6.46	8.51	8.74	6.11	8.49
<b>2</b>	7.14	6.17	8.19	7	5.93	7.93
<b>3</b>	6.99	5.67	7.57	6.52	5.67	7.53
<b>4</b>	6.62	5.59	7.5	6.51	5.14	7.49
<b>5</b>	6.42	5.28	7.37	6.25	5.01	7.36
<b>6</b>	6.37	5.11	7.34	6.1	4.88	7.2
<b>7</b>	6.27	5.05	7.16	6.05	4.87	7.05
<b>8</b>	6.26	4.99	6.98	5.96	4.79	6.48
<b>9</b>	6.2	4.95	6.95	5.86	4.79	6.1
<b>10</b>	6.13	4.92	6.81	5.71	4.69	6.02
...						
<b>41</b>	3.87	3.51	2.62	3.55	2.77	2.96
<b>42</b>	3.75	3.5	2.55	3.45	2.75	2.83
<b>43</b>	3.71	3.42	2.52	3.36	2.71	2.7
<b>44</b>	3.65	3.38	2.5	3.34	2.7	2.36
<b>45</b>	3.61	3.16	2.24	3.09	2.57	1.92
<b>46</b>	3.53	2.92	2.22	3.07	2.13	1.85
<b>47</b>	3.36	2.91	2.16	2.85	2.06	1.84
<b>48</b>	3.26	2.69	1.46	2.62	1.85	1.74
<b>49</b>	2.4	2.17	1.04	1.7	1.5	0.77
<b>50</b>	1.77	1.19	0.27	1.2	0.62	0.12

As shown in the table of results, beyond the failure value in both synthetic time series, the values are significantly above the threshold for irrigation needs. This fact, in combination with those mentioned above, make the risk practically zero on a monthly basis. If the simulated discharge values were zero or negative then the interpretation of the calculated risk would be different. Negative discharges have apparently no physical meaning and would be replaced by zero. Their statistical significance, however, is that the curve of the probability distribution function would be shifted closer to the limit point for the cover of irrigation needs. The failures that would be identified in that case would be statistically significant and so the risk.

It is also reported that if the analysis was on a weekly or daily basis, the risk would be greater. This means that there may be a weakness to cover the needs for a short time during the period July-August. Even if this time period is two or three days, it would be a problem for sensitive crops, such as alfalfa that is common in the Konitsa plain.

In summary, based on the above analysis there is no risk of insufficiency of the water amount at the location of "Konitsa Bridge" to meet the irrigation needs of the Konitsa plain, but the hydrosystem proved to be sensitive during the summer months. Moreover, as it was shown in the study of Mimikou et al. (1999), changes in the hydrological regime of north-western Greece due to the climate change include the reduction of the summer runoff values in all studied cases and basins considering various climate change scenarios. Therefore, the Aaos river water system is vulnerable during the dry summer period and changes in climate or water use can cause its imbalance.

#### **4. Conclusions**

The main goal of this work was to create a solid background for the further study of the Aaos-Voidomatis hydrosystem. Initially the available hydrological data from several hydrometric stations were supplemented and extended for the period 1950-2012, using simple linear regression models with significant accuracy. The original historical time-series data set created can be used in future studies as primary data prior to and after the construction of the Aaos Springs dam.

Furthermore, the monthly water flow rates of the river were forecasted through the creation of a 50-year long synthetic time-series with the use of the autoregressive models AR(1) and AR(2) based on its historical discharge values and thus without taking into consideration any climatic changes. These models were applied to the available data with statistically significant results. Subsequently the risk of failure to cover existing water needs was estimated. The analysis of the results showed that current water flow rates can sufficiently cover the existing water needs; mainly for the irrigation of the Konitsa plain, with practically no risk of failure.

However, it was proven that the hydrosystem is sensitive to climate fluctuations and/or water use, especially during the summer months and thus it requires a long-term sustainable management plan.

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