

Manuscript Number: ASR-D-11-00425

Title: Adaptation of a single-channel generalized method for the estimation of water surface temperature by means of Landsat 7 ETM+ thermal infrared data. Case study: Embalse del Río Tercero (Córdoba, Argentina).

Article Type: ES

Keywords: water surface temperature; reservoir; cooling system; nuclear power plant; single-channel generalized method; atmospheric water vapor.

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Abstract: Monitoring of warm distribution in water is fundamental to understand the performance and functioning of reservoirs and lakes. Surface water temperature is a key parameter in the physics of aquatic systems processes since it is closely related to the energy fluxes through the water-atmosphere interface. Remote sensing applied to water quality studies in inland waterbodies is a powerful tool that can provide additional information impossible to achieve by other means. The combination of good real-time coverage, spatial resolution and free availability of data makes Landsat system a good choice. Many papers have developed algorithms to retrieve surface temperature (principally, land surface temperature) from at-sensor and surface emissivity data. The aim of this study is to adapt the single-channel generalized method developed by Jiménez-Muñoz and Sobrino (2003) for the estimation of water surface temperature by Landsat 7 ETM+ thermal bands. We choose Embalse del Río Tercero (Córdoba, Argentina) as case study because is a reservoir affected by the outlet of the cooling system of a nuclear power plant, which thermal plume could influence the biota's distribution and biodiversity. These characteristics and long term studies make it a very interesting place to test the methodology.

In the method presented here (AWVM), we consider a constant water emissivity value (0.9885) and we compare the results with radiative transfer classic method (RTM). Values of estimated and observed water surface temperatures obtained by the two compared methods were correlated applying a simple regression model. Correlation coefficients were significant ( $R^2$ : 0.9426 for AWVM method and  $R^2$ : 0.9584 for RTM method) while their standard errors were acceptable in both cases. Nevertheless, AWVM could estimate rather small differences in temperature between sites consistently with the results obtained in field measurements, besides it has the advantage that it only uses values of atmospheric water vapor and it can be applied to different thermal sensors using the same equation and coefficients.

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6 **Adaptation of a single-channel generalized method for the estimation of water surface**  
7 **temperature by means of Landsat 7 ETM+ thermal infrared data. Case study: Embalse del Río**  
8 **Tercero (Córdoba, Argentina).**  
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24 **Abstract**

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27 Monitoring of warm distribution in water is fundamental to understand the performance and  
28 functioning of reservoirs and lakes. Surface water temperature is a key parameter in the physics of  
29 aquatic systems processes since it is closely related to the energy fluxes through the water-  
30 atmosphere interface. Remote sensing applied to water quality studies in inland waterbodies is a  
31 powerful tool that can provide additional information impossible to achieve by other means. The  
32 combination of good real-time coverage, spatial resolution and free availability of data makes  
33 Landsat system a good choice. Many papers have developed algorithms to retrieve surface  
34 temperature (principally, land surface temperature) from at-sensor and surface emissivity data.  
35 The aim of this study is to adapt the single-channel generalized method developed by Jiménez-  
36 Muñoz and Sobrino (2003) for the estimation of water surface temperature by Landsat 7 ETM+  
37 thermal bands. We choose Embalse del Río Tercero (Córdoba, Argentina) as case study because is  
38 a reservoir affected by the outlet of the cooling system of a nuclear power plant, which thermal  
39 plume could influence the biota's distribution and biodiversity. These characteristics and long term  
40 studies make it a very interesting place to test the methodology.  
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43 In the method presented here (AWVM), we consider a constant water emissivity value (0.9885)  
44 and we compare the results with radiative transfer classic method (RTM). Values of estimated and  
45 observed water surface temperatures obtained by the two compared methods were correlated  
46 applying a simple regression model. Correlation coefficients were significant ( $R^2$ : 0.9426 for AWVM  
47 method and  $R^2$ : 0.9584 for RTM method) while their standard errors were acceptable in both  
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5 consistently with the results obtained in field measurements, besides it has the advantage that it  
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12 Keywords: water surface temperature; reservoir; cooling system; nuclear power plant; single-  
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14 channel generalized method; atmospheric water vapor.

## 15 16 17 1. Introduction

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20 Reservoirs are built for different purposes like drinking water supply, flood control, irrigation, or  
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22 power generation (Casamitjana et al., 2003). Water temperature governs most of the chemical  
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24 and biochemical processes in aquatic ecosystems. Monitoring of warm distribution in water is  
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26 fundamental to understand the performance and functioning of reservoirs and lakes (Kimmel et  
27  
28 al., 1990) and, its survey is important for water quality management, land-use, and hydrological  
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30 studies (Kay et al., 2005). Surface water temperature is a key parameter in the physics of aquatic  
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32 systems processes since it is closely related to the energy fluxes through the water-atmosphere  
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34 interface.

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36 Water temperature surveys using conventional limnological sampling are expensive and time-  
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38 consuming. Remote sensing applied to water quality studies in inland waterbodies is a powerful  
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40 tool that can provide systematic and periodic coverage and also additional information from the  
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42 non-visible regions of the spectrum impossible to achieve by other means (Dash et al., 2002; Novo  
43  
44 et al., 2006; Alcântara et al., 2009). Moreover, satellite information allows obtaining data in digital  
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46 format that can be easily combined with other geographic information and used to generate  
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48 quantitative models (Chuvienco, 2002) and makes possible the reduction of field sampling  
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50 frequency and costs. The success of water quality parameters quantification in inland  
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52 environments depends on water characteristics and the sensor used. The combination of good  
53  
54 real-time coverage, spatial resolution and free availability of data makes Landsat system a good  
55  
56 choice for studying this kind of waterbodies.

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58 During late 20th century, rapid technological advances resulted in increased availability of thermal  
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60 infrared images and methodologies to calibrate and interpret them. Studies were performed using  
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62 these data for mapping water surface temperature, especially in places with marked variations  
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3 -e.g. cooling channels of nuclear power plants, hot springs, etc.- (Gibbons and Wukelic, 1989,  
4 Mustard et al. 1999, Cherkauer et al. 2005, Zoran et al., 2005, Ahn et al. 2006, Alcântara et al.,  
5 2010).

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8 All the energy exchanges between water mass and atmosphere take place within the very thin  
9 surface skin layer that can be remotely sensed. Temperature in this layer is typically cooler than  
10 the bulk water temperature so a critical factor to consider is the relationship between these  
11 temperatures. Evaporative cooling, wind speed and diurnal energy fluxes affect this complex  
12 relationship, nevertheless Yokoyama et al. (1995) and Schneider and Mauser (1996) demonstrated  
13 that remotely measured skin temperatures are representative of bulk water temperatures.

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Emitted thermal infrared radiation (TIR,  $\lambda = 8$  to  $14 \mu\text{m}$ ) can be used to measure temperature of  
the water surface layer of approximately  $100 \mu\text{m}$  (Kishino et al., 2000, Donlon et al., 2002, Kumar  
et al., 2003; Kay et al., 2005, Becker & Daw, 2005). After removal of atmospheric and emissivity  
effects, recovered temperatures should be similar for all TIR sensor bands. Temperature  
differences between bands can also be consequence of sensor calibration problems, instrument  
noise, and in-scene spectral variability from nonwater materials (Kay et al., 2005).

Many papers have developed algorithms to retrieve surface temperature (principally, land surface  
temperature) from at-sensor and surface emissivity data. In the case of Landsat platform, with one  
thermal band, the only method that can be applied is a single channel one. Traditionally, the main  
disadvantages of this kind of method are that some atmospheric parameters are needed for the  
estimation and in addition, they cannot be used for other thermal sensor (Jiménez-Muñoz and  
Sobrino, 2003).

Jiménez-Muñoz and Sobrino (op. cit.) have developed a generalized single-channel method for  
retrieving land surface temperature from remote sensing data, called here as: *atmospheric water  
vapour method (AWVM)*. This algorithm, assuming that emissivity is known, only uses the total  
atmospheric water vapor content, the effective wavelength of the sensor and the at-sensor data  
(brightness temperature or at-sensor radiance). Land surface emissivity must be estimated, and  
different methods can be applied (Becker and Li, 1995; Goïta and Royer, 1997; Sobrino and  
Raissouni, 2000, Gillespie et al., 1998) while, in water surfaces, emissivity can be considered  
constant. In this paper we followed Snyder et al. (1998) who classified fourteen “emissivity  
classes” and presented their emissivity models and values for each emissivity class by combining  
various spectral component measurements with structural factors, based on MODIS thermal  
channels 31 and 32.

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3 The aim of this study is to adapt the single-channel generalized method (Jiménez-Muñoz and  
4 Sobrino, 2003) for the estimation of water surface temperature by Landsat 7 ETM+ thermal bands.  
5 Embalse del Río Tercero is a reservoir affected by the outlet of the cooling system of a nuclear  
6 power plant, which thermal plume could influence the biota's distribution and biodiversity  
7 (Mariuzzi et al., 1981, 1989). These characteristics and long term studies make it a very interesting  
8 place to test the methodology.  
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## 14 2. Materials and methods

### 16 2.1. Study area

17 Embalse del Río Tercero (ERT) (32° 11'S, 64° 25'W) is a medium size reservoir (5400 Hm<sup>3</sup>, 46 Km<sup>2</sup>,  
18 45 m max. depth, 12 m mean depth, 0.44 y mean residence time), located in the center of  
19 Argentina, in the province of Córdoba (Fig. 1). The reservoir was built in 1936 for flood control,  
20 irrigation and hydroelectric purposes. Since 1983 water level fluctuations are regulated by  
21 requirement of a nuclear power plant of 600 MW, which uses its water for cooling purposes. The  
22 water inflow to the nuclear plant is located at 15 m deep in the Garganta area (G). Water  
23 temperature raises up 7° C in the cooling system and the outflow is directed through an open  
24 cooling channel of 6 km to the southern sector of the reservoir and discharges in a tributary, the  
25 Quillínzo River (Mariuzzi et al., 1992).  
26

27 As part of a monitoring program, since 1996 several physical, chemical and biological properties of  
28 the ERT waters are surveyed bimonthly (Mariñelarena et al., annual reports). Surface water  
29 temperature is measured at stations PAA, PC, M1, M2, C, G and D. Results have shown that  
30 warmer water released at Quillínzo River forms a thermal plume most of the time restricted to the  
31 river channel.  
32

33 According to these reports, temperature in station G is always lower than that of stations PM1 and  
34 PM2. The differences between stations PM1 and G are 3 to 8 °C (average 4.3 °C, n = 74), while  
35 between stations PM2 and G differences are 0.5 to 5 °C (average 2.4 °C, n=74) Mariñelarena (pers.  
36 comm.). Nevertheless, in a few dates in 15 years, the temperature of PM2 was lower than that of  
37 station G, due to programmed nuclear plant shutdowns and the effect of low temperature of  
38 Quillínzo River waters.  
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### 56 2.2. Remote sensing and water surface temperature estimation

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3 In this work we used images from Landsat 7 ETM<sup>+</sup> (Path/Row: 229/82). This sensor has a unique  
4 thermal band (band 6) divided in two: low gain band (B6L) and high gain band (B6H), with a  
5 spectral resolution from 10.4 to 12.5  $\mu\text{m}$  and a spatial resolution of 60 m covering a surface of  
6 3.600 m<sup>2</sup> per pixel (0.36 ha). Images were obtained from the website of the United States  
7 Geological Survey (USGS) (<http://glovis.usgs.gov/>). They were in the UTM projection system, with  
8 WGS84 as ellipsoid and datum. They were resampled to 30 m of spatial resolution.  
9

10 The criteria for the selection of images were:

- 11 (1) Existing *in situ* data in  $\pm 3$  days to the satellite passes.
- 12 (2) Images from different seasons in order to find temperature differences.
- 13 (3) 0% cloud cover when possible.
- 14 (4) No heavy rainfall prior to the image to minimize the effects of changes in water surfaces  
15 that disturb the estimates.

16 Selected images were: March 19<sup>th</sup>, 2000, August 13<sup>rd</sup>, 2001, August 16<sup>th</sup>, 2002 and July 10<sup>th</sup>, 2006  
17 and November 15<sup>th</sup>, 2006.

18 Software ENVI 4.3 (Research Systems, Inc) was used for digital processing of images and  
19 application of algorithm.

#### 20 2.2.1. Generation of “water images”

21 Definition of the specific shoreline for each image is critical to eliminate areas that are not  
22 representative of open waters and give inaccurate values of temperatures near the coast (Wang  
23 et al., 2008). In order to select pure water pixels, an unsupervised classification of near-infrared  
24 band (Band 4) was made. These classes were used as a mask for obtaining the desired image, thus  
25 eliminating the areas of land.

#### 26 2.2.2. Radiometric conversion

27 Conversion of digital numbers (DN) to physical variables (radiance, reflectance and temperature)  
28 is very useful in the interpretation of images and to compare data from the same sensor over time  
29 or between sensors (Chuvieco, 2002). Digital processing for this conversion involves two main  
30 steps: radiometric calibration (Wukelic *et al.*, 1989) and atmospheric correction (Cooper and Asrar  
31 1989).  
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3 DN values were converted to at-sensor radiance by applying the gain and bias of the detectors  
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5 (Irish, 2006):  
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$$L_{\lambda} = \text{gain} * \text{DN} + \text{bias} \quad (1)$$

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12 Where:

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14  $L_{\lambda}$  = uncorrected spectral radiance at  $\lambda$  wavelength in  $\text{W} * \text{m}^2 * \text{sr} * \mu\text{m}$ .

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16 DN = digital number

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18 gain and bias = calibration parameters of detectors (See Table 1).  
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### 22 2.2.3. Calculation of at-sensor radiative temperature ( $T_{\text{sat}}$ )

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24 Radiance values calculated in Eq. (1) were converted to at-sensor radiative temperature using the  
25 following equation based on Planck's law (Wukelic et al., 1989):  
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$$T_{\text{sat}} = \frac{K_2}{\ln\left(\frac{K_1}{L_{\lambda}} + 1\right)} \quad (2)$$

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36 Where:

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38  $T_{\text{sat}}$  = at-sensor radiative temperature (K)

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40  $K_2$  and  $K_1$  = ETM+ thermal constants.  $K_2 = 666.09 \text{ W} * \text{m}^2 * \text{sr} * \mu\text{m}$  and  $K_1 = 1282.71 \text{ }^\circ\text{K}$

41  
42 (Chander et al., 2009).  
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44  $L_{\lambda}$  = uncorrected spectral radiance at  $\lambda$  wavelength in  $\text{W} * \text{m}^2 * \text{sr} * \mu\text{m}$ .  
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### 47 2.3. Methods for water surface temperature estimation

#### 48 2.3.1. Atmospheric water vapor method (AWVM) (adapted from Jiménez-Muñoz and Sobrino, 49 2003)

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52 Water emissivity value was considered to be 0.9885 according to Sydner et al. (1998). Surface  
53 temperature calculation ( $T_s$ ), the equation applied is as follows:  
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$$T_s = \gamma [\varepsilon^{-1} (\psi_1 L_{\lambda} + \psi_2) + \psi_3] + \delta \quad (3)$$

Where:

$\gamma$  = see below, Eq. (7).

$\epsilon$  = water emissivity

$L_\lambda$  = uncorrected spectral radiance calculated in Eq. (1).

$\psi_1, \psi_2$  y  $\psi_3$  = atmospheric functions were calculated by a simulation procedure with MODTRAN 3.5 (Jiménez-Muñoz and Sobrino, 2003).

From these values it is easy to obtain the atmospheric functions for every wavelength and for every atmospheric water vapor content (Jiménez-Muñoz and Sobrino, 2003). The values of these functions were calculated by:

$$\psi_1 = 0.14714 w^2 - 0.15583 w + 1.12340 \quad (4)$$

$$\psi_2 = -1.1836 w^2 - 0.37607 w - 0.52894 \quad (5)$$

$$\psi_3 = -0.04554 w^2 + 1.8719 w - 0.39071 \quad (6)$$

Where:

$w$  = atmospheric water content obtained from <http://weather.uwyo.edu/upperair/sounding.html>, including date, hour and geographic location.

$\delta$  = see below, Eq. (8)

$\gamma$  was calculated from:

$$\gamma = \left\{ \frac{C_2 L_\lambda}{T_{sat}^2} \left[ \frac{\lambda_{ef}^4}{C_1} L_\lambda + \lambda_{ef}^{-1} \right] \right\}^{-1} \quad L_{\lambda = gain * DN + bias} \quad \gamma \quad (7)$$

Where:

$C_2 = 14387.7 \mu\text{m} \cdot \text{K}$

$C_1 = 1.19104 \cdot 10^8 \text{ W } \mu\text{m}^4$

$\lambda_{ef}$  = effective wavelength (11.45  $\mu\text{m}$ )

$L_\lambda$  = uncorrected spectral radiance calculated in Eq. (1).

$T_{sat}$  = at-sensor radiative temperature (K) calculated in Eq. (2).

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3  $\delta$  was calculated from:  
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$$5 \quad \delta = -\gamma L_{\lambda} + T_{sat} \quad (8)$$

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9 Where:

10  $L_{\lambda}$ : uncorrected spectral radiance calculated in Eq. (1).

11  $T_{sat}$ : at-sensor radiative temperature (K) calculated in Eq. (2).

### 12 13 14 15 16 2.3.2. Radiative Transfer Method (RTM)

17 The atmospheric correction of the radiance of thermal band involves the removal of atmospheric  
18 effects that contribute to the signal received by the sensors (up to 90% for water). These effects  
19 change surface water temperature values of  $\pm 2$  °C (Kay et al., 2005) considering a constant water  
20 emissivity. Radiance values must be corrected according to:  
21  
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$$24 \quad L_{\lambda}(T_s) = \frac{L_{\lambda} - L_{\lambda up}}{t \cdot \varepsilon_{\lambda}} - \frac{1 - \varepsilon_{\lambda}}{\varepsilon_{\lambda}} \cdot L_{\lambda down} \quad (9)$$

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31 Where:

32  $L_{\lambda}$  = uncorrected spectral radiance calculated in Eq. (1).

33  $L_{\lambda up}$  = upwelling radiance

34  $t$  = atmospheric transmissivity

35  $\varepsilon_{\lambda}$  = water emissivity

36  $L_{\lambda down}$  = downwelling radiance

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46 Atmospheric parameters ( $L_{\lambda up}$ ,  $t$  and  $L_{\lambda down}$ ) were obtained from Atmospheric Correction  
47 Parameter Calculator (<http://www.atmcorr.gsfc.nasa.gov/>) which uses MODTRAN simulator (Barsi  
48 et al., 2003).  
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54 Calculation of surface temperature ( $T_s$ ), according to Artis and Carnahan, (1982):  
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$$56 \quad T_s = \frac{T_{sat}}{1 + \left( \lambda \cdot \frac{T_{sat}}{\rho} \right) n_{\varepsilon}} \quad (10)$$

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3 Where:

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6  $T_{sat}$  = at-sensor radiative temperature (Eq. 2)

7  $\lambda$  = effective wavelength (11.45  $\mu\text{m}$ )

8  $\epsilon$  = water emissivity

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11  $\rho$  = constant 14380  $\text{m}^*\text{K}$

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$$\rho = h * \frac{c}{\sigma} \tag{11}$$

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17 Where:

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20  $\sigma$  = Boltzmann's constant ( $1.38 * 10^{-23} \text{ J/K}$ )

21  $h$  = Planck's constant ( $6.26 * 10^{-34} \text{ J*s}$ )

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24  $c$  = velocity of light ( $2.998 * 10^8 \text{ m/s}$ )

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#### 2.4. Generation of water surface temperature images

29 For estimation of water surface temperature Landsat 7 ETM+ thermal band 6 Low Gain was used.  
30 We choose this band because its range is greater and is not saturated (Suga et al., 2003, Weng et  
31 al., 2004, Chander et al., 2009). Temperature values estimated for each image were extracted by  
32 interpolating the location of the sampling sites using gvSIG free software.  
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#### 2.5. Statistical data analysis

40 From 35 potential points (7 sampling stations and 5 images), only 27 could be included in the  
41 analysis for different reasons (lacking of field data, cloud pixel in the images or land-water mixed  
42 pixels). At first instance, all points were analyzed together (observed and estimated water surface  
43 temperature). Simple regression analysis was made to evaluate the correlation of estimated water  
44 surface temperature obtained by AWVM with *in situ* data. In order to enhance the results  
45 obtained by the AWVM, the same analysis was performed using the classic method RTM. In  
46 addition, for the determination of significant differences between estimated and observed  
47 temperature values, Chi Square statistic test was applied.  
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### 3. Results and discussion

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3 Values of estimated and observed water surface temperatures obtained by the two compared  
4 methods were correlated applying a simple regression model. Correlation coefficients were  
5 significant ( $R^2$ : 0.9426 for AWVM method and  $R^2$ : 0.9584 for RTM method) while their standard  
6 errors were acceptable in both cases. Both models showed very good adjustment, although the  
7 RTM fits slightly better (Table 2, Figs. 2A and 2B). Chi Square analysis performed to test these  
8 results showed that estimations were accurate (Table 3).  
9

10 Analysis of data in a time sequence allowed a more detailed e of the accuracy of AWVM. *In situ*  
11 and RTM data were included to compare results (Fig. 3). The temperature values estimated by  
12 AWVM were higher than those for the RTM (Figs. 3A, 3B, 3C and 3E), although following the same  
13 trend that matches the behavior of *in situ* values. Both methods have overestimated observed  
14 values, except for data of July 2006, when field measurements were underestimated. This  
15 situation occurs because in winter there is a greater temperature difference between the colder  
16 air layer just above the water and the most superficial water layer. As a consequence both  
17 methods lost accuracy; however the trend curves continue being coherent. In August 2002,  
18 AWVM overestimated *in situ* data in two of the three sampling stations. At sites D, G, C and PM2,  
19 the differences were minor while in the remaining sites (PAA, PC and PM1) differences were  
20 higher (up to 3 ° C).  
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23 Differences in surface water temperature estimated in stations close to (PM1 and PM2) and far  
24 from (G) the cooling channel are in agreement with the conclusions of Mariñelarena et al. (annual  
25 reports). In all cases the estimated temperatures at station PM1 were higher than those at PM2  
26 and these were higher than those at station G, following a dissipation line from the cooling  
27 channel outlet to the lake (Fig. 3).  
28

29 We built the water surface temperature maps based on AWVM (Fig. 4) considering the obtained  
30 results and the fact that AVWM has important advantages over RTM (simple atmospheric data for  
31 correction and possibility of application to different thermal sensors).  
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#### 33 4. Conclusions

34 AWVM and RTM methods adequately estimated water surface temperature in Embalse del Río  
35 Tercero. Nevertheless, AWVM could estimate rather small differences in temperature between  
36 sites consistently with the results obtained in field measurements. So we conclude that AWVM is  
37 the most suitable method for monitoring the thermal plume in this reservoir using satellite  
38 imagery. Besides, as mentioned above, this method has the advantage that it only uses values of  
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3 atmospheric water vapor and it can be applied to different thermal sensors using the same  
4 equation and coefficients.

5  
6 This adaptation of the generalized single-channel method developed by Jiménez-Muñoz and  
7 Sobrino (2003) provides a simpler tool for the calculation of water surface temperature in this  
8 waterbody and expands the possibilities of application in other places with Landsat images or  
9 other existing sensors.

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Figure Captions:

Figure 1. Location of the sampling stations in the Río Tercero Reservoir. References: D: Dique, G: Garganta, C: Confluencia, PM1: Pluma M1, PM2: Pluma M2, PC: Pluma Canal, PAA: Pluma Aguas Arriba.

Figure 2. Plot of estimated water surface temperature versus measured water surface temperature. A. Atmospheric Water Vapour Method (AWVM). B. Radiative Transfer Method (RTM).

Figure 3. Comparison of measured and estimated water surface temperature with RTM and AWVM over sampling stations: A. March 19th, 2000, B. August 13rd, 2001, C. August 16th, 2002, D. July 10th, 2006 and E. November 15th, 2006.

Figure 4. Water surface temperature maps (°C). A. March 19th, 2000, B. August 13rd, 2001, C. August 16th, 2002, D. July 10th, 2006 and E. November 15th, 2006.

Table Captions:

Table 1. Gains and Biases of Landsat 7 ETM+ thermal bands.

Table 2. Regression statistics of comparison between measured and estimated water surface temperature from both methods.

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Table 3. Chi Square test for comparison between measured and estimated water surface temperature from both methods. Critical value:  $p < 0.05$  38.885.

Table

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Satellite/Sensor	Band	Gain (W/m <sup>2</sup> *sr*μm)/DN	Bias W/m <sup>2</sup> *sr*μm	Source
Landsat 7 ETM+	6 Low gain <sup>a</sup>	0.067087	-0.07	Chander <i>et al.</i> , 2009
	6 High gain	0.037205	3.16	

Table

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Methods	Multiple R <sup>2</sup>	Adjusted R <sup>2</sup>	Standard error of estimate	F	P(x)	n
AWVM	0.949855	0.940262	1.224921013	410.2304	0	27
RTM	0.958384	0.956719	1.042623618	575.7314	0	27

Table

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Methods	Chi Square	Degree of freedom	P(x)
AWVM	2.869466	26	<1
RTM	1.883663	26	<1

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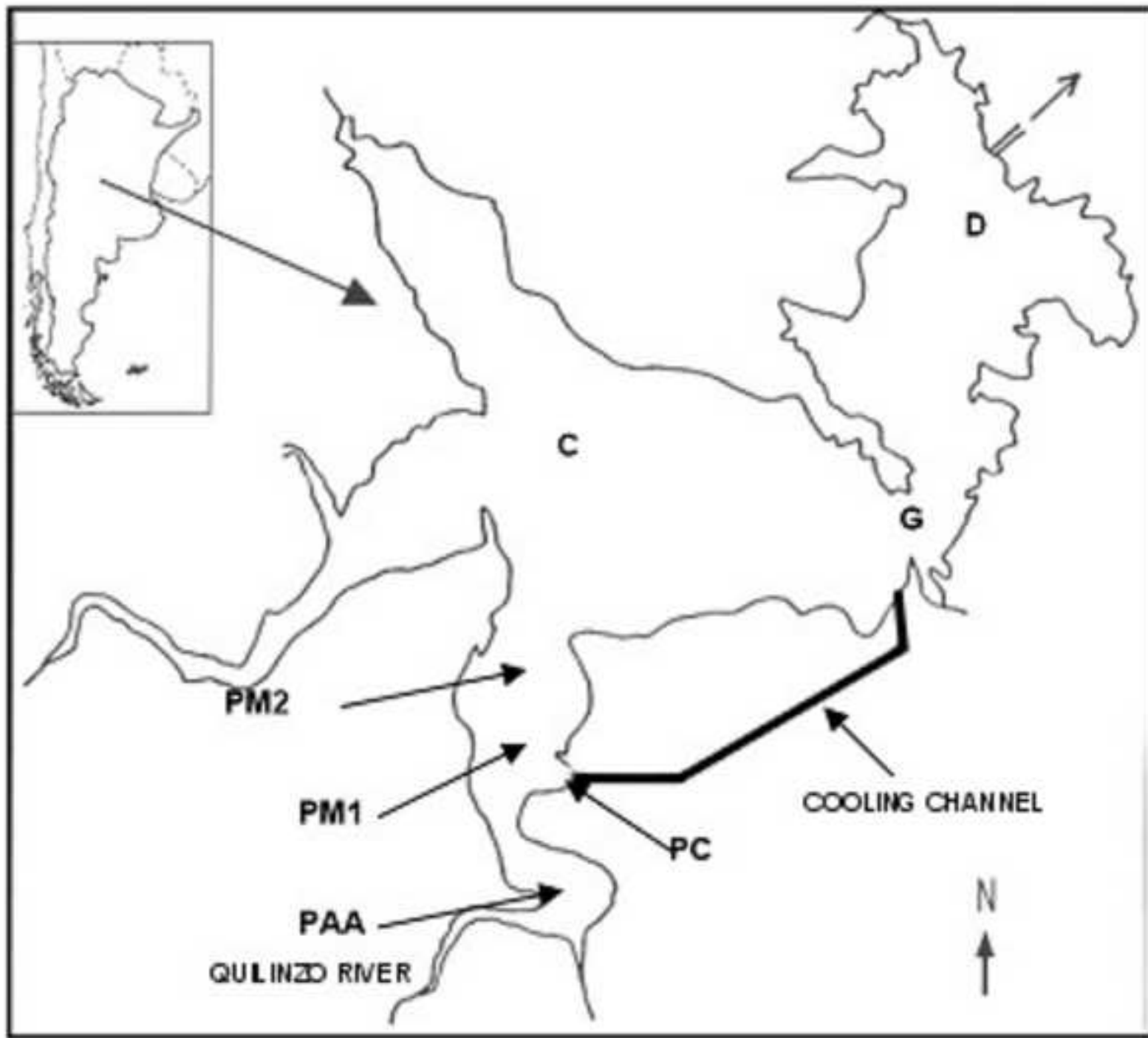


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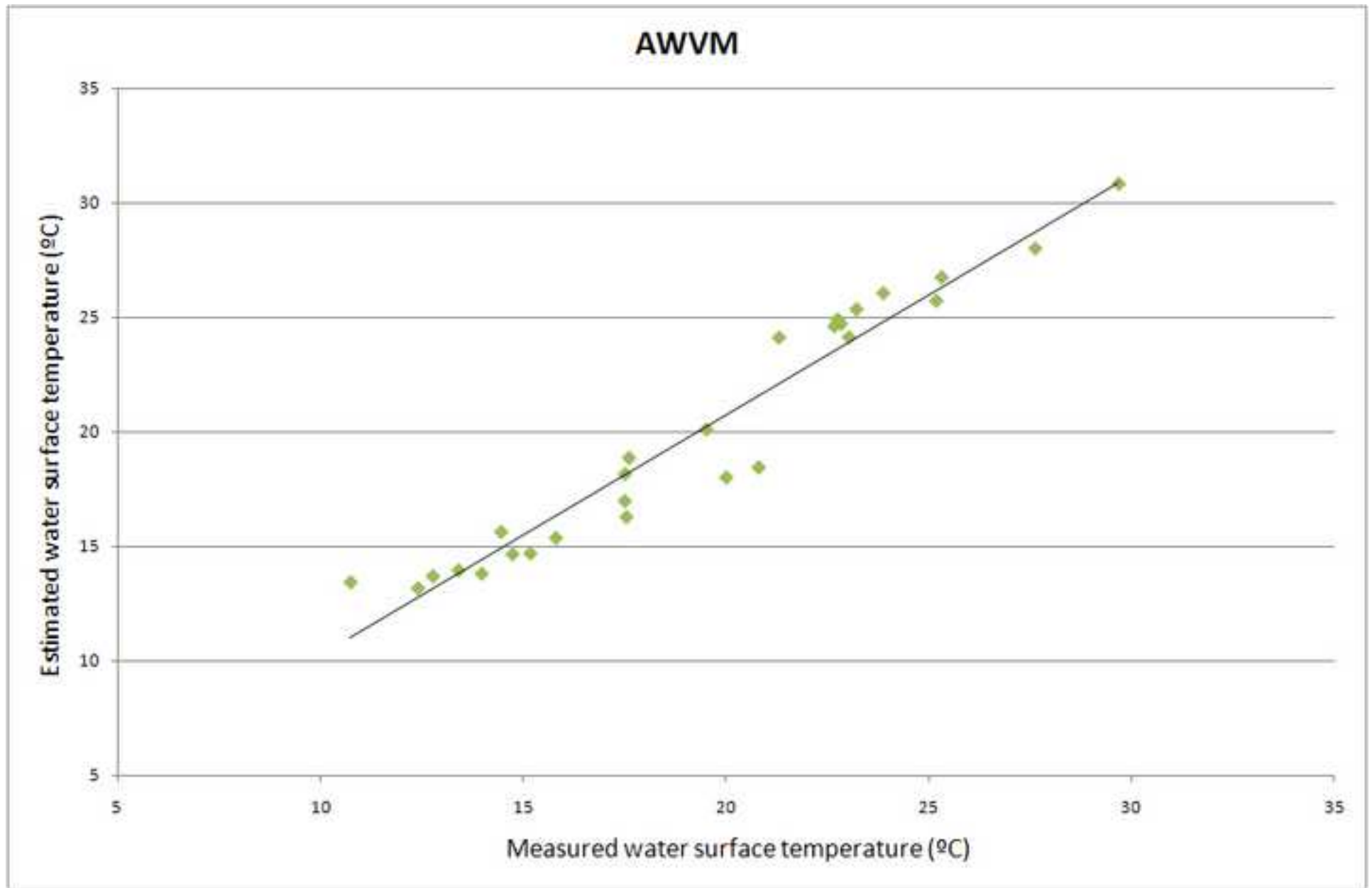


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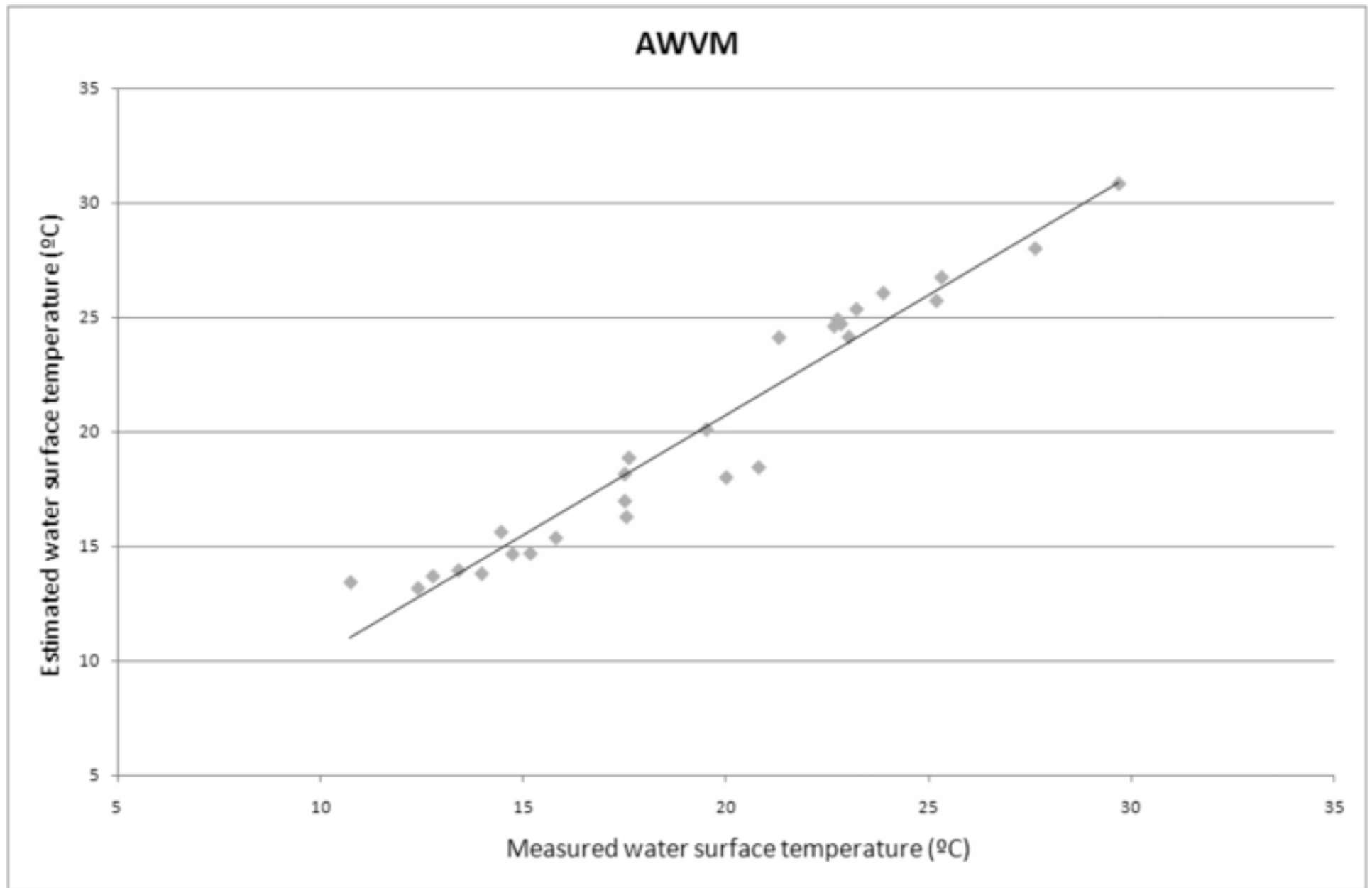


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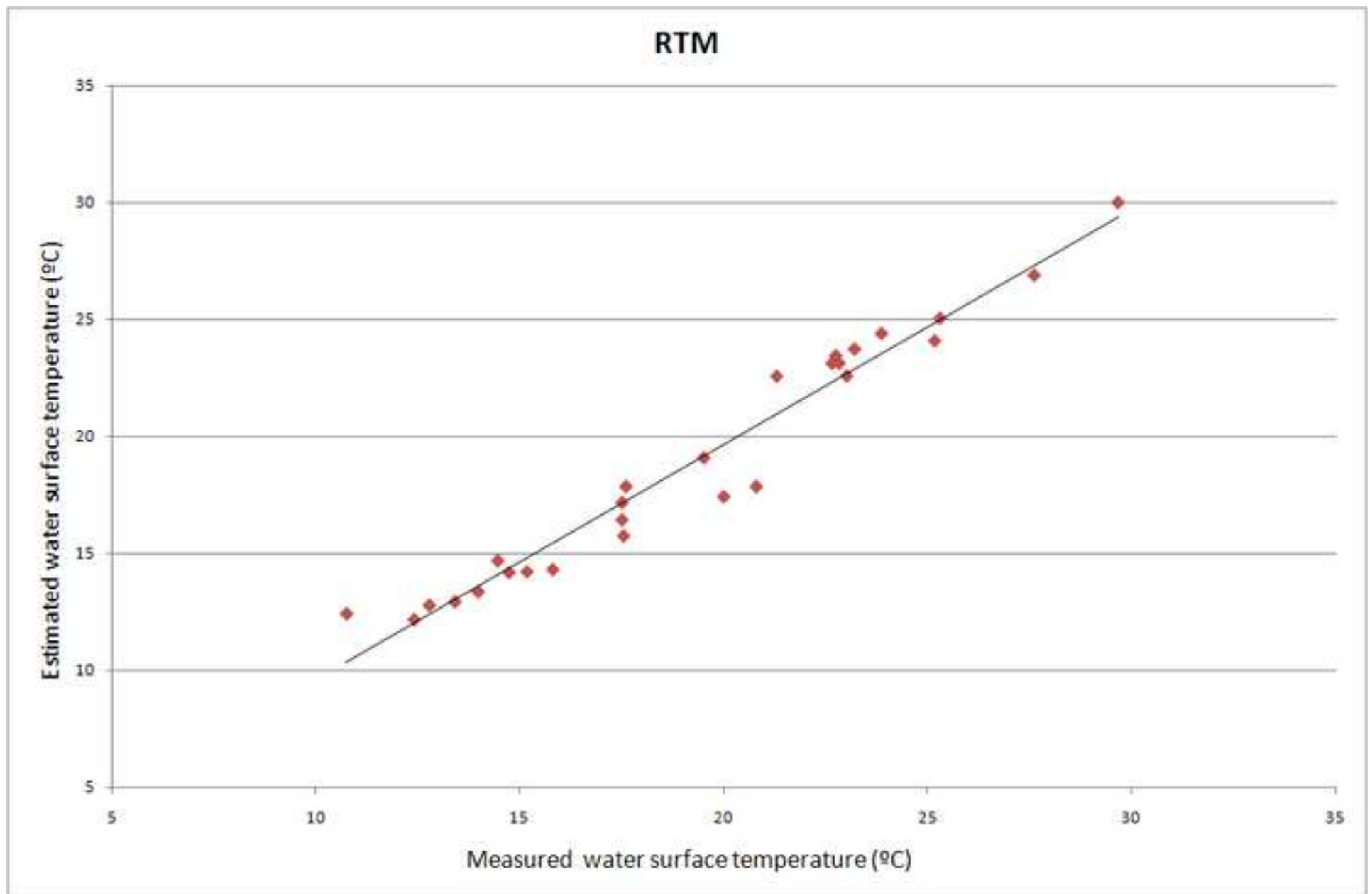


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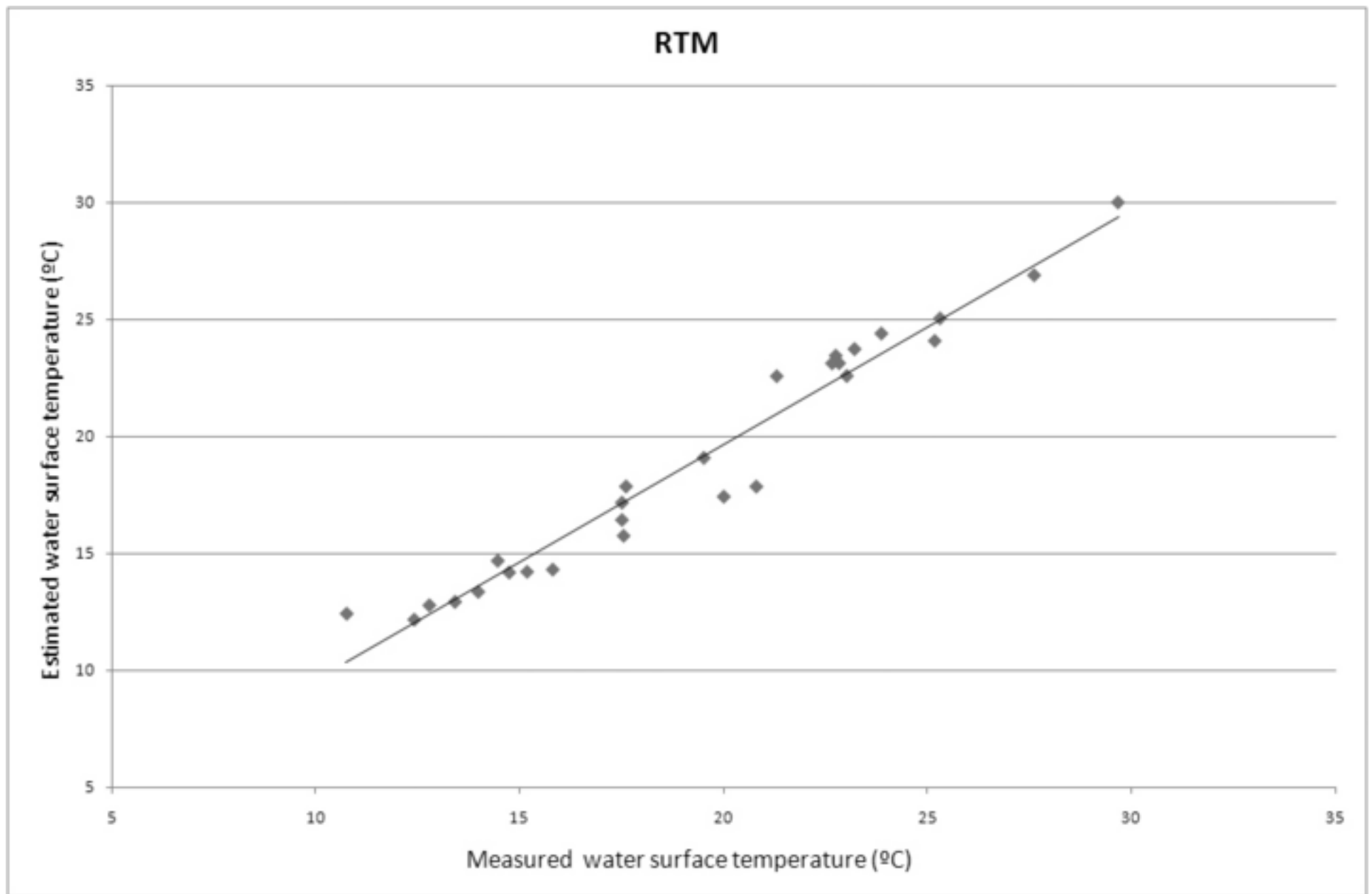


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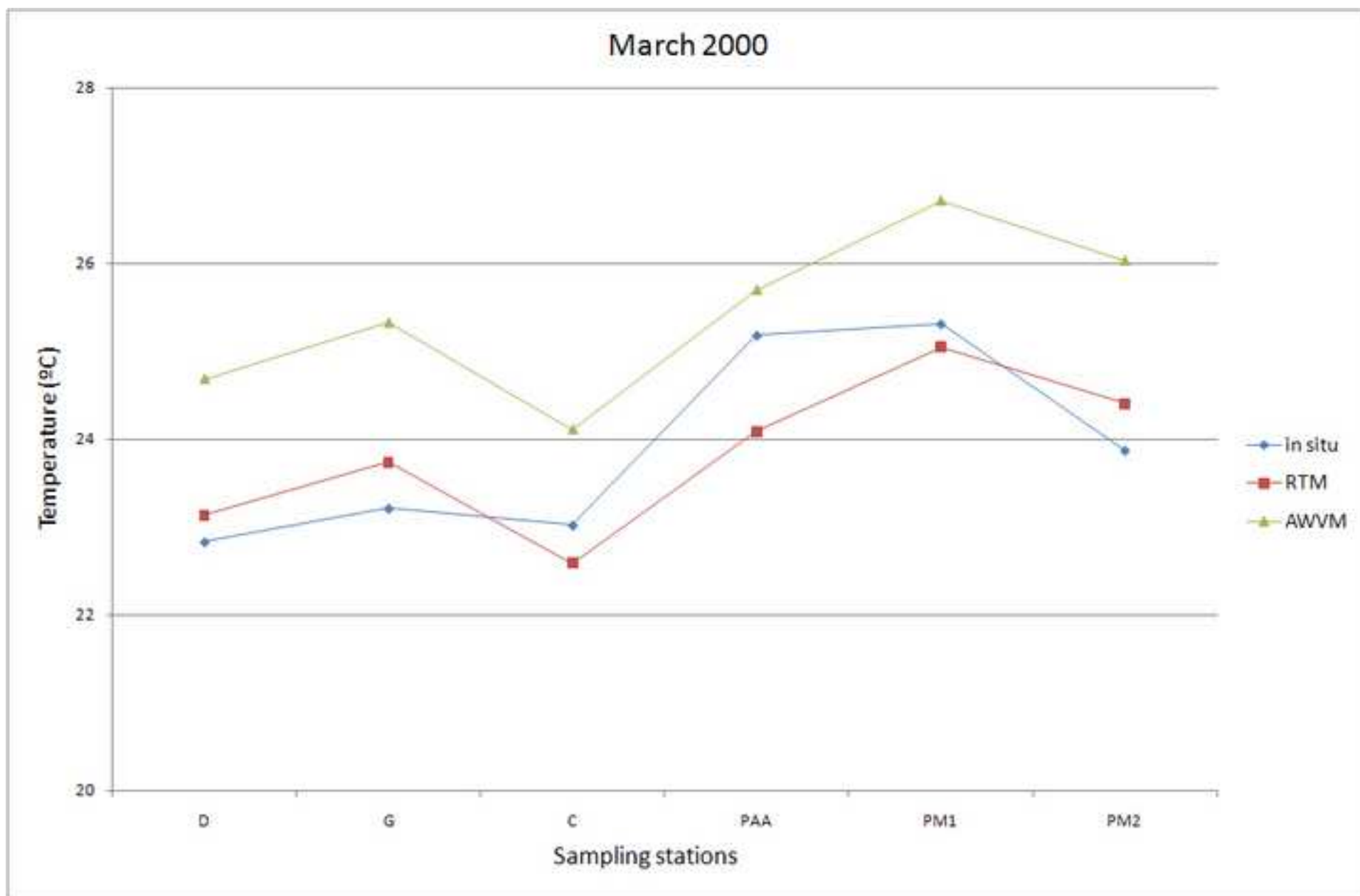


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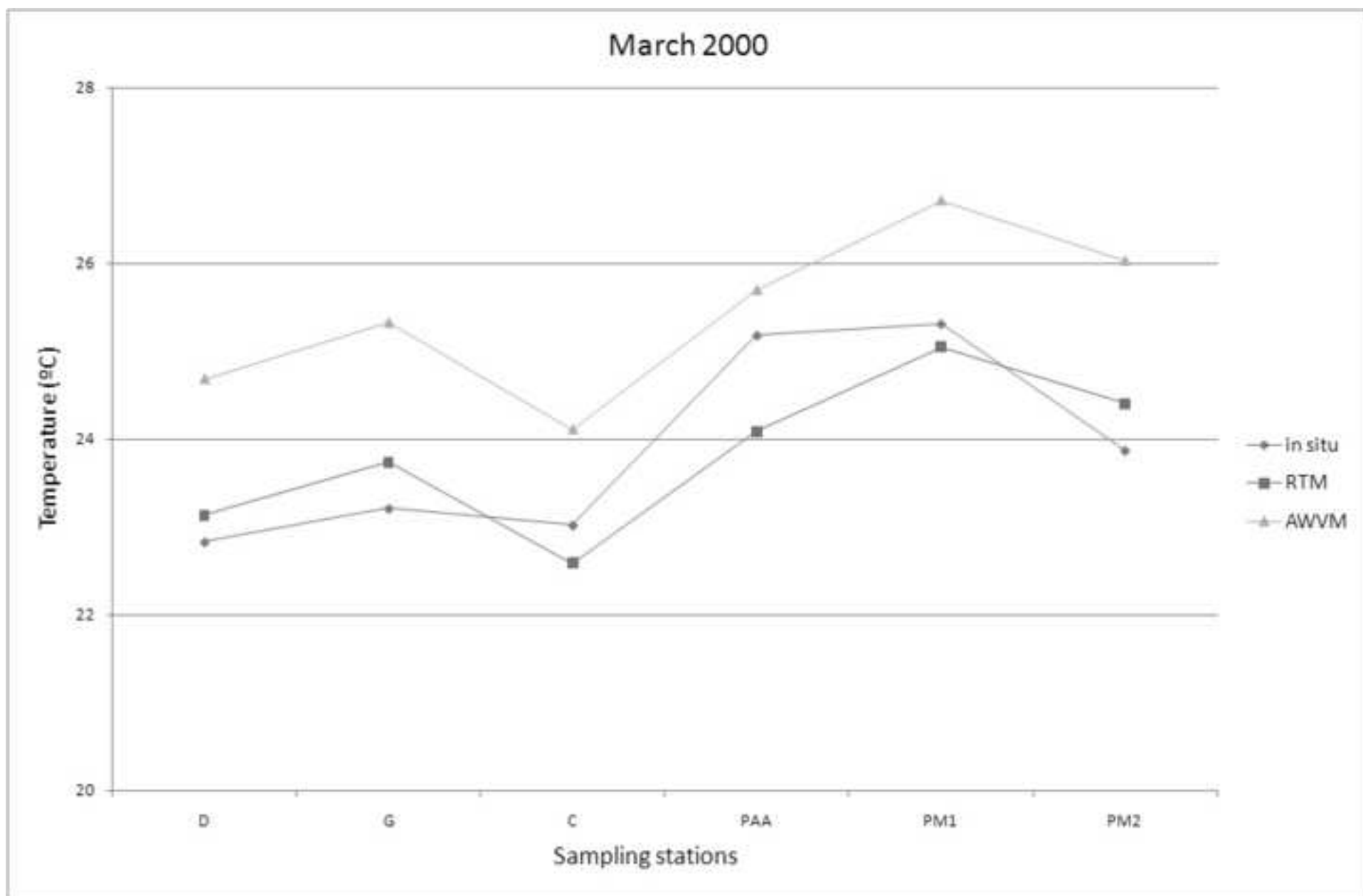


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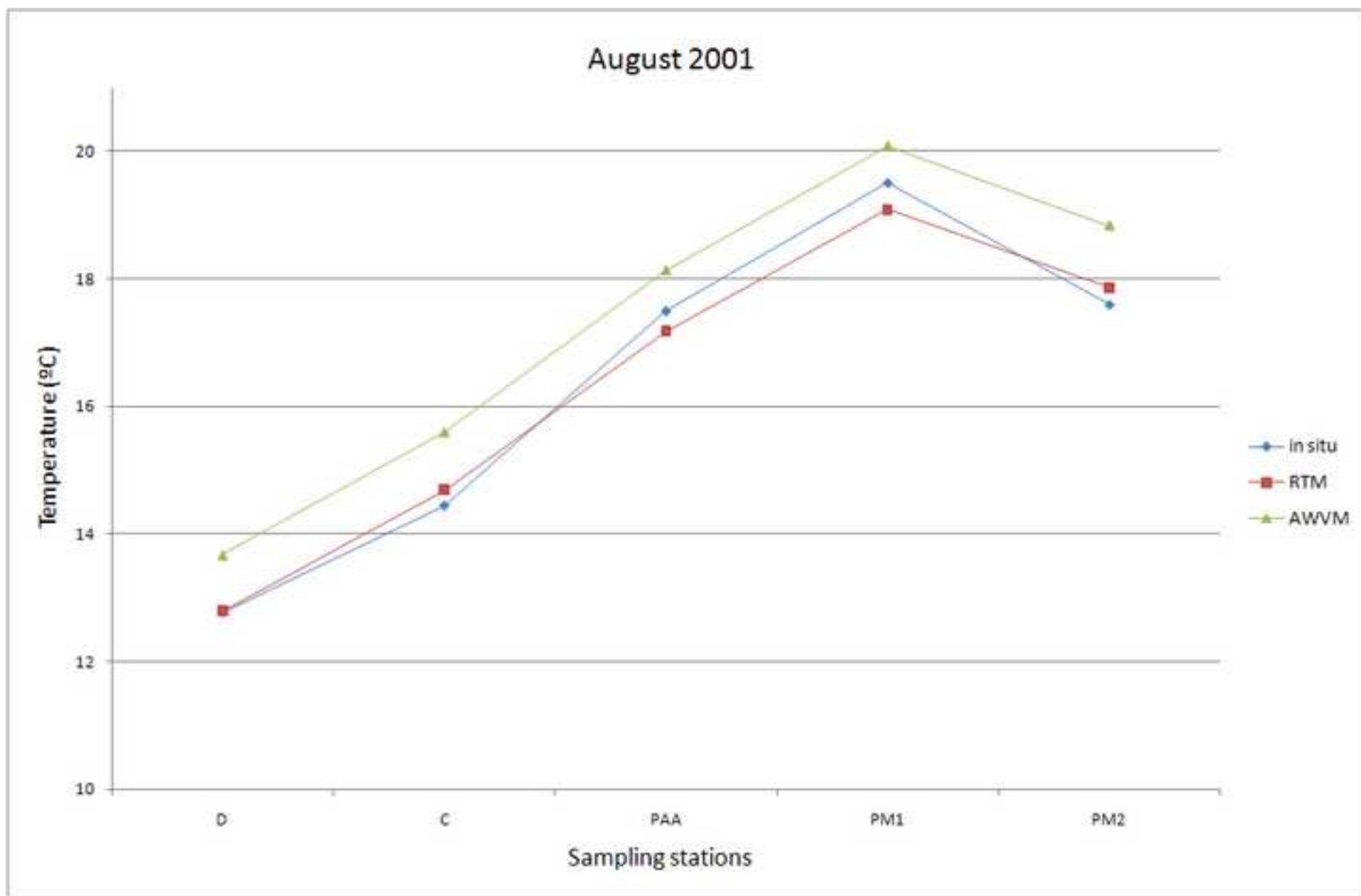


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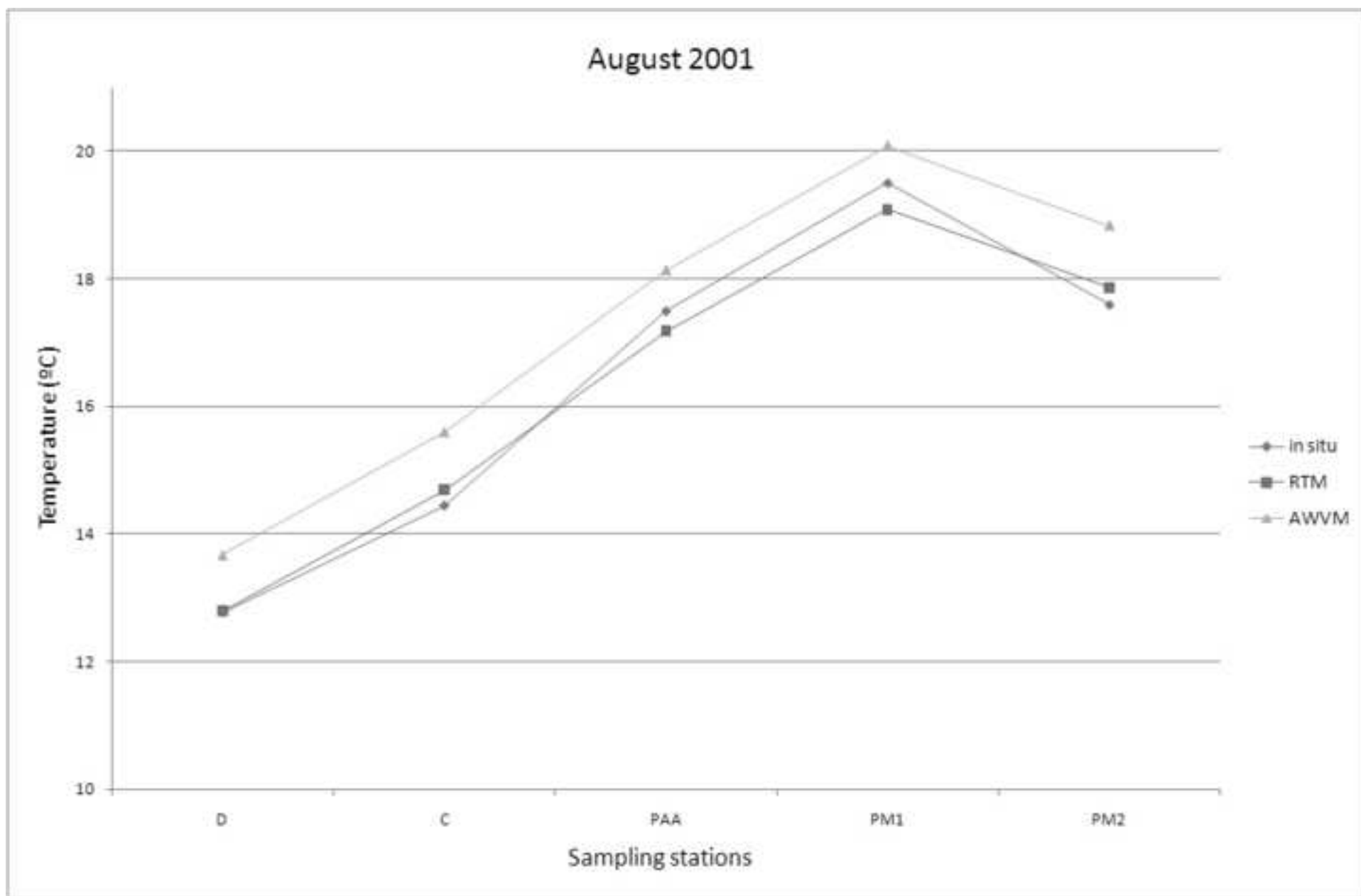


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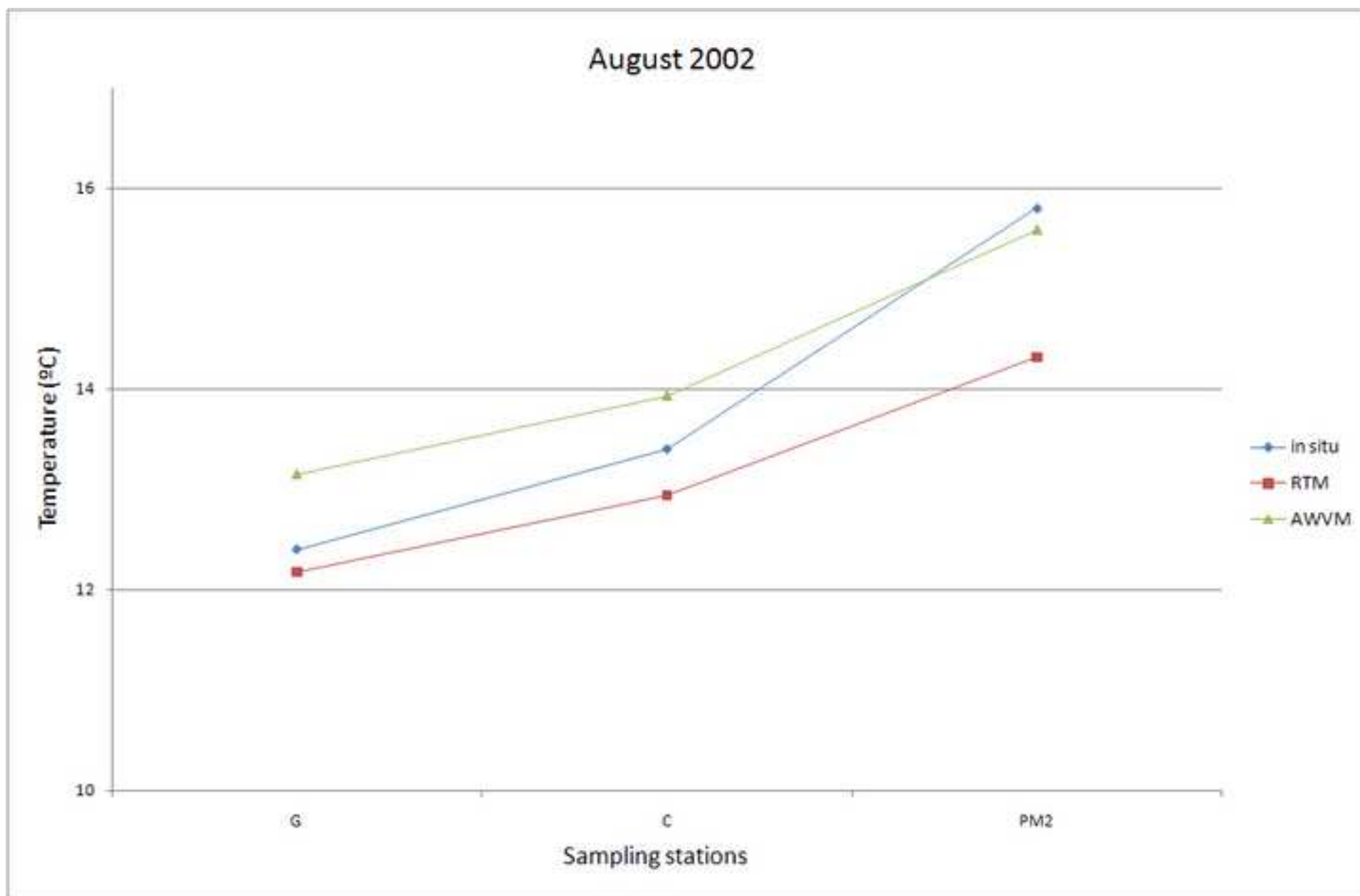


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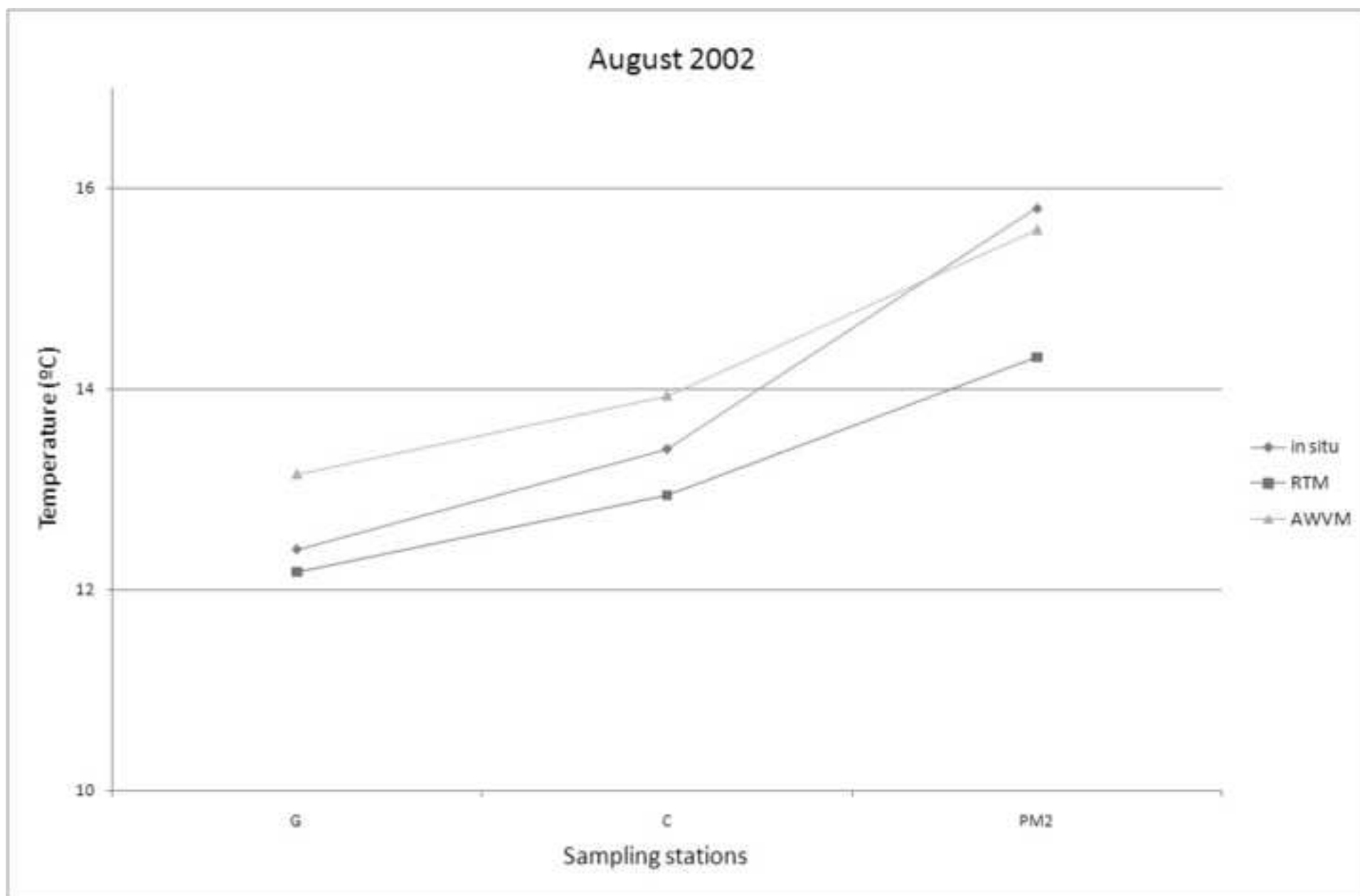


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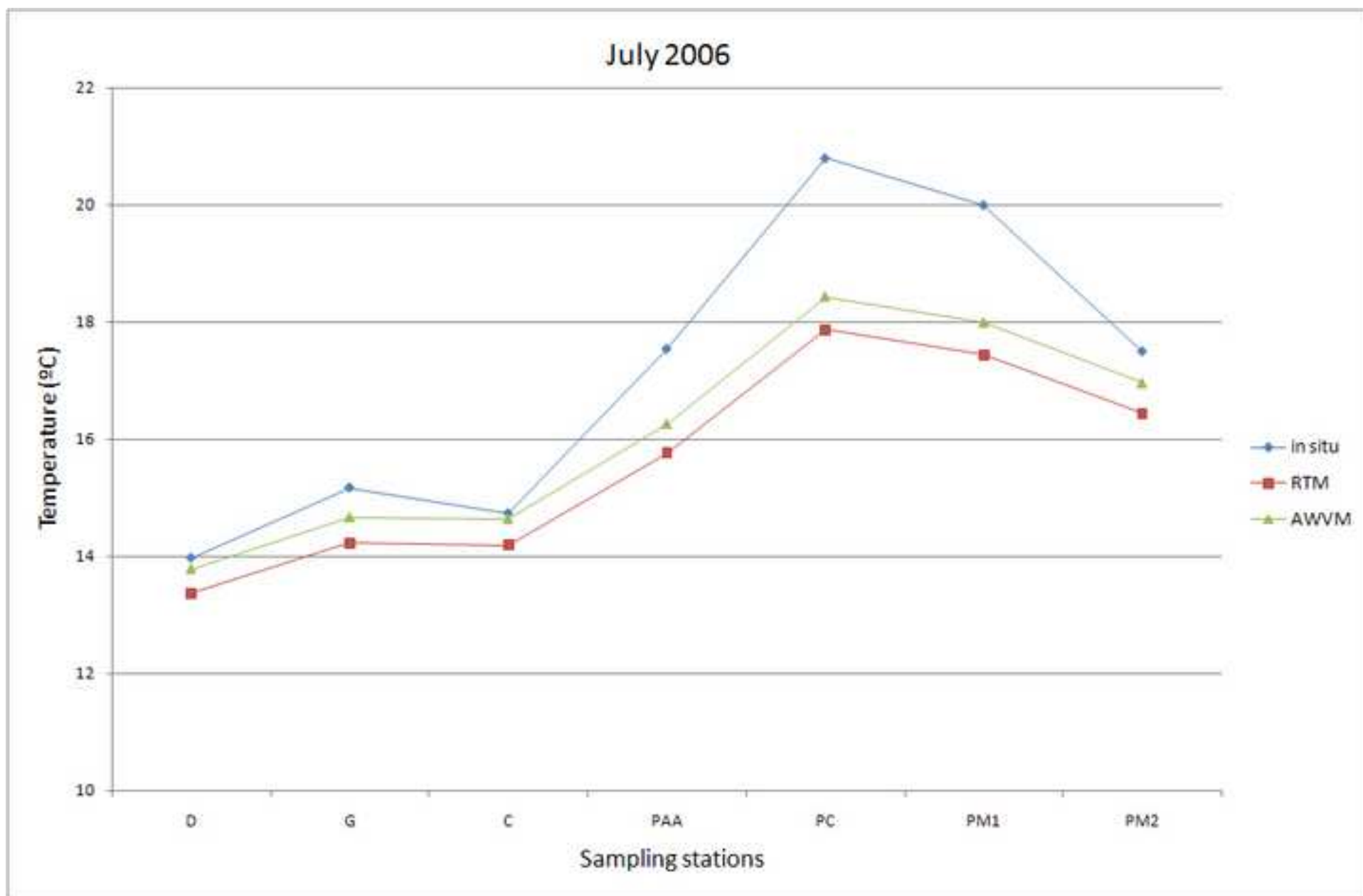


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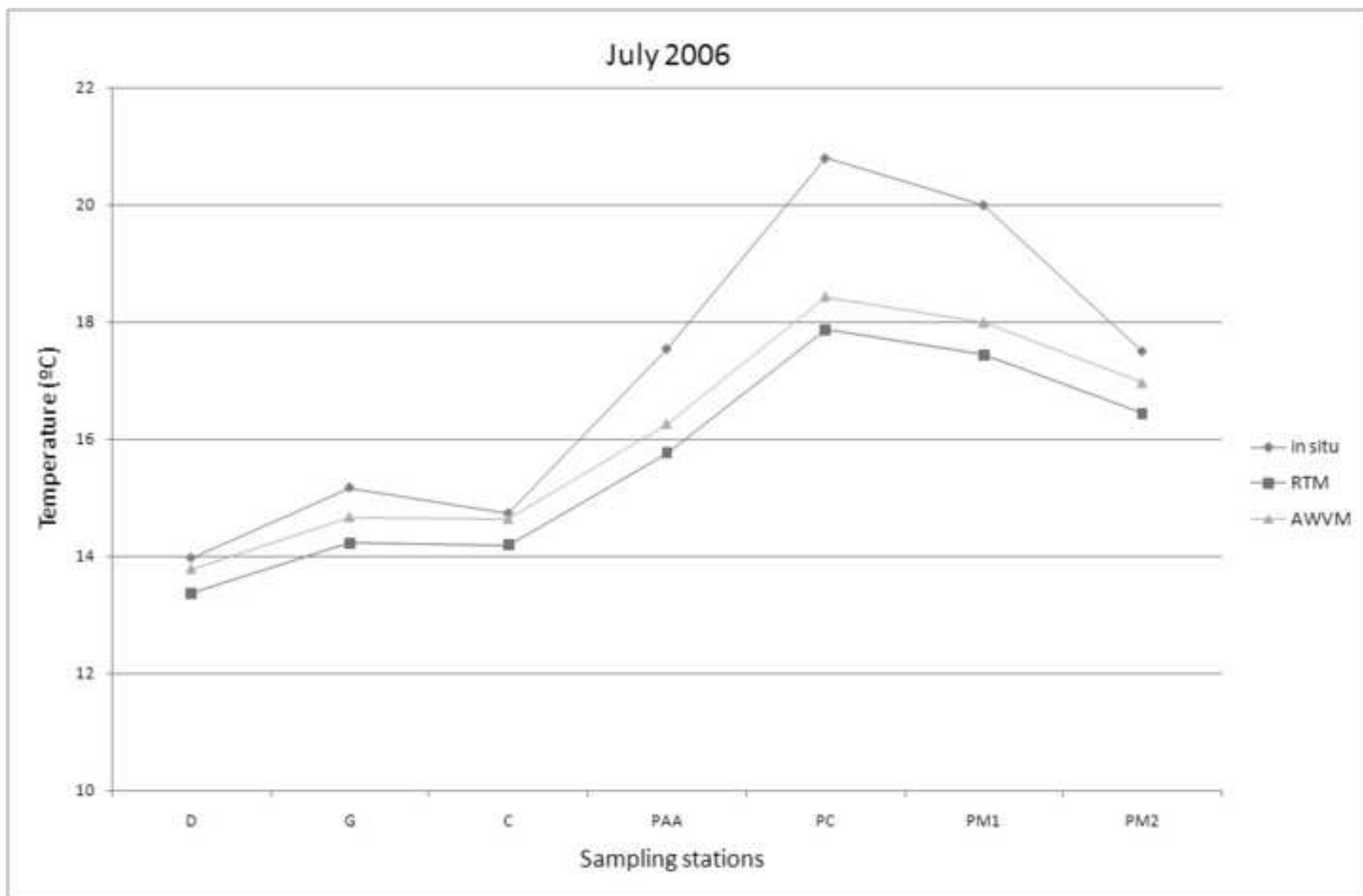


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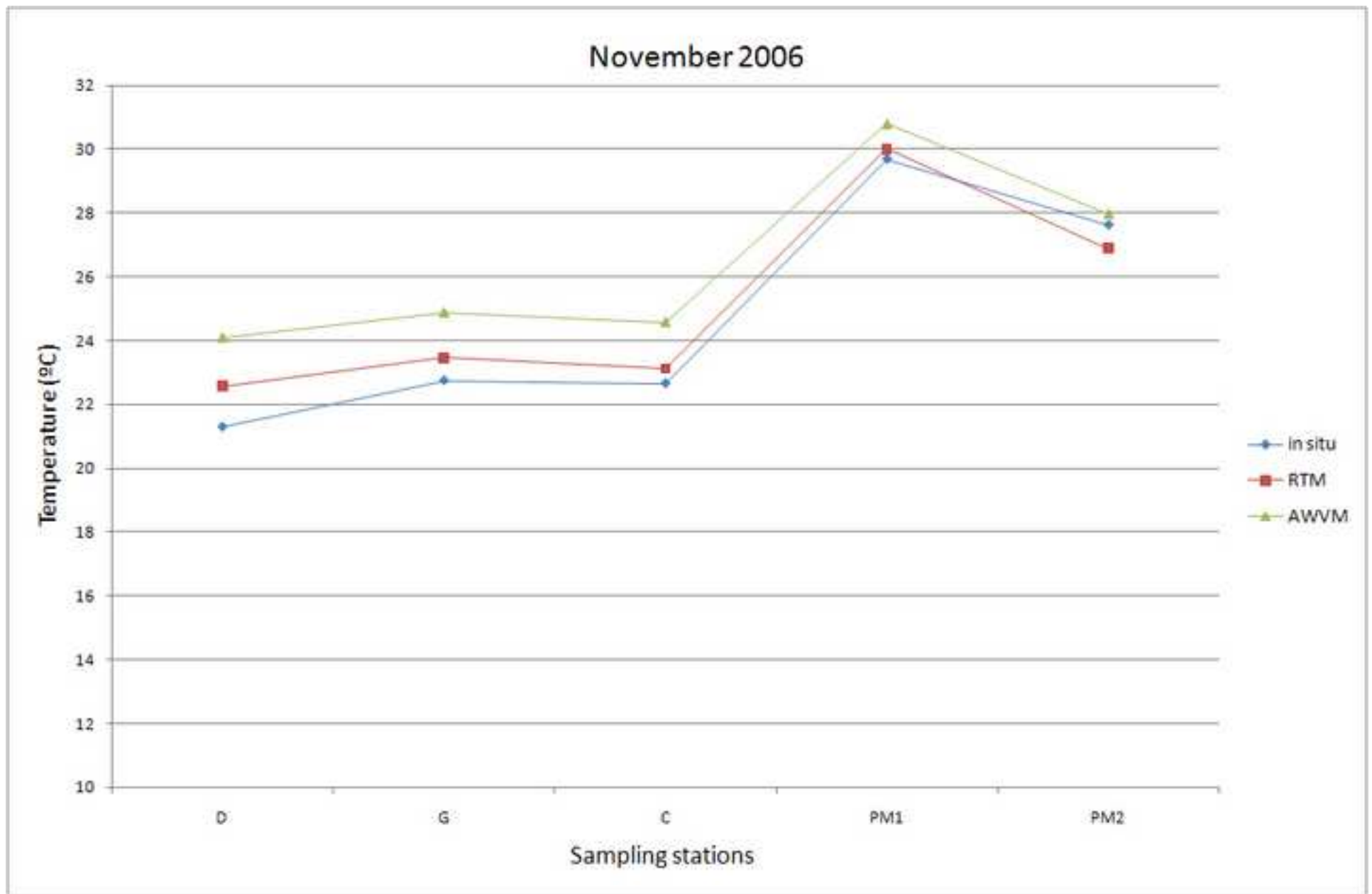


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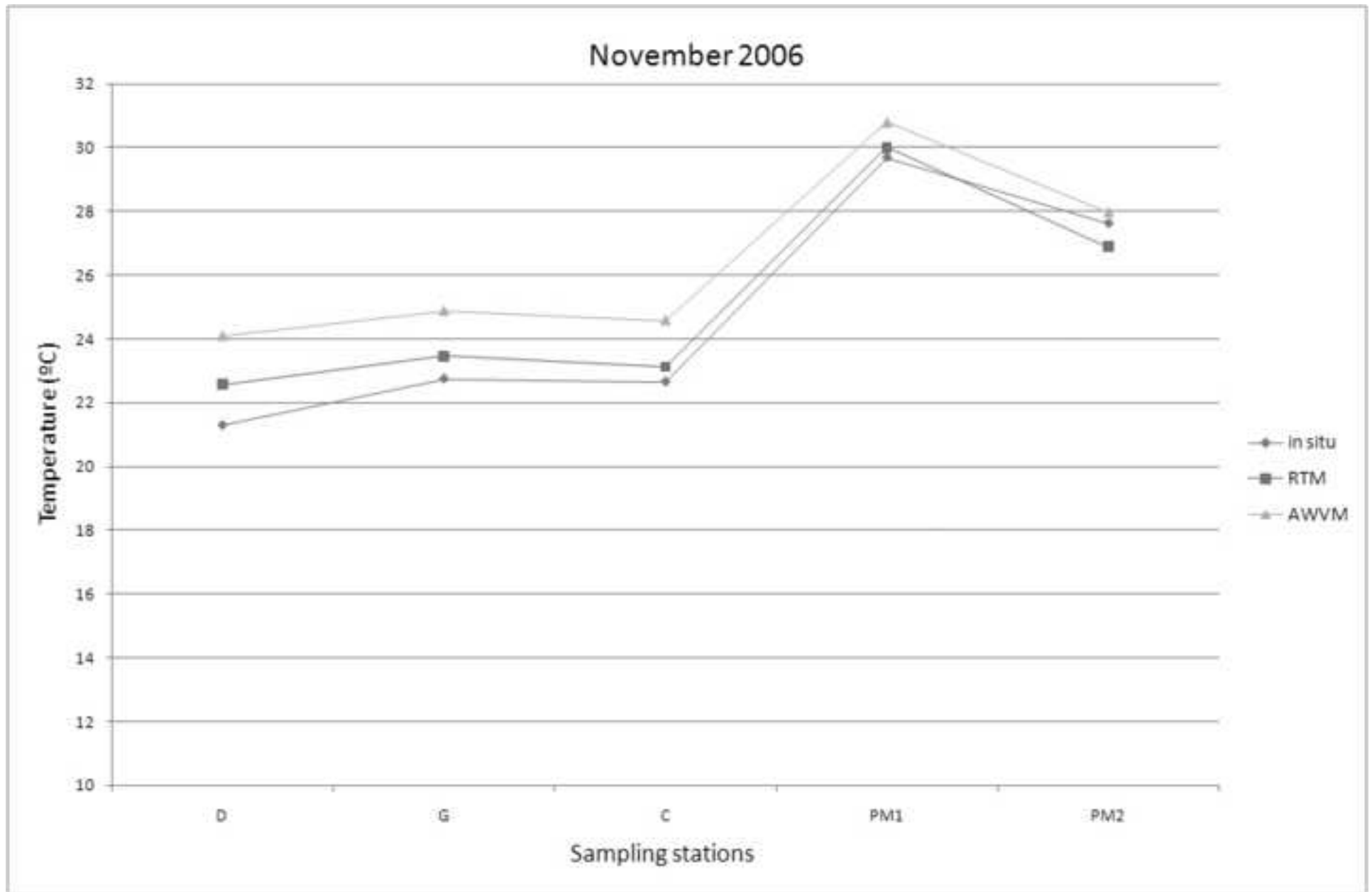
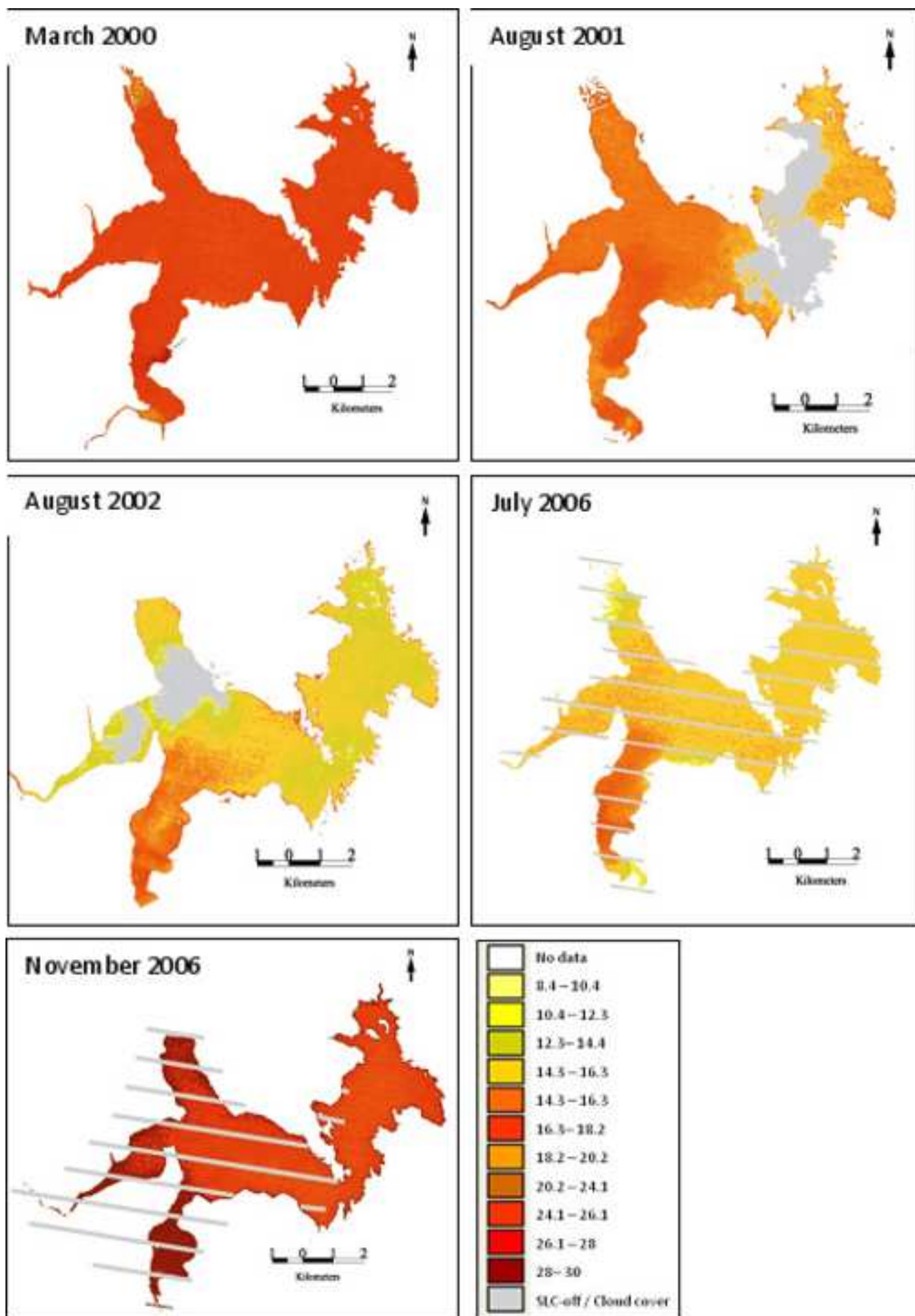
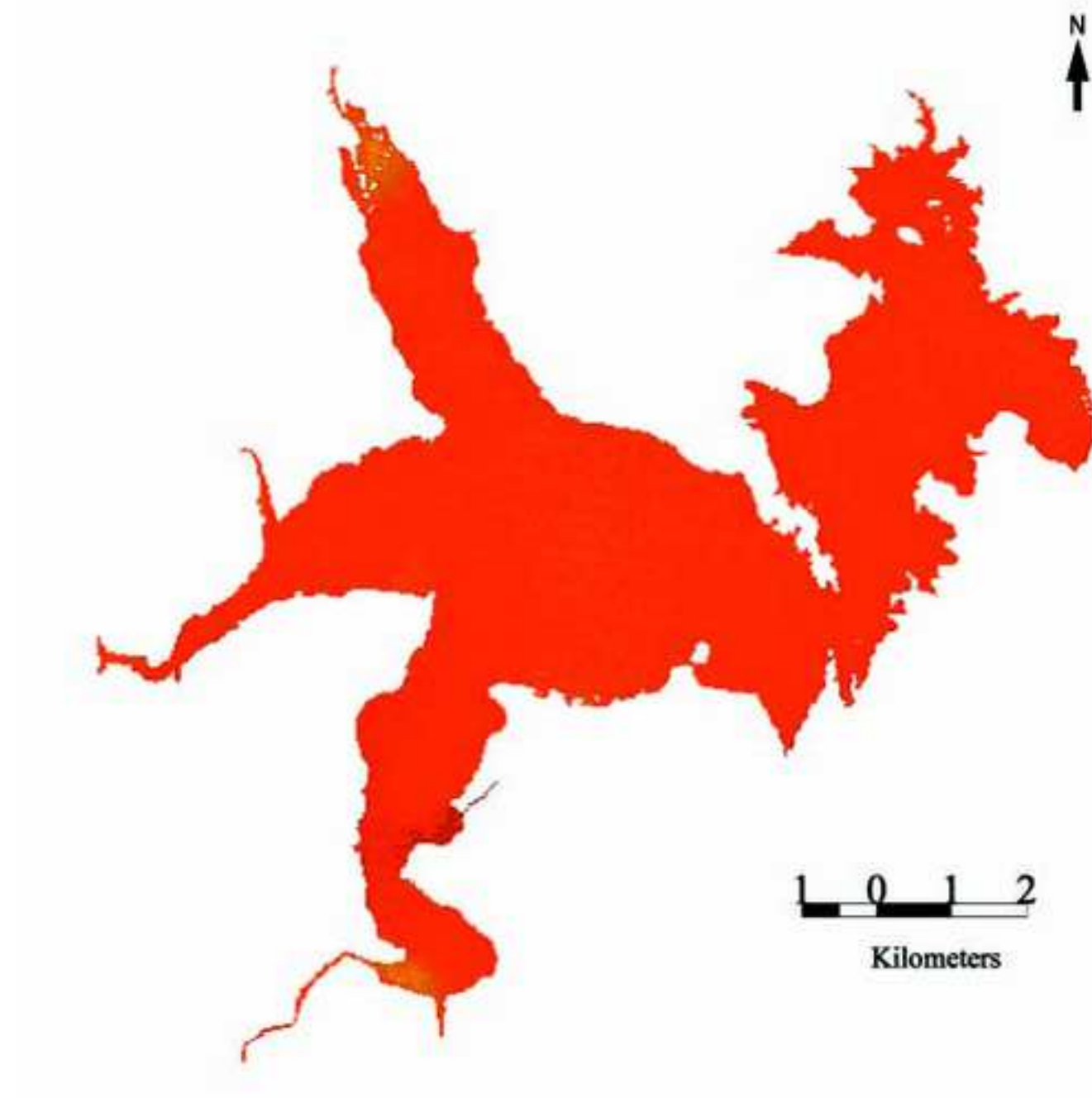


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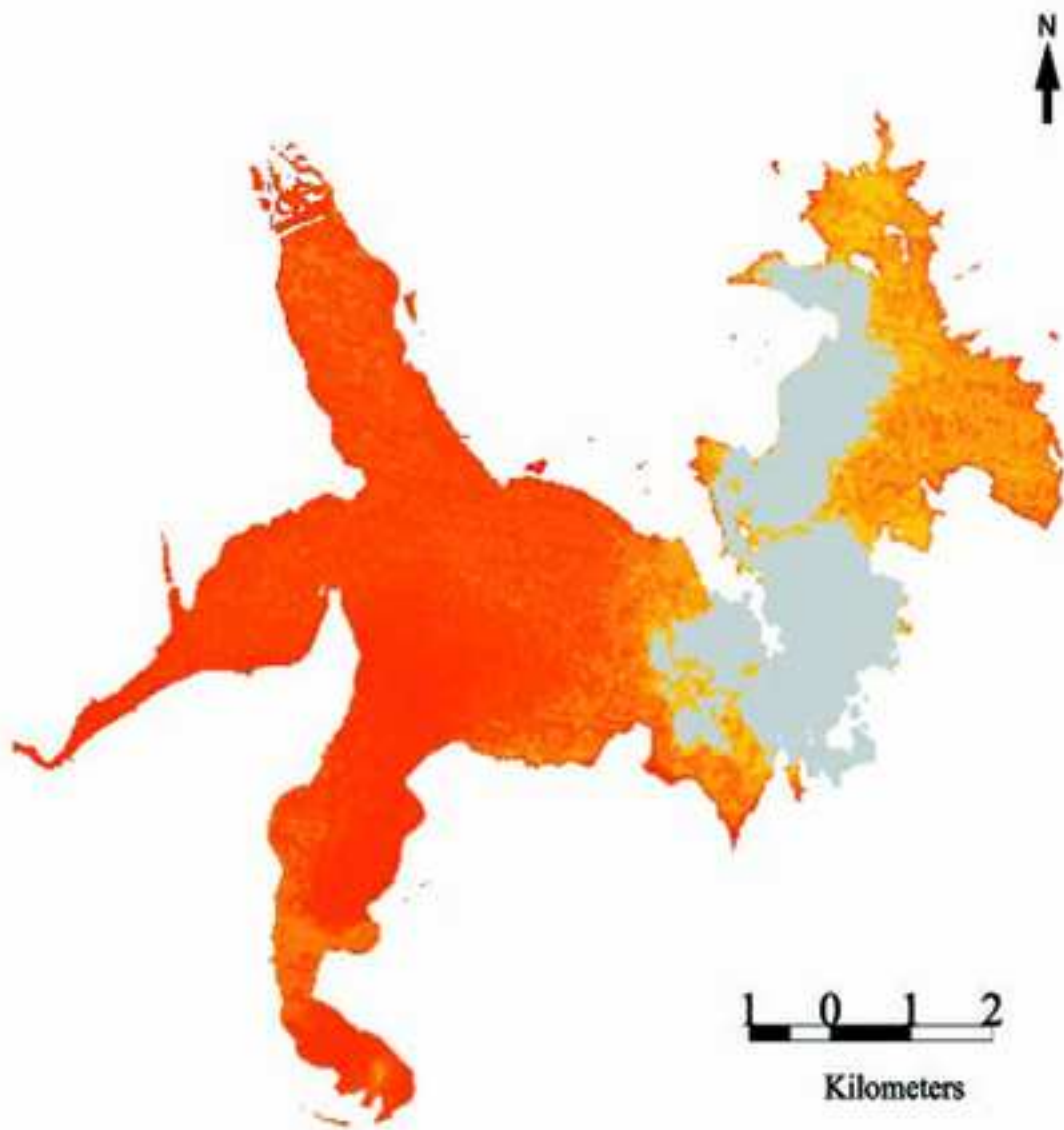
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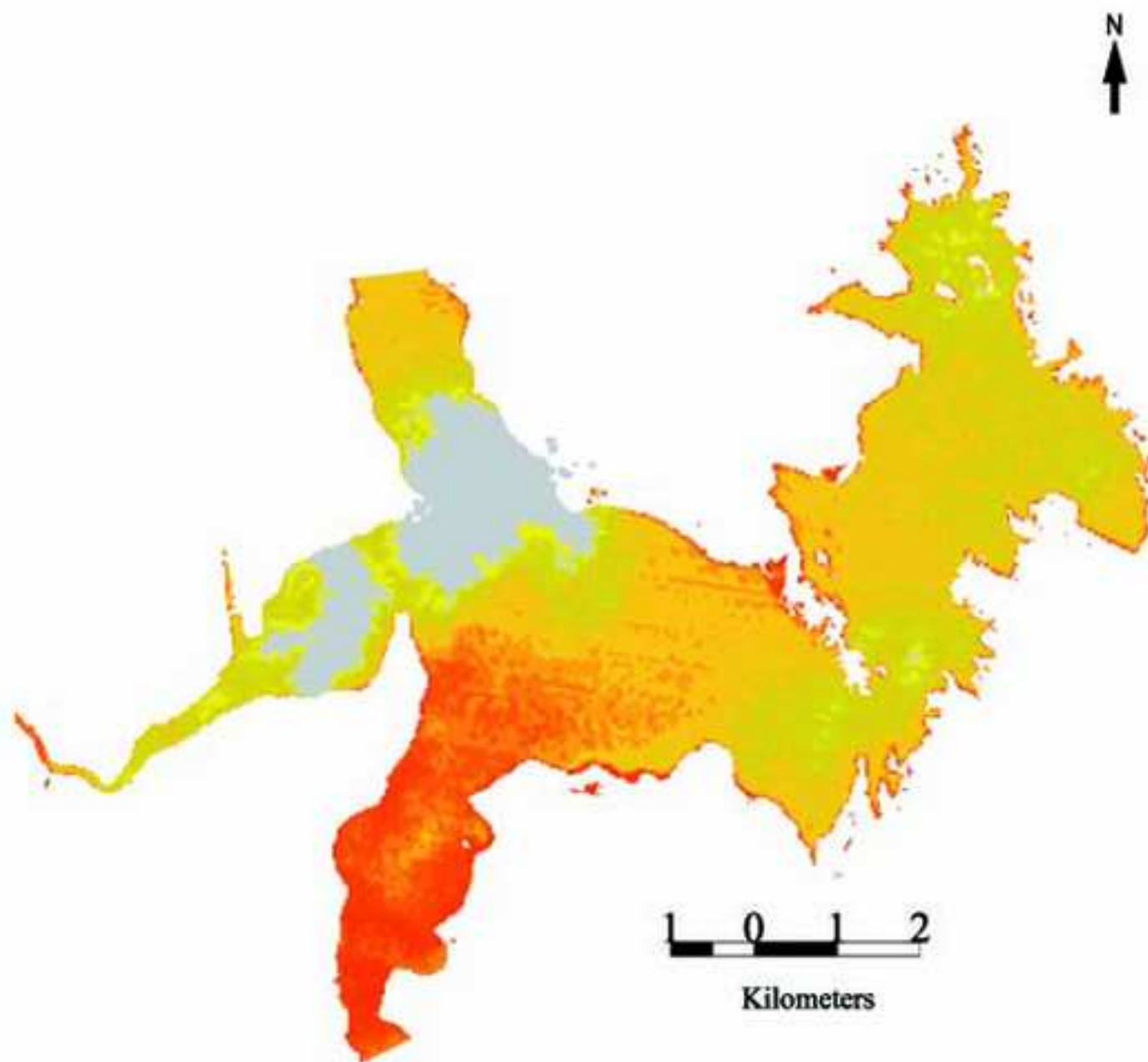


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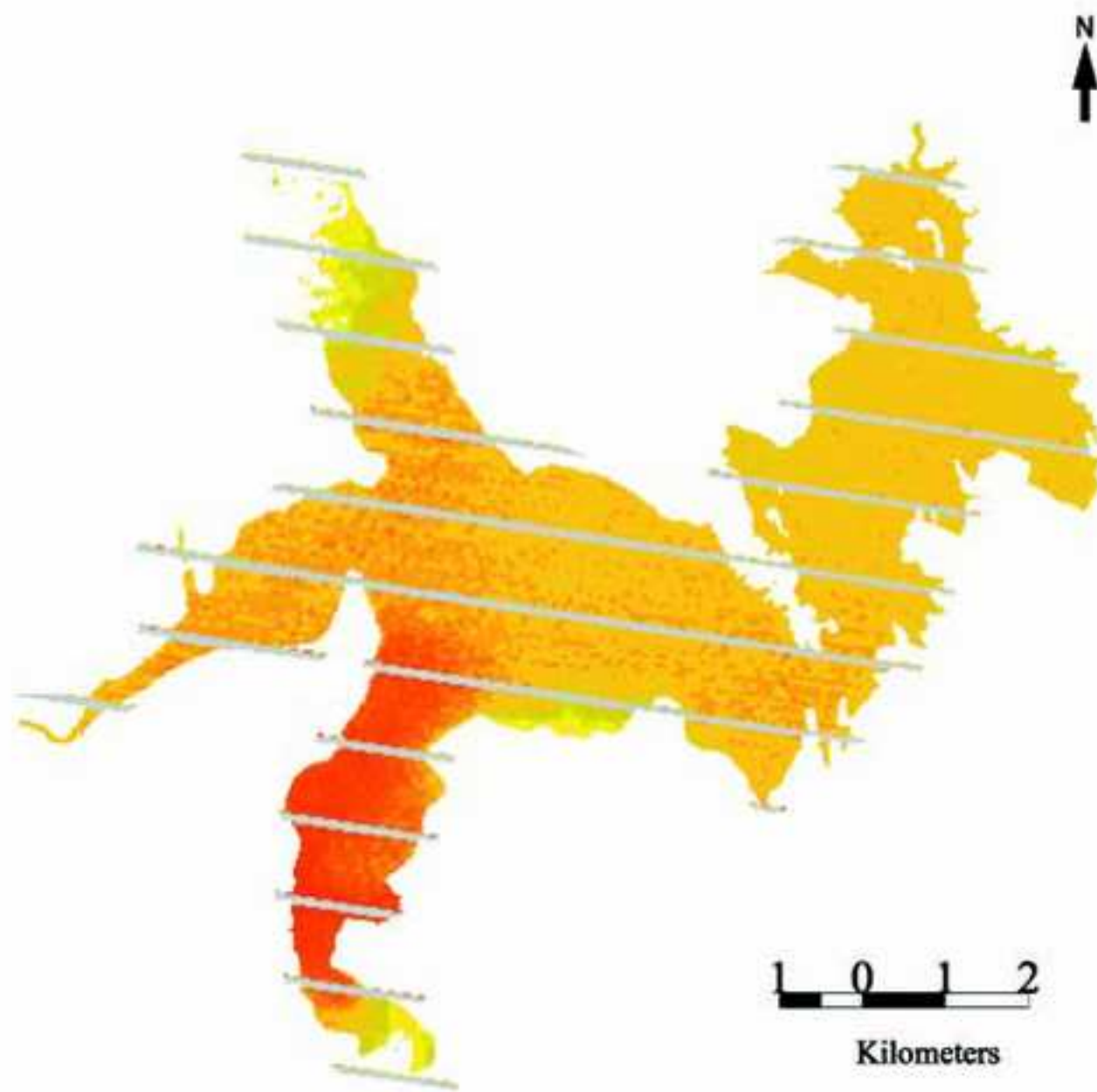
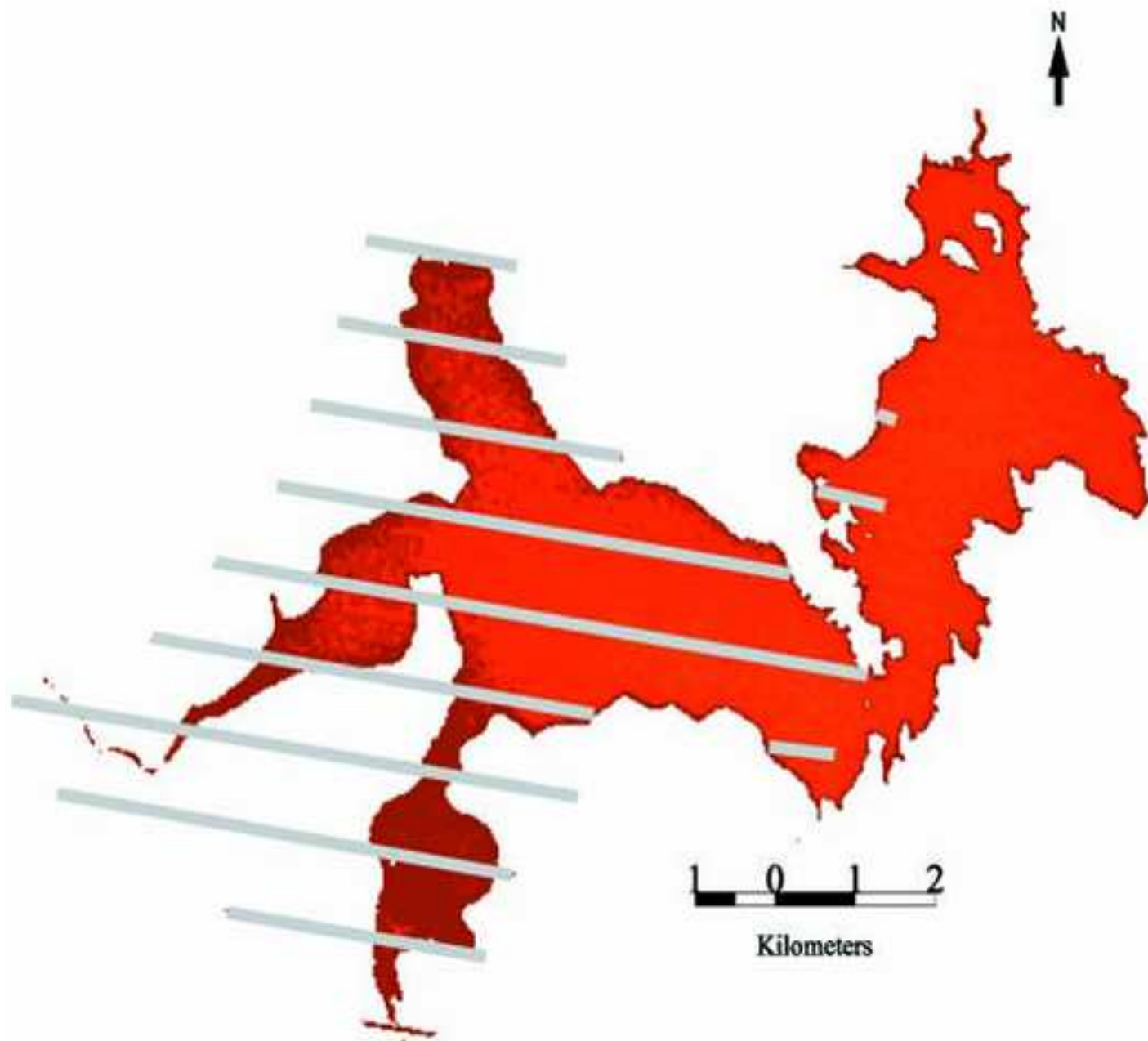


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