Method for classifying sites to Atlantic Rainforest restoration aiming to increase basin’s streamflows

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We propose a method to classify priority sites for Atlantic Rainforest restoration aiming to increase basin streamflows. The Rainfall Forest to Water Production (RFWP) method uses multicriteria analysis supported by GIS techniques and hydrological modeling. The method was applied to the Itapemirim River Basin, southeastern Brazil. The application of RFWP provided a map of areas with different priority for forest restoration by overlapping standardized numerical criteria with different weights (climatological, soil type/land use, and relief). The results indicated the influence of the wide distribution of the restoration sites on the streamflows. The RFWP proved to be suitable for the spatial analysis of the effect of different restoring areas on streamflows. Based on simulated scenarios, an increase in the native forest cover by restoration up to 27.6% of the basin area is expected to significantly enhance water production. The priority areas where forest restoration could better contribute to increase streamflows were delineated, especially at high altitude and in pastures, which are mostly in degraded conditions.

Keywords: GIS Application, Streamflow, Hydrologic Modeling, DHSVM

Introduction

The community seeks to restore the forest cover to improve the supply of ecosystem services (Lamb 2018). The Atlantic Forest biome covers 111 Mha (13% of the Brazilian territory – Souza et al. 2020) in highly heterogeneous environmental conditions. This biome is one of the Earth’s richest biodiversity hotspots (richer than most of the Amazon forests), with high endemism levels (Morelato & Haddad 2000, Ribeiro et al. 2009, Rezende et al. 2018). However, it is the Brazilian biome that suffered the most extensive, human-induced land use and land cover change in the past (Morelato & Haddad 2000). Isolated forest fragments cover 7 to 10% of the biome; most of these fragments are old secondary growth, surrounded by croplands, pasture, forest plantation, cities, and infrastructure (Souza et al. 2020). The high degree of human interference in this biome has affected regional water systems, causing sedimentation at water bodies, water pollution, loss of biodiversity, and reduced streamflows (Costa et al. 2017). The water scarcity and the pollution caused by the accelerated landscape alteration at Atlantic Rainforest point out the need to establish adequate policies for environmental management, especially of the water resources. The potential for maintenance of water quality and regularization of water flow is one of the main ecosystem services offered by the forests (Ferraz et al. 2013, Ellison et al. 2017). Forest cover is associated with water quality improvement, although its relationship in terms of quantity is still controversial (Ellison et al. 2017, Filoso et al. 2017, Bennett & Barton 2018). Previous research showed that in most cases the forest is more related to the regulation of streamflows, rather than being directly linked to the increase or decrease in average annual streamflow (Bruijnzeel 2004, Ferraz et al. 2013, Brogna et al. 2017, Kim et al. 2018).

Forests have the positive hydrologic impact of increasing water infiltration, with a consequent reduction in overland flow and erosion, associated with more significant recharge from aquifers and higher subsurface flow (Bruijnzeel 2004, Ellison et al. 2017). On the other hand, the increase in forest cover is usually associated with the increase in evapotranspiration and rainfall interception (Ellison et al. 2017, Filoso et al. 2017). The potential impact of forest restoration in the basin streamflow is, in a simplified way, the final balance between the positive effects (increasing infiltration) and the adverse effects (increasing evapotranspiration and rainfall interception).

The variables that control the basin hydrology are often interdependent, and their magnitude varies in importance among different basins. The combination of soil and climatic conditions, associated with the wide distribution of land uses, are factors affecting the hydrology of catchments (Dean et al. 2015, Ilistedt et al. 2016, Lozano-Baez et al. 2018). In this respect, assessing the impacts of forest restoration on streamflows must be careful and take into account the wide distribution of the afforested sites, since the land-use change can present a positive, negative or null effect to water production. Therefore, the seek for consistent ways to identify priority areas for forest restoration aiming to improve water availability is a problem that needs investigation (Dean et al. 2015, Alvarenga et al. 2017).

Intending to provide support for territorial planning, some multicriteria approach techniques associated with Geographic Information Systems (GIS) have been applied in the identification of “best” sites to a

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specific goal. This kind of approach, mainly due to the association with GIS, can combine several spatial factors in a single map obtained through overlaying different criterion maps (Malczewski & Rinner 2015, Bonilla Valverde et al. 2016, Vettorazzi & Valente 2016). Regarding forest sites, these techniques were used to identify priority areas aiming to increase aquifers recharge (Rahman et al. 2013, Bonilla Valverde et al. 2016, Rahmati et al. 2016), forest conservation (Vettorazzi & Valente 2016, Valente et al. 2017, Leal et al. 2019), or to improve water quality (Mello et al. 2018). However, there is still a lack of studies focused on identifying sites where forest restoration is most suitable to increase streamflows of catchments, i.e., to increase the basin water availability.

As a premise of this paper, we assume that the application of multicriteria analysis in a GIS environment intending to select areas to achieve a specific objective – such as the water availability – can present more accurate results when supported by simulation models able to predict the effects of land-use changes in the environment, such as basin streamflows. In this sense, hydrological models are useful tools to predict spatial patterns of hydrological processes with land-use changes. Thus, these models configure outstanding support for environmental planning and decision-making that provides quick and low-cost responses.

The main objective of this paper was to propose and apply a method to identify sites with different levels of priority for the restoration of the Atlantic Rainforest, aiming to increase the basin’s water availability (streamflows). The multicriteria analysis supported by GIS and hydrological modeling was taken as the pathways to achieve this goal.

Material and methods

The present paper is divided into three parts. The first comprises the presentation of the method named “Rainfall Forest to Water Production” (RFWP), proposed to identify sites with different levels of priority for the Atlantic Rainforest restoration to increase the water availability (streamflows) of basins. The RFWP requires some spatial information on the basin, which are supported by GIS techniques. The second part describes the application of RFWP to a specific basin. Finally, the third part describes the hydrological modeling of an Atlantic Rainforest basin, conducted by simulating streamflows of different forest restoration scenarios, as built from the RFWP results.

Rainfall Forest to Water Production (RFWP)

The RFWP was developed considering multiple criteria (or factors) for allocating specific land use (Atlantic Forest), identifying the sites where environmental conditions are more or less favorable to improve the recharge of aquifers and increase water availability. The definition of the sites’ priority levels passed through the elaboration and overlay of gridded maps containing the spatial information for each criterion. The following steps were taken: (i) identifying the most relevant set of criteria related to water production; (ii) obtaining a significant set of ratings and weights for each criterion; (iii) identifying an adequate technique to combine the gridded maps of criteria that reflect the relationship between the factors and the objective.

Selection of the criteria (factors) related to water production

Three simple spatial characteristics were chosen as criteria to identify the most suitable locations for forest restoration to increase water availability: one climatological, one related to soil and land use, and one related to the relief. Gridded data sets of each criterion were created. Subsequently, the following procedures were taken with the gridded datasets to achieve the objectives of the RFWP: standardization, definition hierarchy, and overlapping. The selected climatological criterion was the difference between historical annual averages of precipitation and the potential evapotranspiration (P – PET). The reason of this choice was that water production is enhanced in sites with more water input (higher precipitation depth) and lower water “consumption” (lower evapotranspiration depth – Foulon et al. 2018, Pham et al. 2018). Thus, with few input variables, this criterion represents an initial estimate of the water depth available to be converted into water availability. P – PET is not an exact representation of water depth to be converted into streamflow. PET itself is not reached throughout the year, and actual evapotranspiration (ET) would be a better estimator. However, ET is not as simple to be determined as PET, depending on a large amount of meteorological, edaphic, and vegetational information which are not always available, especially at the Atlantic Rainforest biome (Cecílio et al. 2020).

Soil class and vegetation are determinant factors in assessing water availability (Campos et al. 2016), being incorporated into the RFWP as another criterion. The Curve Number (CN) parameter of the Curve Number Method was selected to reflect soil and land use attributes, as it intrinsically represents the potential infiltration of the area. The higher the CN (which varies between 0 and 100), the lower the potential infiltration, as it is dependent on factors related to the soil, the surface, and the management of the soil (Sartori et al. 2005). Since forests increase water infiltration (Cooper et al. 2013, Dean et al. 2015, Lozano-Baez et al. 2018, Lozano-Baez et al. 2019), the inclusion of this criterion allow the selection of areas where the change of land use from the original vegetation to the native forest will potentially increase the infiltration capacity as a consequence of afforestation.

The premise of using a criterion related to the relief is to reflect the distance to the nearest stream and the water table. RFWP considers that the further away from both, the more appropriate the site. Often crops with deeper roots (such as species from the Atlantic Forest) can uptake water from the aquifers (Beyer et al. 2018). Therefore, in sites distant from surface and groundwater, the root system of trees uses mostly water from the soil’s unsaturated zone, leaving the water table to increase its contribution to the streamflow. It should be made clear that the RFWP is proposed for the Atlantic Rainforest biome, where the distance from surface and groundwater sources does not limit forest growth due to the abundant rainfall. In mountainous sites, where the soil drains faster thus limiting groundwater storage, the forest is responsible for maintaining more significant infiltration and, consequently, the aquifer recharge (Cooper et al. 2013, Dean et al. 2015, Lozano-Baez et al. 2018, Lozano-Baez et al. 2019).

RFWP uses a unique criterion that reflects the depth of the water table and the proximity to surface water together with the topography. To quantify the influence of relief on physical processes, such as the movement of water in the subsoil, we used topographic indexes (Liang & Chan 2017, Sheshukov et al. 2018), namely, the topographic wetness index (TWI – eqn. 1):

\[ TWI = \ln \left( \frac{a_j}{\tan \beta} \right) \]

where TWI is the topographic wetness index of the cell, \( a \) is the specific contributing area upslope cell \( j \) \((m^2)\), \( \beta \) is the average slope upslope cell \( j \) \(^{\circ}\). This index characterizes the surface water saturation zones and the water content in soils, spatially identifying the sites most suitable to saturation (Drover et al. 2015, Ediriweera et al. 2016). The sites located along the streams and around the springs are saturated zones that do not allow significant infiltration gains, contributing more to surface flow (Bressiani & Schmidt 2016).

Sites with higher TWI are those at lower altitudes, close to streams, where the water table is shallower, while the sites with lower TWI are drier and located on mountain top, where the water table tends to be deeper (Sørensen et al. 2006, Lima et al. 2012). Thus, it is preferred to restore the Atlantic Forest in sites with lower TWI.

Criteria standardization

In the multicriteria analysis, numerical values of each criterion are standardized over a normal range and then aggregated through a linear function of the product of weights and criteria values (Bonilla Valverde et al. 2016). To this purpose, we standardized the gridded dataset using a common scale of 256 levels which maintains the intrinsic characteristics of the original grid values (Mello et al. 2018) using the following equations (Vettorazzi & Valente...
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\[ x_{i,j} = \frac{R_{i,j} - R_{i,\text{min}}}{R_{i,\text{max}} - R_{i,\text{min}}} \times 255 \]  
\( i = 1 \) to \( i = \text{number of criteria} \)  
\( j = 1 \) to \( j = \text{number of grid cells} \)

where \( x_{i,j} \) is the standardized value of criterion \( i \) at cell \( j \), \( R_{i,j} \) is the original value of criterion \( i \) at cell \( j \), and \( R_{i,\text{max}} \) and \( R_{i,\text{min}} \) are the maximum and minimum observed values of the criterion \( i \). Eqn. 2 was applied to both \( \text{P-PET} \) and \( \text{CN} \), where the highest values indicate the desired conditions according to the study’s objective (increase water availability), while eqn. 3 was applied to the \( \text{TWI} \) criterion, where the lowest values is the desired condition. After standardization, the grid values for each criterion ranged between 0 (unwanted conditions) and 255 (desired conditions).

### Assignment of the criteria’s weights

The weight expresses the relative importance of each criterion (Valente et al. 2017) as different criteria affect water availability. To achieve the weight of each standardized criterion, RFWP uses a decision-making process known as Analytical Hierarchical Process (AHP - Saaty 1977), which employs a pairwise comparison (two by two) between the criteria to determine their relative importance and calculate their respective weights, whose final sum must be equal to one.

The \( \text{P-PET} \) criterion was the most important because it represents the total volume of water available to be converted in streamflow. TWI was considered the second most important as it indicates the water table’s depth, the potential consumption of groundwater, and the aquifer recharge. Finally, NC was considered the less critical criterion.

### Combination of standardized criteria

RFWP proposes the overlap of the standardized criteria grids based on the aggregation of the product of the weights and values of the standardized criteria (Malczewski & Rinner 2015), according to eqn. 4:

\[ S_i = \sum w_i x_{i,j} \]  
where \( x_{i,j} \) is the standardized value of the \( i \)-th criterion at the \( j \)-th grid cell, \( S_i \) is the resulting numerical value for the cell \( j \) and \( w_i \) is the weight of criterion \( i \). High values of \( S_i \) indicate sites with high priority and low value areas with low priority. Fig. 1 shows an example of a set of criteria and their relationship to the study’s objective after applying RFWP.

### Application of the RFWP

The hydrological modeling of an Atlantic Rainforest basin was performed using the “Distributed Hydrology Soil-Vegetation Model” (DHSVM – Wigmosta et al. 1994) to evaluate RFWP. The basin under study was the Itapemirim River Basin upstream of the “Usina Paineiras” stream gauge station – BHPAIN (Fig. 2). The basin is located in the south of Espirito Santo (Brazil), between the geographic coordinates 20° 10’ S and 21° 15’ S; and 40° 55’ W and 41° 50’ W, with an area of approximately 5170 km². According to the Köppen classification, the regions’ predominant climatic types are “Cwa” (Humid Subtropical climate with dry winter and hot summer) and “Cwb” (Humid Subtropical climate with a dry winter and temperate summer). The altitudes vary from 5 to 2854 meters, and the relief is dominated by mountains, which indicates suitability for applying the DHSVM.

The gridded data that characterize the soils and the land use (for 2015) at BHPAIN were obtained from the Institute of Environment and Water Resources of Espirito Santo (GEOBASES 2015). BHPAIN’s land uses were determined from aerial photographs using remote sensing photo-interpretation techniques (GEOBASES 2015) and showed that the three mainland uses are pastures (42% of the area), forests (23.4%), and coffee plantations (15.9%). According to the Brazilian Soil Classification System, the predominant soil types in the basin are Interceptosols, Ultisols, and Oxisols, representing 31.6%, 30%, and 20.4% of the BHPAIN. The gridded dataset used in the application of RFWP to BHPAIN was standardized to 90 m spatial resolution, indicated as the most suitable to hydrologic modeling of BHPAIN with DHSVM model (Mendes et al. 2020).

The climatological criterion considers the Brazilian daily gridded database containing rainfall depths, with a spatial resolution of 0.25° x 0.25° (Xavier et al. 2016), which was used to compute the mean annual rainfall.
depths (P). This dataset was resampled to the 90 meters spatial resolution, using the Inverse Distance Weighting interpolator (Xavier et al. 2016).

The historical average of the monthly reference evapotranspiration (PET$_{ref}$) was calculated using the Hargreaves & Samani equation calibrated for Espírito Santo, according to eqn. 5 (Zanetti et al. 2019). The sum of PET$_{ref}$ is the average monthly evapotranspiration (PET – eqn. 5).

$$\text{PET}_{\text{ref}} = \text{MND} \cdot 1.08 \cdot \left[ 0.0023Q_s \left( T_{\text{max}} - T_{\text{min}} \right) ^{0.5} \left( 17.8 + T_{\text{min}} \right) - 0.81 \right]$$

where PET$_{\text{ref}}$ is the average monthly evapotranspiration (mm), $Q_s$ is the monthly extraterrestrial solar irradiation (mm day$^{-1}$), $T_{\text{max}}$ is the historical average monthly maximum air temperature (°C), $T_{\text{min}}$ is the historical average monthly minimum air temperature (°C), and MND is the month’s number of days (day).

In order to consider the variation of evapotranspiration in altitude, the historical average air temperatures were calculated using multiple linear regression equations established for Espírito Santo State, according to the general model in eqn. 6 (Castro et al. 2010):

$$T_i = \beta_0 + \beta_1 \text{Alt} + \beta_2 \text{Lat} + \beta_3 \text{Lon}$$

where $T_i$ is the historical average monthly air temperatures (maximum, minimum, or mean - °C), Alt is the altitude (m a.s.l.), Lat is the latitude (°), Lon is the longitude (°), and $\beta_0$, $\beta_1$, $\beta_2$, $\beta_3$ are regression parameters (Castro et al. 2010).

A SRTM digital elevation model (DEM) with a spatial resolution of 90 meters was used. The DEM was manipulated with GIS tools to become a hydrologically conditioned digital elevation model (HC-DEM) to serve as input data for the DHSVM model. The calculations of the variables $Q_p$, $T_{\text{max}}$, $T_{\text{min}}$, $\text{PET}_{\text{ref}}$ (eqn. 5, eqn. 6), and PET were all performed with a GIS tool based on the HC-DEM. The maps containing the variables $P$ and PET were then subtracted to generate the $P$ – PET gridded dataset.

The establishment of the CN gridded dataset (soil and land use criterion) passed through the association between the land use classes and the soil hydrological groups. BHFPAIN soils were classified into hydrological groups based on the recommendation to Brazilian soils (Sartori et al. 2005). Therefore, the Curve Number method tables were used to associate land use and soil group to CN values. This procedure generated a gridded file with the 90 m spatial resolution adopted. Areas whose land use could not be converted to forests (rocky sites, mining sites, urbanized areas, and water bodies) were excluded from the posterior analysis by assigning the “No Data” information.

Finally, TWI was calculated applying eqn. 1 to the HC-DEM, with the SAGA-GIS Wetness Index algorithm (Conrad et al. 2015). The criteria weights ($\omega_i$) were defined based on the paired comparison using AHP (Saaty 1977) as described above, resulting in the following values: (i) $P$ – PET: $\omega_i = 0.5813$; (ii) TWI: $\omega_i = 0.3092$; and (iii) CN: $\omega_i = 0.1096$.

The three criteria were combined using the standardized criteria grids ($x_i$) and their respective weights ($\omega_i$), resulting in a single priority map with values between 0 and 255. The combined priority values were then reclassified into eight priority classes, from 1 (most suitable sites for forest restoration) to 8 (least suitable sites). This reclassification was performed using the GIS natural breaks method, which maximizes the difference between classes.

Hydrological modeling of land-use change scenarios

The effectiveness of RFWP in achieving its objectives was verified through hydrological modeling, using the DHSVM model (Wigmesta et al. 1994). For the application of the DHSVM, we used the input data of a previous calibration for BHFPAIN (Mendes et al. 2020). More information about the values of the DHSVM’s input parameters used, are reported by Mendes et al. (2020).

The modeling consisted of building 16 scenarios of restoration of the native forest (Tab. 1) in sites belonging to each of the eight classes of priority (and their combinations), thereby verifying the impact over the water availability. Only sites with agriculture, pastures, exposed soil, vegetation, and early-stage regeneration were considered eligible to be converted into Atlantic Forest. In all scenarios, daily streamflow series for BHFPAIN were simulated using the daily meteorological data from 2008 to 2011. The evaluation of each scenario in matching the objectives was performed by the analysis of relative changes in two reference flows: the average streamflow ($Q_{\text{avg}}$) and the minimum streamflow, represented by the flow of permanence in 90% of the time ($Q_{\text{p90}}$). The simulated scenarios are described in Tab. 1.

The steps for applying the RFWP and verifying land-use changes with hydrological modeling are summarized in Fig. 3.

Results and discussion

Application of RFWP

The final result of RFWP to BHFPAIN is shown in Fig. 4a, and its reclassification in the eight different classes is shown in Fig. 4b.

The highest S-values of Fig. 4a represents sites most suitable for the Atlantic Rainforest restoration to increase water availability. The maximum value of S was 231, and the minimum was 9. The “blank” areas are those where land use cannot be changed (rocky sites, mining sites, urbanized areas, and water bodies).

Comparing Fig. 4 and Fig. 5b, one can observe that the criterion $P$ – PET was determinant to delineate the three classes of higher priority (1, 2, and 3) due to its higher $\omega_i$. In classes 4 and 5, there is a prevalence of TWI, but the classes are still influenced by the $P$ – PET, especially in sites with good water availability (higher $P$ – PET). The delineation of classes 6, 7, and 8 was influenced by the TWI in sites close to water bodies and by lower $P$ – PET values.

Class 1 sites (S range: 171-231) were found in areas close to important remaining forest fragments in the upper basin: Northern-western portion (Caparaó) and Northeastern portion (Castelo and Vargem Alta mu-

| Tab. 1 - Forest restoration scenarios simulated based on the results of RFWP applied in the study area (BHFPAIN). |
|-----------------|-----------------|-----------------|-----------------|
| Scenario | Description | Area of forest restoration (km$^2$) | Area of forest restoration (%) |
|-------------|-----------------|-----------------|-----------------|
| SC0 | No land-use change (current scenario) | - | - |
| SC1 | Forest restoration at eligible sites of class 1 | 44.2 | 0.9 |
| SC2 | Forest restoration at eligible sites of class 2 | 153.2 | 3.0 |
| SC3 | Forest restoration at eligible sites of class 3 | 419.8 | 8.1 |
| SC4 | Forest restoration at eligible sites of class 4 | 807.2 | 15.6 |
| SC5 | Forest restoration at eligible sites of class 5 | 961.5 | 18.6 |
| SC6 | Forest restoration at eligible sites of class 6 | 781.2 | 15.1 |
| SC7 | Forest restoration at eligible sites of class 7 | 510.2 | 9.9 |
| SC8 | Forest restoration at eligible sites of class 8 | 211.8 | 4.1 |
| SCA12 | Forest restoration at eligible sites of class 1 and 2 | 197.4 | 3.8 |
| SCA13 | Forest restoration at eligible sites of class 1 to 3 | 617.2 | 12.0 |
| SCA14 | Forest restoration at eligible sites of class 1 to 4 | 1424.3 | 27.6 |
| SCA15 | Forest restoration at eligible sites of class 1 to 5 | 2385.9 | 46.2 |
| SCA16 | Forest restoration at eligible sites of class 1 to 6 | 3167.1 | 61.3 |
| SCA17 | Forest restoration at eligible sites of class 1 to 7 | 3677.3 | 71.2 |
| SCA18 | Forest restoration at eligible sites of class 1 to 8 | 3889.1 | 75.3 |
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The sites of class 2 (S range: 147-171) are situated around class 1 sites and reaches areas under pasture, agriculture, and also remaining forest fragments in the northwestern (Ibitirama) and northeastern (Castelo, Vargem Alta, and Venda Nova do Imigrante municipalities) headwaters of BHPAIN. Some sites at the high altitude of the BHPAIN’s center were also class 1 (Alegre, Cachoeiro de Itapemirim, Castelo, Muiz Freire, and Iúna). The forest fragments of class 1 and class 2 sites include the Caparoá National Park, the Forno Grande State Park (Castelo), and the part of the Pedra Azul State Park belonging to BHPAIN (Vargem Alta), which are protected areas included by the government as a priority for biodiversity conservation. Therefore, the sites of these classes can be taken as fundamental not only for water production but also for biodiversity conservation.

Sites class 3 (S range: 129-147) and site class 4 (S range: 115-129) include pastures, agriculture crops, and forest fragments at intermediate altitudes and in the middle basin. In class 5 (S range: 102-115), pastures at intermediate altitudes are predominant. Classes 6 (S range: 88-102), class 7 (S range: 72-88), and class 8 (S range: 9:72) occupies the lower basin and the lands along the valleys near the streams.

Our results confirmed that the RFWP decision rules applied can be considered adequate, since the priority sites for restoration were delineated in major P - PET areas far away from the water table and streams and with a significant potential infiltration. Indeed, these sites present the potential for the infiltration gain to exceed the evapotranspiration losses.

For the application of the RFWP in other areas or biomes, it is recommended to test different weights and the relative importance of criteria to assess which process is dominant to achieve the objective (Schneiderman et al. 2007), or even include additional criteria. A sensitivity analysis must be incorporated to check the ordering of priority and the criteria' weights (Malczewski & Rinner 2015).

Hydrological modeling of land-use change scenarios

The percentages of change in the reference flow for each scenario compared to current situation (SC0) are shown in Fig. 5a. Since the results in Fig. 5a may be masked according to the different restored areas of each class, the rate of variation of the flows (in %) as the surface (in km²) covered by native vegetation increases in the study area is displayed in Fig. 5b.

Fig. 6 shows that the percentage change in streamflows were more pronounced for Q90 than for Q9 as for all the scenarios, suggesting that forest restoration in BHPAIN could affect more significantly the annual water availability than that of dry periods. The discussion of the results presented in Fig. 6 will follow two main aspects: the comparison of the scenarios related to forest restoration individually in each class (SC1 to SC8), and the analysis of the scenarios that simulated the restoration of more classes together (SCA12 to SCA18).

Comparison of different scenarios of forest restoration (SC1 to SC8)

Fig. 5a shows that class 5 (SC5) is the one...
with the most significant change in $Q_{\text{ave}}$ (1.29%) and $Q_{90}$ (0.87%), while the order of the other classes’ impacts varies considering $Q_{\text{ave}}$ or $Q_{90}$. Forest restoration in SC7 and SC8 (worse to water production) contributed to the reduction of $Q_{\text{ave}}$, showing that RFWP results are consistent. The root system of the forest, in classes 7 and 8, tends to be closer to groundwater sources.

In these areas, the forest maintains the evapotranspiration closer to PET for a longer period, increasing water consumption (Móricz et al. 2016). Therefore, forest restoration in these areas contributes more to water consumption (Gríbovszki et al. 2017) than with the increase of infiltration, resulting in the reduction of water availability in dry periods. Based on the above considerations, we concluded that site belonging to classes 1 to 6 are more suitable for forest restoration aimed to water production than sites belonging to classes 7 and 8.

Fig. 6 presents the percentage of BHLPAIN area affected by land-use changes in the simulated scenarios SC1 to SC8, where only a little afforestation occurs at SC1 (0.86% of the basin’s area), SC2 (2.97%), and SC3 (8.13%) compared to the other scenarios (SC4 to SC6). SC4, SC5, SC6 scenarios simulated an effective change of pastures and abandoned pastures to forest (about 15 to 19% of BHLPAIN). Although classes 1 to 3 are the priority levels, their area available for forest restoration is smaller (Tab. 1, Fig. 6). This confirms the minor effect of SC1 to SC3 over the streamflows (Fig. 5a), and highlights the need to evaluate the relative importance of afforestation in each scenario, which can be calculated by dividing the percentage change in the streamflow by the restored area (Fig. 5b), resulting in the percentage change in streamflow per 1.0 km$^2$ of afforested area.

Fig. 5b shows that the SC1 has the highest rate of change in flows per km$^2$ of forest restoration. For all the other scenarios, the flow gains are decreasing. For $Q_{\text{ave}}$, the rate of change in SC1 is almost double that for SC3, suggesting the effectiveness of forest restoration for water production in these areas. The results in Fig. 5b support our initial assumption on the existence of areas more suitable for increasing water availability, and the effectiveness of method adopted in separating sites based on their potential increase of streamflow after forest restoration.

Combination of forest restoration scenarios (SCA12 to SCA18)

The SCA12 to SCA18 scenarios consisted of combining two or more classes of sites eligible to forest restoration. According to Fig. 5a, the most substantial increment for $Q_{\text{ave}}$ is predicted under SCA18, which simulated the most significant increase in forest cover (388.1 km$^2$ or almost 75% of the basin area). The results of scenarios SCA15 to SCA18 showed little difference from

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![Fig. 5 - Results of the simulated scenarios of the Atlantic Forest restoration at BHLPAIN. (a) Percentage of change in $Q_{\text{ave}}$ and $Q_{90}$; (b) percentage variation of the streamflows per increment of the Atlantic Forest area.](image)

![Fig. 6 - Change of each land use (expressed as percentage of the basin area) considering the afforestation of all eligible areas inside each suitability class for forest restoration at BHLPAIN.](image)
each other. The more significant increase of $Q_{0w}$ was obtained under scenario SCA16, which simulated forest restoration over an area of 3167.1 km² (61% of the whole basin). Yet, the increase in both flows was not high, being only 4.72% for $Q_{0w}$ and 2.45% for $Q_{0s}$. This corroborates the results of previous studies on the BHPAIN, which demonstrated that the rainfall exerts more influence than forest cover over BHPAIN streamflow (Mendes et al. 2018).

Fig. 5a also suggests that the combined scenarios (SCA) could lead to higher increases in streamflow than the individual scenarios (SC). Indeed, a considerable increase in the streamflows was observed under SCA12 to SCA15, while smaller increases in the $Q_{0w}$ and $Q_{0s}$ are predicted by the subsequent scenarios. However, Fig. 5b shows that the percentage gain in $Q_{0s}$ per km² of afforestation is expected to be higher under SC12 to SC14. Therefore, the sites of classes 1 to 4 are priority sites, and have more potential to increase $Q_{0w}$ and $Q_{0s}$ and therefore water availability.

Overall, our results delineate the areas of the BHPAIN where forest restoration could better contribute to increase streamflows, especially pastures, which are mostly in degraded conditions (Rocha Junior et al. 2017). Based on simulated scenarios, an increase in the native forest cover up to 27.6% of the BHPAIN area, namely at sites indicated by classes 1 to 4 (SCA14), is expected to significantly enhance local water production.

General considerations about RFWP

The results found are the first step towards spatial planning aimed to identify suitable sites for afforestation to support the environmental and socioeconomic functions of the landscape. The results of RFWP must be assessed carefully for each basin, as it could have negative consequences in different ecosystems, such as in the warmer regions with high evapotranspiration. For example, it has been reported that an intermediate forest cover in regions with drier climate can maximize groundwater recharge (Listedt et al. 2016, Oliveira et al. 2016).

The RFWP is a useful tool that allows the delineation of sites where forest restoration potentially changes the balance between infiltration and evapotranspiration, thus achieving net water production. Our results showed that in specific sites, the forest impact on the streamflow was positive, while in other sites the flow is reduced. The results of hydrological modeling confirmed the suitability of the method used in this study for spatial analysis and mapping.

Land use planning and management can greatly benefit from the use of RFWP to identify the areas of favorable conditions for the infiltration and aquifer recharge, increase the water availability of catchments and, as an immediate consequence, reduce erosion processes. The benefits of forest restoration in the study area can be enhanced by adopting best management practices (Cristan et al. 2016) and sustainable agricultural systems (Wilson & Lovell 2016), such as agroforestry. In this context, due to the vast predominance of pastures at BHPAIN, the adoption of Integrated Crop-Livestock-Forest Systems (ICLFS) is recommended to achieve sustainability (Alves et al. 2017).

Finally, identifying priority areas for forest restoration intending to increase water availability is still a challenging problem. Further research is encouraged, especially in different regions, to test the applicability of the RFWP. Besides, we consider essential the validation of RFWP with observational data in the field.

Conclusions

The methodology applied in this study (RFWP) allowed to map sites with different levels of priority for the restoration of Atlantic Rainforest aimed at increasing water availability (streamflows) in the BHPAIN (SE Brazil). The hydrological modeling proved that the method adopted is effective for the above purposes. However, caution should be taken in the choice of criteria to be included and in the weighting of criteria, and their prior assessment is recommended to improve the RFWP performance. Overall, our results support the use of RFWP as a helpful tool in land use planning and management.

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