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Identification of river basins within northwestern slope of Crimean Mountains using various digital elevation models (ASTER GDEM, ALOS World 3D, Copernicus DEM, and SRTM DEM)

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Since the end of the 20th century, the use of geographic information systems and digital elevation models has reduced the time required for and improved the quality of morphometric analysis of the relief within river basins. However, researchers are constantly faced with the problem of choosing the most accurate and suitable digital terrain model for their task. Many global, regional, and local digital elevation models are available. In this study, we comparatively analyzed the accuracy of the ASTER GDEM, ALOS World 3D, Copernicus DEM, and SRTM DEM spatial datasets for the purpose of catchment basin modeling for the river basins of the northwestern slope of the Crimean Mountains (Zapadnyy Bulganak, Alma, Kacha, Belbek, and Chernaya Rivers) as an example. For each river basin, we calculated the systematic, root mean square, mean absolute, standard root mean square (Bessel's correction), and centered mean absolute errors by comparing ASTER GDEM, ALOS World 3D, Copernicus DEM, and SRTM DEM data with a 1:100,000 topographic map within the considered river basins. We found the smallest error values for the Copernicus DEM and ALOS World 3D datasets; furthermore, we used the Copernicus DEM dataset to model the river basins and sub-basins of the northwestern slope of the Crimean Mountains. As a result, we identified these river basins and sub-basins for the Zapadnyy Bulganak, Alma, Kacha, Belbek, and Chernaya Rivers, which are represented by stream basins, valleys, gullies, and ravine systems.

KEYWORDS

GIS, digital elevation model, ASTER GDEM, ALOS World 3D, Copernicus DEM, SRTM, river, river basin

1 Introduction

Digital elevation model (DEM) is a generic term for digital topographic and/or bathymetric data in all their forms (Manune, 2007). A large number of studies have been devoted to assessing the accuracy of DEM, considering both the practical and theoretical aspects of this issue (del Rosario Gonzalez-Moradas and Viveen, 2020; Mesa-

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Mingorance and Ariza-López, 2020; Uuemaa et al., 2020; Yamazaki et al., 2017). The number of these scientific studies has been constantly increasing due to the improvement in existing geodatasets and the market entry of new sets of geodata, which is of considerable interest to researchers. In recent years, DEMs created using unmanned aerial vehicles (UAVs) (Uysal et al., 2015; Hashemi-Beni et al., 2018; Escobar Villanueva et al., 2019; Annis et al., 2020) have been extensively used. UAVs have high accuracy but, in almost all cases, are inaccessible to a wide range of researchers. DEMs are actively used in the study of glaciers (Fischer et al., 2015; Bodin et al., 2018), forests (Balzter et al., 2015; Liu et al., 2018), and celestial bodies (Florinsky and Filippov, 2017; Fawdon et al., 2018), among others. One of the largest niche areas is occupied by studies of the morphometry of river basins (Pyankov and Shikhov, 2017; Fang et al., 2019; Sarkar et al., 2020; Yermolaev et al., 2021; Zhao et al., 2021), in which DEMs are actively used to identify and characterize river basins.

ASTER and SRTM DEMs have been thoroughly compared. For example, Rajasekhar et al. (2018) studied lineament extraction from ASTER DEM, SRTM, and Cartosat for the Jilledubanderu River basin, Anantapur district, India. Thomas and Prasannakumar, 2015 studied basin morphometry derived from topographic maps, ASTER, and SRTM DEMs, considering an example from Kerala, India. Nikolakopoulos et al. (2006) compared ASTER and SRTM DEMs in Greece using two regions of Crete Island. Zhao et al. (2021) compared the performance among typical open global DEM datasets for the Fenhe River Basin in China.

Due to the emergence of a large number of new open datasets, researchers have been increasingly using Copernicus DEM and comparing different datasets (Karlson et al., 2021; Garrote, 2022; Yuan et al., 2022). Mutar et al. (2021) considered the river basins flowing into the Mosul reservoir (Iraq), finding that the Copernicus DEM model was more accurate than SRTM DEM and ASTER GDEM. However, other authors (Kramm and Hoffmeister, 2021) reported that Copernicus DEM data can produce ambiguous results. Many studies (Karionov, 2010; Yeritsian, 2013; Trofimov and Filippova, 2014) have described the accuracy of the SRTM dataset. Nevertheless, other authors (Karionov, 2010; Yeritsian, 2013) have emphasized that the accuracy of the cartographic material obtained using the SRTM datasets is equivalent or close to that of topographic map. In recent years, ALOS World 3D DEM (Tadono et al., 2016) was also introduced and its accuracy, as well as advantages and disadvantages compared with other DEMs (Courty et al., 2019; Viel et al., 2020), have been analyzed. For the Crimean Peninsula, SRTM are mainly used by researchers, but comparisons with other DEMs are not given and measurement errors are not evaluated.

Several global datasets contain information on river basins on a global scale (Tang and Lettenmaier, 2012; Lehner and Grill G., 2013; Dallaire et al., 2019). For example, the HydroBASINS Version 1.0 dataset contains information on river basins and sub-basins worldwide. This dataset is not suitable for the Crimean Peninsula due to a large number of errors, in particular, incorrect allocation of the catchment basins' boundaries (for example, merged boundaries of the South Coast of the Crimean Peninsula river basins, unreasonable basin division of the largest rivers of the Crimean Peninsula into logically unreasonable parts, etc.). Most regional models in Europe do not include the Crimean Peninsula in the research area, which complicates further analysis and comparison of catchment basins (Vanham and Bidoglio, 2014). Additionally, the identification of small river basins or sub-basins for most large rivers of the Crimean Peninsula is limited only to the main tributaries, whereas the catchment basins of tributaries and their tributaries are practically not considered, with the exception of the most studied and largest river of the Crimean Peninsula, the Salgir River, as well as a small number of Crimean rivers.

Three groups can be distinguished among the studies on the catchment basins of the Crimean Peninsula. The first small group consists of studies (Dunaieva and Kovalenko, 2013; Narozhnyaya, 2021) that considered the river basins of the Crimean Peninsula as a whole. Almost always (with the exception of one study (Pozachenyuk, 2009), which is of historical value), river basins have been automatically identified using geographic information systems and DEMs. The second group consists of studies devoted to river basin groups in certain regions of the Crimean Peninsula: the river basins of the northwestern slope of the Crimean Mountains (Vermaat et al., 2012; Tabunshchik, 2018), the Kerch Peninsula (Krivoguz, 2016), the Sivash region (Timchenko et al., 2020), and the northern macroslope of the Crimean Mountains (Timchenko, 2000). The third and most numerous groups of studies has focused on the catchment basins of different separate rivers of the Crimean Peninsula (Vlasova, 2011; Pozachenyuk et al., 2014; Ergina and Timchenko, 2016; Kayukova, 2016; Amelichev et al., 2017). Moreover, in the third group, the most studied river basins are the basins of the largest rivers and their main tributaries.

The purpose of the study is to select the most accurate DEMs and, on its basis, to identify the basins and sub-basins of the rivers within northwestern slope of Crimean Mountains. Specifically, the main contents of this study are as follows. In the Section 2 "Materials and Methods," four DEMs are compared with a topographic map and a general scheme of research using geoinformation research methods are presented. In the Section 3 "Results," the calculation of measurement errors typical for various DEMs are shown. Also, the result of modeling the allocation of basins and sub-basins of the rivers of the northwestern slope of the Crimean Mountains are presented. In the Section 4, discussion of the obtained results and their comparison with other regions of the world are shown. Also, in the Section 4, the difficulties that the authors encountered while working on the article and ways to solve them are described. In the Section 5, conclusions and implications are given.

2 Materials and methods

2.1 Study area

The basins of the Zapadnyy Bulganak, Alma, Kacha, Belbek, Chernaya Rivers are located in the southwestern part of the Crimean Peninsula (Figure 1). The area of the studied territory comprises approximately 2,299 sq km. The rivers originate in the Crimean Mountains and flow into the Black Sea (Tabunshchik et al., 2022).

2.2 Materials and methods

We chose the DEMs for identifying river basins by selecting those most suitable for the study geodatasets, with the highest spatial resolution, and distributed under an open license. We thus selected





ASTER GDEM (Version 3, 2019), ALOS World 3D (Version 3.2, 2021), Copernicus DEM (Version 3, 2021), and SRTM DEM (Version 3, 2013) for this study among all DEMs available to us. The spatial resolution of these DEMs is 30 m/pixel.

As a DEM differs from the real terrain elevation, as for each pixel, an average value is given, we verified the selected DEMs by comparing the height marks and elevation values obtained from the topographic map. From a topographic map with a scale of 1: 100,000 [which was previously linked to the WGS 84 UTM zone 36 N (EPSG: 32,636) projection], we obtained sample values of several peaks (mountains, points) and isohypses, which we then compared with the elevation values of the same points on each of the considered DEMs through a simple spatial relationship. For these purposes, a point shapefile was created containing the elevation

values of points from a topographic map. Then, using the tools "Spatial Join" and "Extract Values to Points," the elevation values from each DTM for each point were obtained. In total, we selected 100 points for each catchment basin, and then we compared the obtained data. Accuracy was calculated according to a previously reported method (Onkov, 2011). An additive error model was adopted during statistical data processing. According to this additive error model, we calculated the difference in heights of the DEM H_{DEM} and the topographic relief H_{TOPO} as

$$\Delta H = H_{DEM} - H_{TOPO} \tag{1}$$

Land was considered as the sum of systematic Δ_H and random Δh errors:

$$\Delta H = \Delta_H + \Delta h. \tag{2}$$

TABLE 1 Comparison of accuracy of absolute heights in study area according to a topographic map at scale of 1:100,000 for various DEM sets.

Error type	DEM				
	ASTER	ALOS	Copernicus	SRTM	
Systematic error, m	-7.7	-3.8	-3.4	-9.5	
Root mean square error, m	14.0	12.8	8.7	14.9	
Mean absolute error, m	10.3	7.4	6.0	10.9	
Standard root mean square error, m	21.9	19.0	14.3	11.3	
Centered mean absolute error, m	3.9	3.9	3.4	2.7	

TABLE 2 Comparison of accuracy of absolute heights in Zapadnyy Bulganak River basin according to topographic map at scale of 1:100,000 for various DEM sets.

Error type	DEM				
	ASTER	ALOS	Copernicus	SRTM	
Systematic error, m	-3.8	-2.7	-1.5	-1.9	
Root mean square error, m	6.7	4.4	3.6	3.7	
Mean absolute error, m	4.7	2.8	2.2	2.4	
Standard root mean square error, m	14.7	12.8	9.6	4.8	
Centered mean absolute error, m	3.0	3.2	2.7	1.8	

TABLE 3 Comparison of accuracy of absolute heights in Alma River basin according to topographic map at scale of 1:100,000 for various DEM sets.

Error type	DEM			
	ASTER	ALOS	Copernicus	SRTM
Systematic error, m	-10.5	-7.0	-5.2	-11.4
Root mean square error, m	13.3	12.1	10.0	17.3
Mean absolute error, m	11.0	8.0	6.7	12.1
Standard root mean square error, m	23.3	20.8	16.5	13.6
Centered mean absolute error, m	4.2	4.3	3.8	2.9

TABLE 4 Comparison of accuracy of absolute heights in Kacha River basin according to topographic map at scale of 1:100,000 for various DEM sets.

Error type	DEM			
	ASTER	ALOS	Copernicus	SRTM
Systematic error, m	-12.6	-9.1	-7.1	-16.1
Root mean square error, m	19.6	13.9	12.0	19.6
Mean absolute error, m	14.8	10.9	9.5	16.7
Standard root mean square error, m	28.2	22.9	18.7	14.2
Centered mean absolute error, m	4.5	4.5	4.0	3.3

Error type			DEM	
	ASTER	ALOS	Copernicus	SRTM
Systematic error. m	-6.2	-0.2	-3.4	-10.4
Root mean square error. m	15.3	20.5	8.8	16.3
Mean absolute error. m	11.7	10.2	6.8	13.4
Standard root mean square error. m	22.4	23.6	14.7	12.9
Centered mean absolute error. m	4.1	4.1	3.5	3.0

After eliminating the systematic error from the measurement results using

$$\Delta h_i = \Delta H_i - \bar{\Delta}_{H\,i} \tag{3}$$

the parameters of the random component Δh were estimated.

The following types of errors were calculated in the study (where n is the number of measurements):

1. Average elevation difference (systematic error), m;

$$\bar{\Delta}_H = \frac{1}{n} \sum_{i=1}^n \Delta H_i \tag{4}$$

2. Root mean square error, m;

$$RMSE_{\Delta H} = \sqrt{\frac{1}{n} \sum_{i=1}^{n} \Delta H_i^2}$$
(5)

3. Mean absolute error, м;

$$MAE_{\Delta H} = \frac{1}{n} \sum_{i=1}^{n} |\Delta H_i|$$
(6)

4. Standard root mean square error (Bessel's correction), m;

$$\sigma_{\Delta h} = \sqrt{\frac{1}{n-1} \sum_{i=1}^{n} \Delta h_i^2}$$
(7)

5. Centered mean absolute error, m.

$$\Theta_{\Delta h} = \sqrt{\frac{1}{n-1} \sum_{i=1}^{n} |\Delta h_i|}$$
(8)

The methodology for delineating river basins is based on the utilization of the ArcGIS software suite in conjunction with the DEM. It encompasses a systematic algorithm comprising a series of steps executed using the "Hydrology" toolbox within the "Spatial Analyst" tool.

- 1. The DEM is imported into the ArcGIS software suite.
- 2. The "Fill" tool from the "Hydrology" toolbox in the "Spatial Analyst" toolset is employed to rectify erroneous depressions within the DEM.
- 3. The "Flow Direction" tool from the "Hydrology" toolbox in the "Spatial Analyst" toolset is applied to ascertain the flow direction for each pixel of the DEM, which has been preprocessed (in step 2) using the "Fill" tool.
- 4. The "Flow Accumulation" tool from the "Hydrology" toolbox in the "Spatial Analyst" toolset is utilized to compute the cumulative flow, representing the aggregated weight of all pixels that drain into each downslope pixel in the resulting raster. The flow direction raster created in step 3 serves as the input.
- 5. The "Raster Calculator" tool from the "Map Algebra" toolbox in the "Spatial Analyst" toolset is used to select pixels with a flow accumulation value exceeding 25. As a result, a new raster is generated with flow accumulation values above 25.
- 6. The "Stream Link" tool from the "Hydrology" toolbox in the "Spatial Analyst" toolset is employed to create a raster linear network wherein each section of the network is assigned unique values, representing individual stream links. The input rasters consist of the flow direction raster generated in step 3 and the

Error type	DEM				
	ASTER	ALOS	Copernicus	SRTM	
Systematic error, m	-6.1	-1.6	-0.7	-9.2	
Root mean square error, m	12.4	7.2	6.8	12.2	
Mean absolute error, m	10.5	5.7	5.2	10.5	
Standard root mean square error, m	20.6	14.7	11.8	8.2	
Centered mean absolute error, m	3.9	3.7	3.3	2.6	

TABLE 6 Comparison of accuracy of absolute heights in Chernaya River basin according to topographic map at scale of 1:100,000 for various DEM sets.



FIGURE 3

Stream sub-basins on the northwestern slope of the Crimean Mountains within the basins of the Zapadnyy Bulganak, Alma, Kacha, Belbek, and Chernaya Rivers.

flow accumulation raster with values exceeding 25, produced in step 5.

- 7. The "Stream Order" tool from the "Hydrology" toolbox in the "Spatial Analyst" toolset is utilized to assign a stream order to each link within the stream network raster created in step 6.
- The "Basin" tool from the "Hydrology" toolbox in the "Spatial Analyst" toolset is employed to generate a raster depicting river basins based on the constructed flow direction raster (step 3).
- 9. The "Raster to Polygon" tool from the "Conversion" toolbox within the "From Raster" toolset is applied to transform the stream network raster obtained in step 7 into a polygon shapefile.

We supplemented and implemented the method for identifying river basins and their sub-basins using ArcGIS 10.8 software and the builtin model editor "Model Builder," which allowed us to automate and speed up the delineation process (Figure 2). Our identification of river basins is based on a previously described method (Elkhrachy, 2018; Bajirao et al., 2019; Garrote, 2022; Samsonov, 2022). The theoretical and methodological foundations of delineating river basins and sub-basins were extensively discussed by Bai et al. (2015a) (Bai et al., 2015b). We automated it using the built-in ArcGIS Model Builder (Figure 2).

We note the sensitivity of this method to the incoming sets of spatial data: the type of DEM and its accuracy, as well as the accuracy of tying the points of the river mouths.

3 Results

We calculated the values of five different types of errors for the territory of the river basins of the northwestern slope of the Crimean Mountains. These values are presented in Table 1. Table 1 shows that the errors for ASTER GDEM and SRTM DEM were the largest; those of ALOS World 3D and Copernicus DEM were the smallest. Additionally, the values of errors within the basins of the Zapadnyy Bulganak, Alma, Kacha, Belbek, and Chernaya Rivers were analyzed (Table 2, Table 3, Table 4, Table 5, Table 6).

The Copernicus DEM is most suitable for the analysis of the morphometric characteristics of the river basins of the northwestern slope of the Crimean Mountains. This DEM has a spatial resolution of 30 m/pixel.

Based on the Copernicus DEM, the boundaries of the streams sub-basins within the basins of the Zapadnyy Bulganak, Alma, Kacha, Belbek, and Chernaya Rivers were identified as a result of modeling and partial manual correction of the obtained model results (we selectively checked for the presence of errors in the boundaries of the selected sub-basins). The obtained results are shown in Figure 3.

Figure 3 shows that we identified 3,293 sub-basins in the study area, which form the valley, gully, and ravine systems of the study area. The basin of the Western Bulganak River has 207 sub-basins, the Alma River has 860 sub-basins, the Kacha River has 855 sub-basins, the Belbek River has 747 sub-basins, and the Chernaya River has 624 sub-basins.

4 Discussion

After analyzing the data presented in Table 1, Table 2, Table 3, Table 4, Table 5, Table 6, we found that the error values of the

Basin	Area, km ²			Difference from literature data, %	
	SRTM DEM	Copernicus DEM	According to (Lisovsky et al., 2011)	SRTM DEM	Copernicus DEM
Zapadnyy Bulganak	177.1	174.6	180	2	3
Alma	641.8	631.8	635	-1	0
Kacha	570.9	573.3	573	0	0
Belbek	492.1	491.1	505	3	3
Chernaya	430.5	428.0	427	-1	0

TABLE 7 Comparison of basin areas of the Zapadnyy Bulganak, Alma, Kacha, Belbek, and Chernaya Rivers.

Copernicus DEM, which has a spatial resolution of 30 m/pixel, were the lowest. Using the same technique with different data as the input to the model may have led to the obtained results slightly differing. This may concern both the boundaries of the study region and the software products on which the data were processed. For example, if we compare our earlier calculations of the area of catchment basins using SRTM DEM (Tabunshchik, 2021a) with those obtained using Copernicus DEM and data from the literature, insignificant differences are observed. Here, on the entire-basin scale, these changes are insignificant, and the differences are mainly related to the number of points along which the outer boundaries of the river basins are drawn.

We compared our data with those calculated by researchers for other regions. Karwel and Ewiak (2008) reported that the accuracy of SRTM within the flat part of the territory of Poland is 2.9 m, and 5.4 m for mountainous and foothill areas. Calculations (Orlyankin and Aleshina, 2019) showed that within the river basins of the northwestern slope of the Crimean Mountains, the systematic error of elevation calculated from the SRTM dataset, with a spatial resolution of 90 × 90 m, is +1 m.

Mutar et al. (2021) indicated that the RMSE of Copernicus DEM is 1.3 m in Iraq, which is 2.6 times more accurate than the SRTM DEM dataset and 5.2 times more accurate than the ASTER GDEM dataset. The accuracy of the Copernicus DEM dataset in China is 6.73 m (Li et al., 2022). Santillan and Makinano-Santillan (2016) found that when comparing datasets within the Philippines, the AW3D30 dataset most accurately represents true heights compared with the SRTM and ASTER GDEM datasets, because the AW3D30 dataset has the lowest mean error, RMSE, and standard deviation. Elkhrachy (2018), for the territory of Saudi Arabia, reported that when comparing DEM and a topographic map at a scale of 1: 10,000, which was chosen as a reference, the vertical accuracy of the SRTM and ASTER datasets is ± 6.87 and ± 7.97 m, respectively. Dong et al. (2015) conducted an accuracy assessment of ZY-3, SRTM, DLR-SRTM, and GDEM in Northeast China. GPS data was used as the accuracy evaluation criterion for ZY-3, and the RMSE for SRTM was found to be ±2.82 m. Zhang et al. (2019) compared ASTER, SRTM, ALOS, and TanDEM-X for flood risk mapping on the island of Hispaniola, using GPS and LiDAR measurements. They found that ASTER had the highest errors, while ALOS and TanDEM-X had the lowest errors. Karabörk et al. (2021) compared AlosPalsar, Sentinel-1A, AW3D30, SRTM, and ASTER GDEM with ground control points (GCP) obtained from digital aerial photographs, photogrammetric maps, or orthophotos. They found that the mean error values for ALOS were 1.1 m on flat terrain and 8.2 m in mountainous areas, while SRTM had mean errors of 1.8 m on flat terrain and 7.9 m in mountainous areas. ASTER had mean errors of 1.0 m on flat terrain and 8.4 m in mountainous areas. Purinton and Bookhagen (2021) compared the accuracy of SRTM, ASTER, ALOS, TanDEM-X, and Copernicus DEM in the Arid Central Andes. They found that the Copernicus DEM provided the most accurate representation of the landscape and should be the preferred DEM model for topographic analysis in areas where local high-quality DEM coverage is not available.

To demonstrate the changes in the areas and morphometric characteristics of river basins, Table 7 presents the results of a comparison of the basins area of the Zapadnyy Bulganak, Alma, Kacha, Belbek, and Chernaya rivers, calculated using the SRTM DEM and Copernicus DEM datasets, as well as a comparison with the data on river basin area given in the literature.

Using GIS, both the river basins (Ermolaev et al., 2014; Ali et al., 2023; Sharma et al., 2023) and sub-basins (Vanham and Bidoglio, 2014; Dallaire et al., 2019) of large rivers can be identified. However, the low accuracy of the DEM and new techniques can often lead to distortion of the output results. As an example, consider a previously described technique (Tabunshchik, 2021b; Samsonov, 2022), which is based on the PCRaster Python Library and automated by Van der Kwast as a PCRaster Tools plugin for QGIS. The application of this technique to the river basins of the northwestern slope of the Crimean Mountains showed a rather mixed picture that defies logical classification (Figure 4).

Figure 4 shows that the sub-basins of the rivers were not identified. The resulting processing result contained many errors, and the boundaries of sub-basins were identified without considering watersheds, which indicates the impossibility of applying this method in the study area.

The method proposed by Samsonov (2022) produced the best result of modeling the sub-basins of the five largest rivers of the northwestern slope of the Crimean Mountains (Zapadnyy Bulganak, Alma, Kacha, Belbek, and Chernaya Rivers). We propose using a buffer value that exceeds twice the DEM resolution's value to separate the points of the mouth of smaller streams flowing into the main stream. In our measurements, we found that this value should be equal to the pixel resolution, given the large error in constructing and visualizing data with a buffer size of two pixels (Figure 5).





In some areas, we identified single image pixels that did not belong to catchment basins, so we then manually identified them. Additionally, the resolution of various open DEM datasets impacts the accuracy of sub-basin identification. When comparing DEM data with a resolution of 90 m/pixel with those with a resolution of 30 m/pixel, the most accurate results were achieved by the latter.

Many studies have focused on the identification of river basins using SRTM DEM for the Crimean Peninsula (Vlasova, 2012; Pozachenyuk et al., 2014; Pozachenyuk et al., 2015; Tabunshchik, 2021a; Narozhnyaya, 2021; Drygval, 2022). Other DEMs have practically not been used, which is probably due to the popularity and widespread use of SRTM DEM.

Narozhnyaya (2021) provided a detailed description of the morphometric analysis of the river basins of the Crimean Peninsula; however, the description is based on the use of SRTM DEM. A detailed description of the individual basins of the large rivers of the Crimean Peninsula was not provided, including the five basins that we considered. Narozhnyaya (2021) did not distinguish the sub-basins of the rivers: only separate maps were presented that allowed judging the distribution of certain morphometric indicators of river basins.

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We identified the sub-basins within the basins of the Zapadnyy Bulganak, Alma, Kacha, Belbek, and Chernaya Rivers for the first time, and we substantially refined the available information on the quantitative characteristics of the basins of these rivers. This, in turn, creates many opportunities for researchers of the nature of the Crimean Peninsula to solve many problems with more accurate initial data. Additionally, the use of DEMs is advisable when conducting complex engineering, geological, hydrological, and hydrogeological studies. Several directions for further research of the river basins and sub-basins of the northwestern slope of the Crimean Mountains emerge from our study, in particular, their anthropogenic transformation, searching for relationships between climate change and changes in vegetation and land cover types; assessing their geoecological state; preparing landscape planning maps; and developing and implementing recommendations for sustainable development.

In further research, it is necessary to continue comparing different DEM datasets and identifying the most accurate and suitable ones for specific tasks. While the study focused on analyzing specific DEM datasets for the studied region, there is potential in future research to examine the accuracy of various DEM datasets in different regions and compare them to each other. Additionally, advancements in remote sensing technologies and data processing methods present opportunities to improve the accuracy of DEM datasets. Further research can explore the integration of data from multiple sources, such as LiDAR, satellite imagery, and ground-based measurements, to enhance the accuracy and reliability of DEMs. The use of Unmanned Aerial Vehicles (UAVs) at the local level of investigation also holds great interest.

5 Conclusion

Despite more than two centuries of hydrological studies of the rivers and river basins of the Crimean Peninsula, many unsolved problems remain. In general, the Crimean Peninsula remains insufficiently hydrologically studied. This primarily applies to important tasks such as the hydrological characteristics of rivers and the morphometric characteristics of river basins, determining the types of water management use of rivers and developing schemes for optimizing river and river basin use, studying and predicting possible ecogeodynamic processes under the influence of anthropogenic factors within river basins, studying the degree of anthropogenic transformation of the river basin, etc. Only at the end of the 20th to the beginning of the 21st century did detailed work begin on the identification and description of the basins and subbasins of the rivers of the Crimean Peninsula, which continues to this day. However, these studies are still extremely scarce, and the data are scattered. This study of the river basins and sub-basins identifying the northwestern slope of the Crimean Mountains provides a distinctive contribution to the unresolved history of hydrological research within the Crimean Peninsula.

The use of DEMs enables the study of the main morphometric characteristics of the river basins of the northwestern slope of the Crimean Mountains. However, when choosing the initial data, the least error-prone datasets should be used. The performed calculations showed that the smallest errors in the selection of DEM were obtained for Copernicus DEM, which has a resolution of 30 m/pixel. Copernicus DEM provides a sufficiently high level of accuracy and detail, which was shown in the calculation model of the Zapadnyy Bulganak, Alma, Kacha, Belbek, and Chernaya River basins, as well as their sub-basins.

Data availability statement

The original contributions presented in the study are included in the article/supplementary material, further inquiries can be directed to the corresponding author.

Author contributions

VT conceived and designed the experiments; TG performed the experiments; VT and RG analyzed the data and contributed to the analysis tools; VT, TG, CP, AK, and RG wrote the paper, but all authors discussed the results and enhanced the final draft of the manuscript. All authors contributed to the article and approved the submitted version.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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