

Dynamics of Soil Salinity in Irrigation Areas in South Kazakhstan

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Abstract

The analysis has been conducted of possibility of determining the extent of soil salinity by indirect and direct decoding of satellite images Pleiades 1A / 1B (in the year of survey) and LANDSAT (archive). It was found that indirect decoding of salinity based on spectral characteristics of vegetation images is strongly dependent on crop growth phase (period of shoot-ing). LANDSAT archive images with normalized soil salinity index (NDSI) allow to develop salinity maps and soil salinity dynamics maps at semi-quantitative level. Based on computer analysis of LANDSAT images it was determined that soil salinity at study object during the pe-riod from 1987 to 2014 has increased due to significant decrease of the area of non-saline soil by 41.5% and increasing of the areas of low and moderately saline soils by 34.9%, also regions with heavily saline soils at 6.6% of total surveyed area have been detected, which previously were absent.

Keywords: regression analysis, soil salinity, geostatistical methods, information system, soil salinity dynamics map

Introduction

As a natural system, each terrain has important properties: self-organization, self-regulation, and self-renewal – all of which facilitate sustainable development of an ecosystem and its ability to resist degradation and crisis processes. Sustainability of terrain is associated with the ability of its components to conserve their structure and function under external influences [1].

At present, the impact of anthropogenic factors often exceeds maximal permissible limits, destroying natural ecosystems. First of all, land degradation results in expanded water and wind erosion, reduction of groundwater level, pollution, water logging and salinity of irrigated and non-irrigated soils [2]. Especially in arid regions where productivity of agricultural production is dependent on land irrigation, and soils are mostly subjected to salinity. The following six groups of degradation have been outlined in the World Atlas of Desertification [3]: water erosion, wind erosion, soil fertility decreases, salinization,

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waterlogging, and lowering of the groundwater level. Among them, salinization has been explored as the most relevant issue for Central Asia [4-5] and Kazakhstan [6-7].

Currently in the Republic of Kazakhstan, saline and alkaline soils occupy 111.6 million ha., or 41.0% of the nation [4]. The share of salt marshes in soil surface structure is significantly increasing in the southern part of the country, which represents a closed inland area that does not have free effluent to ocean basins. Extensive use of fertility of irrigated soils during the transition period (after the collapse of the Soviet Union, in particular), the poor state of irrigation and drainage networks, and non-compliance of their technical parameters with project standards have resulted in sharp deterioration of soil-reclamation conditions of irrigated areas. Here, 42,912 hectares of soil have unsatisfactory reclamation status due to salinity, 80,005 ha due to the rise of groundwater level, and 24,909 ha due to both factors [5]. Lack of a permanent operational monitoring system of irrigated soils has also contributed to this fact. Currently, traditional methods of exploring salinization dynamics by means of major salt surveying and soil salinity mapping are not used, mainly due to high costs. Therefore, currently one of the main problems in the Republic is that there is no reliable and timely data on soil-reclamation status of irrigated areas, as well as data on the development of secondary soil salinization and its impact on crop yields.

In this regard, currently there is an urgent need for the development of satellite techniques for soil salinity monitoring, which is considered one of the relevant new directions in research on dynamics of changes in saline soils. These methods are fundamentally different from their analogue land surface mode of exploring soil-salt regimes, and it is more operative, accurate, and comparatively cheaper. A review of recent scientific publications shows that the issue of analyzing saline soil dynamics in large areas has not been addressed [8-12]. A large number of publications has been devoted to the possibility of exploring soil salinity based on decoding satellite data, which is regarded as one of the main sources of information for the analysis of soil salinity dynamics [13-17]. Currently most researchers use Landsat TM and ETM + satellite data and can also use Landsat MSS data [18-21] for monitoring. Several researchers use Ikonos IRS-II LISS-II and III data. Works also use ASTER and MODIS data [22-25]. An overview of scientific publications shows that currently there is no global unified methodology for space monitoring of soil salinity, and all works in Kazakhstan still have an exploratory character. In this regard, the main objective of this work is to develop a method of determining soil salinity dynamics based on satellite data.

Research Area

The research object includes soil patterns of the Shilik and Aktobe rural districts in the Shoulder irrigation area in the Otyrar District of southern Kazakhstan (Fig. 1). The

figure shows that from east and south, the area is framed by a gently sloping plain of ridges of the West Tien Shan and Karatau. The Kyzyl Kum sandy area serves as a natural boundary in the west, and in the north and northwest it borders with the Turkestan irrigation area. Most of the area is used as pasture for grazing farm animals, and the southern and eastern parts along the valleys of the rivers Shayan and Bugun are plowed as irrigated arable land. Terrain of the area is represented by slightly wavy or horizontal surface with poor and monotonous flora. Here dominate various species of wormwood (*Artemisia*), salt plants (*Salsola*), and jantak (*Alhagipseudalhagi*). The river valleys are rich in meadow grasses, thickets of wild rose (*Rosa*), and groves of poplar (*Populus*) and elm (*Ulmusfoliaceae*). Saline ajrek (*Aeluropuslittoralis*) are near floodplains [26].

The climate is sharply continental and desert. Temperature in the winter falls to -25°C . The Syrdarya River usually freezes in early December and ice stays until March. Often in spring the Syrdarya and Aris overflow their banks, flooding a large area [27].

The leading crops in these rural districts are fodder crops: corn for grain, alfalfa, and rarely cereals, vegetables, and melons. The Bugun River is the main source for irrigation water, and on this river the reservoir with the same name is constructed. The water supply network is represented by different open-type sprinklers that are laid in natural ground and, of course, serve as an additional source of groundwater supply.

The area is dominated by meadow-gray soil saline (Solonchak, sometimes saline) soils (Xerosols) occupying mid-level surfaces that are formed on saline weakly layer loamy and clay sediments in medium depth (4-6 m) saline groundwater under the sparse grass-halophytic shrubs with ephemerals and wormwood

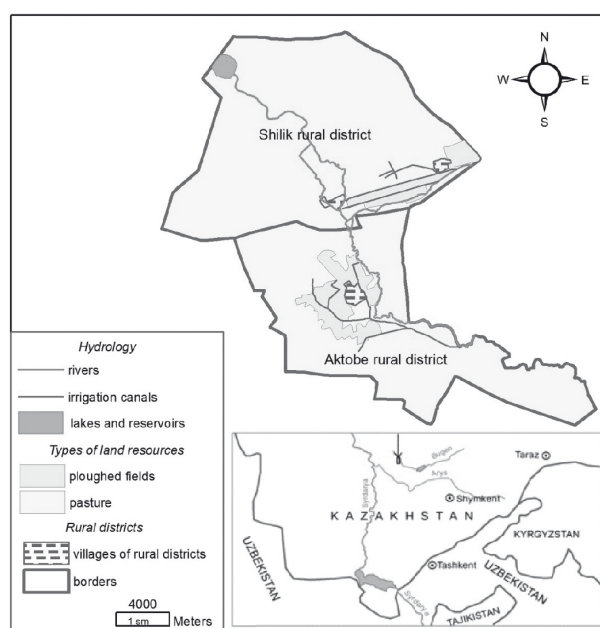


Fig. 1. Research area.

(*Artemisia*). Also here are meadow-gray SolonetzGeyic soils, occupying average level surfaces and micro-terrain depressions, under-halophytic wormwood, wormwood-halophytic, and halophytic vegetation with ephemeral plants, as well as Solonchaks-Gleyic, located on micro-terrain rises to 20-30 (50) cm of relative height under sparse halophyte plants – mostly leafless barnyardgrass (*Echinochloa*).

Solonchaks and Solonetz are formed here, as a rule, on heavier and saline deposits under conditions of high mineralization of average-depth groundwater. In terrain depressions with close (up to 3 m) groundwater, the following soils are formed: meadow-swamp saline soils under meadow-marsh vegetation at very close (1.5 m) slightly mineralized waters, gray-meadow saline salt marshes under halophytic and grass-halophytic vegetation on close slightly mineralized waters, meadow marshes under halophytic grass-halophytic vegetation at close (1.5-3 m) slightly mineralized waters, and ordinary salt marshes under halophytic vegetation at close heavy mineralized groundwater.

Salt marshes laying in complexes and combinations usually occur at elevated areas of micro- and mesoterrain [28] relative to other soils. The prevailing salinity type is chloride-sulphate and sulphate-chloride, sometimes with the presence of normal soda. All soils of the area are carbonate and are characterized by high alkalinity (pH 8-9). Water-physical, physical, and physical-chemical properties of soils depend on the degree of salinization and alkalization.

According to conditions of groundwater supply and outflow, the area refers to hydro-geological region of intense external inflow and hindered outflow of groundwater and due to this fact, soils in this area are subject to secondary salinization. Former farm canals, collectors, and vertical drainage wells have been neglected, and their parameters do not meet the project standards and also result in raising the groundwater level and thus result in secondary salinization. We also know that in terms of irrigation, soil formation processes are sufficiently intense and have a relatively high rate of mobilization and migration processes. Therefore, monitoring fertility levels, especially salinization of irrigated soils, must be carried out regularly and include a wider range of soil properties to be determined.

Materials and Methods

Our work was carried out by means of space and ground soil surveys of the research area that were synchronous in time and location. For this purpose, by means of conducting ground reconnaissance survey of the area, we have chosen subsatellite areas with contrasting soil salinization, and satellite images have been ordered from Pleiades 1A/1B in panchromatic mode with spatial resolution of 0.5 m on the site, and in multispectral mode (4-channel mode) with spatial time-resolution about 2.0 m on the site. In addition, LANDSAT satellite data

collection for previous years was done from open sources (USGS) aimed at exploring the long-term dynamics of soil salinization.

Corn for grain and alfalfa are the major crops on the research object, as well as in the whole Shoulder irrigation area. Based on biological features of these crops and the length of their vegetation period data at the Institute of Soil Science, the last week of July was selected as the period for space and ground surveys. It was assumed that at this time corn will be in the buttonhole phase, and alfalfa at the beginning of the flowering phase (second mowing), i.e., at this time it is possible to measure vegetation index using ground and space-based techniques.

In regard to satellite data processing methods and decoding of soil salinity state, we can outline the following main approaches:

- Calculating vegetation indexes that facilitate identification of soil salinity.
- Using statistical models and methods (multiple regression, method of principal components, maximum probability method, regression of partial least squares).
- Using geostatistical techniques (kriging, co-kriging, modified kriging).

A review of literature revealed that such methods as partial least squares regression (among regression methods), principal component analysis, co-kriging, and modified kriging provide reliable results and high accuracy. It should be noted that when using the regression approach, it is also recommended to use the probability ratio test, the Lagrange multiplier, and corrected information criterion Akaike [29-32]. Note that decoding saline soils can be done using the character of exposed soil surface as well as by analyzing vegetation. In the latter case, the applied methods of decoding may be the same, but during decoding of soil salinity based on the status of vegetation, it is necessary to include the unit of identification of crop types cultivated in a particular growing season into the satellite monitoring system. In addition, we should pay attention to the fact that hyperspectral and radar satellite data can also be used for researching saline soils [33]. However, methods for their use are insufficiently developed.

Regression analysis of the correlation between magnitude of electrical conductivity measured in the field, and space images, was used for decoding satellite images. It was performed in the program Statistica v10.0 media for each soil layer (0-20 cm, 20-50 cm, 50-100 cm) and separately for alfalfa and corn. The sampling for cornfields was 25 points, and 30 points for fields with alfalfa. The following predictors were used for regression analysis:

1. Spectral channels of satellite Pleiades 1A / 1B:
 - a) Blue (Channel1), Green (Channel 2), Red (Channel 3), and Near Infrared (Channel 4).
2. Correlation channels of Pleiades 1A / 1B:
 - a) $B / G = \text{Channel 1} / \text{Channel 2}$, $R / G = \text{Channel 3} / \text{Channel 2}$, $B / \text{NIR} = \text{Channel 1} / \text{Channel 4}$, $B / R = \text{Channel 1} / \text{Channel 3}$, $\text{NIR} / G = \text{Channel 4} / \text{Channel 2}$.
3. Vegetation index, describing the state of vegetation [34]:

- a) $NDVI = \frac{(\text{Channel4}-\text{Channel3})}{(\text{Channel4}+\text{Channel3})}$, $IR_R = \frac{\text{Channel4}-\text{Channel3}}{\text{Channel4}+\text{Channel3}}$, $SQRT = \sqrt{\frac{\text{Channel4}-\text{Channel3}}{\text{Channel4}+\text{Channel3}}}$.
- b) $VEGI = \frac{\text{Channel4}-\text{Channel3}}{(\frac{\text{Channel4}-\text{Channel3}}{\text{Channel4}+\text{Channel3}})+0.5}$, $TND\ VII = \frac{\text{Channel4}-\text{Channel3}}{(\frac{\text{Channel4}-\text{Channel3}}{\text{Channel4}+\text{Channel3}})+0.5}$.
- c) $GNDVI = \frac{(\text{Channel4}-\text{Channel2})}{(\text{Channel4}+\text{Channel2})}$, $NDGR = \frac{(\text{Channel2}-\text{Channel3})}{(\text{Channel2}+\text{Channel3})}$.

Ground-based research was conducted according to the “All-union instructions ...” and “Guide on conducting ...” [35-36]. In addition, during the fieldwork we used the latest equipment for the study of soil salinity and global positioning systems. For conducting salt shooting, along with the traditional method (layer survey, well drilling) we also used field portable Progress 1T salt-meters [37]. To clarify the shapes of saline soils based on satellite imagery we used a Garmin GPS 18 paired with an ASUS netbook, and to determine layer point coordinates we used a Garmin 62 GPS unit.

Results and Discussion

Our work began with the creation of the basic GIS layers within the Shilik and Aktobe rural districts (research objects), including borders of irrigated sites, settlements, roads, lakes, rivers, irrigation network, etc. All the layers that make up the GIS were obtained by decoding Pleiades space imagery for 2014, and also land use maps of local farmers. After conducting fieldwork, field observation points and results of satellite imagery of the research object were added into GIS.

Also, we conducted an analysis of availability of archived satellite data on the research territory. It was established that from 1985 to 2014 in the archive there is a sufficient number (46) of LANDSAT images of high spatial resolution. These images were later used as the basis for identifying soil salinity dynamics in the research region.

As a result of regression analysis for the Pleiades satellite image 1A/1B of 15 July 2014, the statistically significant regression with determination coefficient (R^2) of 0.51 was determined only for soil salinity in the 20-50 cm layer for plots with corn.

The above-determined pattern to some extent can be explained by the fact that in the Shoulder irrigation area, irrigation of corn begins in the phase of 7-8 leaves at the end of June and finishes in the second half of August. During this time, 4-5 irrigations with a norm of 600-700 m³/ha were undertaken. These irrigations promote desalination of upper topsoil that can strongly reduce linkage between soil salinity and spectral properties of satellite imagery.

Multiple regressions have also been developed for Landsat satellite imagery for day 284 in October 2014, but they also have very low values of determination coefficients (R^2) of about 0.3. The only significant regression was built for one 50-100 cm layer of soil under corn. The best predictors were 5 and 4 channels of the image.

As it was not possible to build sufficiently reliable re-

gression models using the images for mapping salinity, we used a different approach using image classification and determination of salinity index NDSI. Since in most cases fields in the research area are abandoned due to high soil salinity, first of all masks of abandoned fields were built. For this purpose a pair of Landsat satellite images acquired during recent years have been used to exclude temporarily unused fields. In the images of 2013 and 2014 for days 280-285 (October), the abandoned fields have been automatically decoded using ILWIS 3.3.1, with the help of supervised classification on training sampling by the “greatest similarity” method. Classification was carried out in a combination of channels (4-3-2). The imagery acquired in 1988 shows that all agricultural plots were used. Thus the abandoned fields have outlined which soils were considered the most saline.

Less saline soils have been decoded based Landsat 5TM on satellite images during the second half of autumn, when images show the open soil surface: day 312 in 1988 (October) and day 332 in 2014 (November). Normalized index of salinity (NDSI) was calculated for each image. A variation of the indexes of channels 5 and 4 of satellite Landsat was used, as these channels are the most frequent as significant predictors in construction of salinity regression models: $NDSI = \frac{(\text{Channel5} - \text{Channel4})}{(\text{Channel5} + \text{Channel4})}$, in which Channel4, Channel5 – respective channels of Landsat satellite.

Then the image of NDSI indices were divided into three classes corresponding to non-saline, slightly saline, and medium-saline soils. Quantile counted on imagery for day 332 in 2014 were used as class boundaries. Quantile values are $p(0.33) = 0.26$ and $p(0.66) = 0.28$. Furthermore,

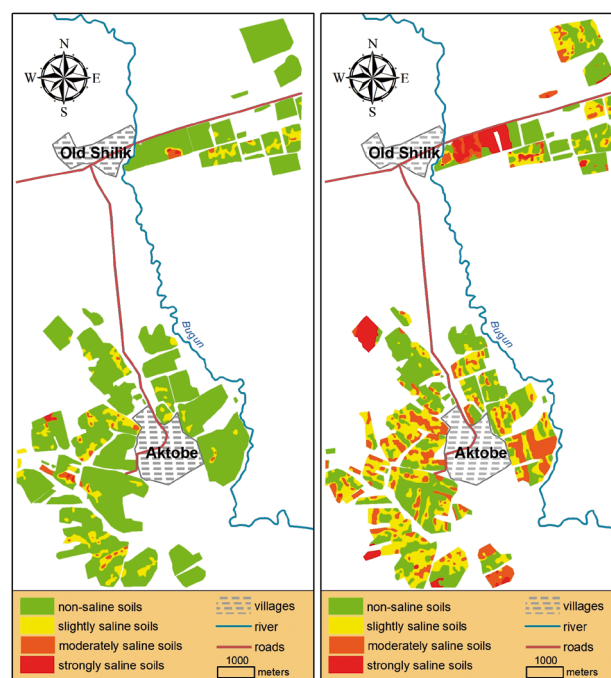


Fig. 2. Salinity maps (left - in 1988, right - 2014, Signs: 1 - non-saline, 2 - slightly saline, 3 - moderately, 4 - strongly saline) as a result of LANDSAT images decoding.

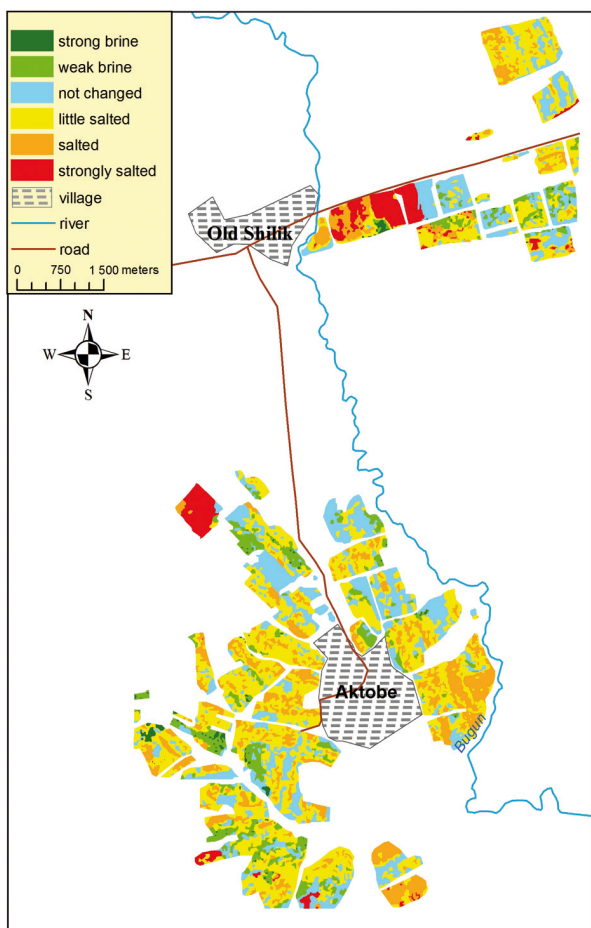


Fig. 3. Map of soil salinization the dynamics of research object for the 1988-2014.

we combined masks of heavily saline soils (unused fields) and non-saline masks, slightly saline, and medium-saline soils final salinity maps for 1988 and 2014 (Fig. 2).

Also, a map of salinity dynamics was built to assess the dynamics of salinity (Fig. 3). The rate of salinity change was marked in green for desalination process; gray for unchanged salinity; and yellow, orange, or red for increased levels of salinity. So, on the map the most

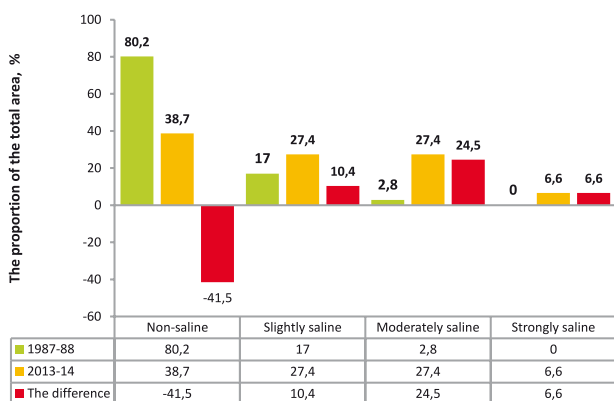


Fig. 4. The dynamics of soil salinization in the period from 1987 to 2014.

contrast comes from the images of soils where salinity has significantly increased or decreased.

Based on the map of soil salinity dynamics of research object (2262.9 ha), soil contour areas with varying degrees of salinity for 1987-88 and 2013-14 were calculated in GIS media, and the degree of change was determined (Fig. 4). As shown in Figure 4, for the past 26 years non-saline soil areas have significantly changed, reducing the area by 41.5% or 939.4 ha.

Due to reduction of the area of non-saline soils, the areas of medium and strongly saline soils have increased, respectively, by 10.4 and 224.5% or 235.3 and 555.2 hectares. There appeared contours of strongly saline soils, which occupy 6.6% of the surveyed area. That is a result of studying the dynamics of soil salinity using satellite images, and it has been determined that over the last 26 years in Shilik and Aktobe rural districts progressive salinization is taking place. Highly saline soils cover mainly the peripheral parts of the area, and the end portions of the main irrigation canals, and contours of moderate soils are found everywhere.

Conclusion

Using regression analysis of linkage between spectral properties of Pleiades space image 1A/1B and LANDSAT archival images and electrical conductivity of irrigated soils, it became possible to build salinization regression models only for particular soil layers under corn, and they are not reliable enough statistically, apparently due to the inappropriate choice of the image period.

Probably indirect decoding of salinity on spectral characteristics of vegetation images would enable us to get more qualitative regression models. In this case, we should take into account crop vegetation phases when selecting images.

Salinity maps and a map of salinity dynamics were built on a semi-quantitative level using automated image classification on learning selection using “maximum similarity” and dividing soil imagery based on NDSI salinity index values. Strongly saline soils were identified most accurately, while soils with other salinity degrees were outlined with less precision. However, analysis conducted using LANDSAT archive images showed that soil salinity in the research object for the period from 1987 to 2014 has increased significantly. This was due to a significant decrease in the area of non-saline soils and an increase in the area of low and moderately saline soils. Also, previously absent contours of strongly saline soils have appeared.

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