Animals were sampled within the experimental area using traps to investigate the spatial and temporal variation in abundance, species richness, and species composition of invertebrate communities. A total of 60 traps were operated simultaneously during each sampling period. Traps were emptied 26 times every 7-9 days each year. Plant water availability, precipitation, wind speed, air temperature (minimum, maximum, daily mean), air humidity, and atmospheric pressure were used as ecological predictors of invertebrate community status and structure. Two-dimensional geographic coordinates of sampling locations were used to create a set of orthogonal spatial variables based on eigenvectors. We used time series of sampling dates to produce a set of orthogonal eigenvector time variables. The moisture content in technosols was the most important factor determining the terrestrial invertebrate community's temporal dynamics under semi-arid climate and reclaimed ecosystem conditions. Each ecological group of terrestrial invertebrates is homogeneous in terms of moisture gradient (xerophilic, xerozoophilic, mesophilic) and has a specific set of patterns best explain the species response to water content in technosols. However, one should consider the fact that the species response to soil water content is influenced not only by soil water content but also by a complex of other environmental, temporal and spatial factors. That is why the effect of other factors on the species response must be extracted previously to find real estimations of the species optima and tolerance. This task can be solved using the constrained correspondence analysis (CCA) or constrained redundancy analysis (RDA) depending on the type of response to ecological factors prevailing in the community – monotone or unimodal. We found that in more dry conditions, the prevalent species responses are unimodal asymmetric, in moister – bimodal, and in moderate conditions, the distributions are symmetric unimodal. The asymmetric species response to soil moisture in different parts of the soil humidity range may be assumed as predominantly due to the abiotic factors in the gradient's aridest margin and due predominantly to the biotic factors in the most humid margin of the gradient.

**Keywords:** species response, niche, optima, tolerance, reclamation, gradient, temporal dynamic

**Introduction**

The measurement of niche properties concerns such questions as the problem of range, spacing, and the problem of nonlinearity (Colwell, Futuyma, 1971). The shape of the response curves has theoretical and practical importance (Austin, 1999). The niche may be quantitatively described by niche position and niche width (Gregory, Gaston, 2000). The species response parameters like optima or niche width depend on the model shape (Jansen, Oksanen, 2013) and can be used to explain the ecological behavior of species (Michaelis, Diekmann, 2017). Apart from the optimum, all model attributes are heavily influenced by the data set's properties, which is why it appears to be challenging to establish the "true" niche characteristics of a species (Michaelis, Diekmann, 2017). Niche width (or niche breadth, niche size, versatility) is the distance through a niche along some particular line in niche space (Colwell, Futuyma, 1971; Smith, 1982). Niche width is essentially the inverse of ecological specialization (Kohn, 1968). The different aspects of the niche size such as climatic tolerance, habitat, or diet breadth were revealed to affect the species range size (Gaston et al., 1997). The term specialization is used for different biological organization levels (individual, species, population, or community) and measured at very different spatial scales (Devictor et al., 2010). The intraspecific and interspecific interactions and environmental conditions affect niche position and niche width (Brown, 1999; Lawton, 1999). Interspecific competition is a dominant force in animal communities that induces niche shifts (Tarjuelo et al., 2017). The general diversity theory integrating metabolic principles with niche-based community assembly predicts that a decrease in temperature should reduce niche widths along environmental gradients due to decreasing growth rates. The changing niche widths should lead to contrasting αβ-diversity patterns (Okie et al., 2015). The temporal and spatial dynamics of population abundance or community diversity may be explained based on the niche theory (Schoener, 1989; Tokeshi, 1999). Niche theory predicts that species' distribution is forced by the differences between species optima and environmental factors (Hutchinson, 1957). Environmental filtering, biotic interactions, and interspecific competition have important implications for
species diversity and composition (Chase, Myers, 2011). A set of biotic and abiotic conditions under which a given organism can survive and reproduce is considered its ecological niche (Hutchinson, 1957). The ecological niche may be considered in the context of the two dimensions: Grinnellian and Eltonian (Soberon, 2007). The Grinnellian niche considers the importance of a given set of resources for the survival of a species (Deviotor et al., 2010). The Grinnellian niche is considered in two ways: on the one hand, as a complex of the habitat conditions and, on the other hand, as behavioral adaptations allowing organisms to persist and produce offspring (Grinnell, 1917). The Eltonian niche is concentrated on considering species response to environmental impact and the consideration of species impact in the environment. The niche reflects a species’ place in the biotic environment, its relations to food and enemies (Elton, 1927).

For the unimodal species response model, the simplest way to estimate the species optimum is by calculating the weighted average of the value of environmental variables in the samples where the species is present (Šmilauer, Lepš, 2014). The weighted averaging approach for species relative importance assessment was developed by Curtis and McIntosh (1951). The Gaussian bell-shaped response model is the classical and frequently used model (Gauch & Whittaker, 1972). Suppose the relationships between species occurrences and values of a quantitative environmental variable conform to bell-shaped curves. In that case, the optimal for species values of the environmental variable may be found by averaging over the samples in which a species occurs or can be weighted by species abundance (ter Braak & Looman, 1986). Correspondence analysis is based on the assumption of symmetrical unimodal species responses along environmental gradients (ter Braak, 1985), which is why weighted averaging is the basis of this ordination technique (Hill, 1973). The generalized linear models allow the modeling of symmetric bell-shaped response curves (ter Braak & Looman, 1986). The Huismann-Olf-Fresco (HOF) approach provides a comprehensive set of models characterized by statistical correctness, flexibility, and ecological interpretability (Huismann et al., 1993). This model framework was expanded to comprise seven ecological niche patterns (Jansen & Oksanen, 2013).

The terrestrial macroinvertebrate communities of arid and semi-arid ecosystems are represented by insects, spiders, mollusks, and myriapods (Gerlach et al., 2013). Among insects, the ground-dwelling beetles (especially Carabidae, Scarabaeidae, and Tenebrionidae) form the most considerable community component in the arid ecosystems (Konstantinov et al., 2009). Spiders are dominant predators and a species-rich group in most terrestrial ecosystems (Wise, 1993; Rushton, Eyre, 1992). Saprophagous macroarthropods provide a crucial role in organic matter turnover within most ecosystems (Paolletti et al., 2007). Millipedes, woodlice, and other saprophagous macroarthropods are classified as litter transformers (Lavelle et al., 1997) and represent critical regulators of plant litter decomposition within ecosystems (David, Handa, 2010). Saprophagous macroarthropods’ sensitivity to environmental impacts makes them invaluable bioindicators (Paolletti, Hassall, 1999; Souty-Grosset et al., 2005). Humidity and moisture affect the activity and distribution of terrestrial isopods (Warburg et al., 1984). Climatic factors, predation, and parasitism are the frequent causes of the Isopoda population dynamic (Warburg et al., 1984). Shrub cover and herbaceous species richness were shown to be the best drivers of detritivores and species composition of the herbivores beetle community. For predator species composition, shrub height and cover were the best predictors (Liu et al., 2016). The herbivorous and predatory arthropods’ range characteristics were correlated with diet breadth (Brandle et al., 2002; Beck, Kitching, 2007) and habitat breadth (Brandle et al., 2002). The distribution and abundance of spiders depend on three niche axes: wind, moisture, and temperature (Wise, 1993). Spider species occurrence is frequently related to vegetation structure, soil moisture, disturbance, and management regime (Marc et al., 1999; Bonte et al., 2002). The importance of habitat landscape properties on spider assemblages was reported (Gallé et al., 2011). Four separable spider species groups (salt marshes, dunes, meadows, and floodplain forests) along a soil salinity and moisture gradient were extracted (Buchholz, 2009). It is also important to note that several studies failed to find convincing correlations between the environment and spiders’ occurrence (Mallis, Hurd, 2005). The spatial distribution of carabids around arable field-woodlot boundaries was explored using logistic curves with the Huismann-Olf-Fresco models (Knapp et al., 2019). The species-specific effects of shrub cover on Arthropoda animal groups that act as indicators of degradation were revealed due to focusing on species niche breadths and optima (Hering et al., 2019). Coexistence mechanisms at multiple scales in insect assemblages were explored with Huismann-Olf-Fresco models (Laporta, Sallum, 2014). The vegetation and vegetation-soil interactions are essential determinants of beetle community assemblages. The vegetation changes had more substantial effects on the functional group composition than changes in the soil (Liu et al., 2016).

Deep underground hard-rock mining has a considerable impact on the landscape, transforming original habitats and leaving landscapes in altered states (Szczepanska, Twardowska, 1999; Hodecek et al., 2015; Klímkina et al., 2018; Zhukov, Maslikova, 2018; Yorkina et al., 2019). The landscape transformation due to open-pit coal mining causes significant changes in terrain structures, waterways, microclimates, land uses, and living organism communities (Sklenicka et al., 2004; Hendrychova et al., 2011; Yorkina et al., 2018; Zadorozhnaya et al., 2018). The technical and biological reclamation phase is essential for managing a disturbed landscape (Hildmann, Wunsche 1996; Rehor et al., 2006; Hodecek et al., 2016). The reclamations processes’ technical phase is a significant disturbance that slows down the successional rate of beetle communities (Hodecek et al., 2016). The ecological niche principle has to be used for a conceptualization of the restoration of the mine- degraded lands (Zhenqi et al., 2012; Kunakh et al., 2018). The soil formation is essential for successful reclamation of post-mining sites (Madej, Kozub, 2014). The soil macrofauna distribution pattern was found to be strongly related to soil development after reclamation (Ge et al., 2014; Zhukov et al., 2018). Animal post-reclamation succession can be influenced by plant community diversity or abiotic conditions (Hendrychova, 2008; Buchori et al., 2018; Zhukov, Gadorozhnaya, 2016). The Coleoptera communities on reclaimed ecosystems were shown to be dependent upon the herb and soil cover (Hodecek et al., 2016). The species richness of invertebrate groups representing various trophic levels and diverse spatial niches was revealed to be affected by different soil properties, microclimatic conditions, and management history (Hendrychová et al., 2012). The diversity of insects in reclamation areas was revealed as being affected by environmental factors such as tree age, vegetation diversity, and soil chemistry (Dunger et al., 2001; Buchori et al., 2018). The microclimatic condition, moisture, herb cover, and forest composition were identified as the
most significant environmental variables affecting an invertebrate community of areas remaining after opencast brown coal mining (Hendrychová et al., 2008). The epigec invertebrates in the early stages of technosols' colonization presented a succession, indirectly linked to soil parameter changes (Hedde et al., 2018). The application of organic waste combined with revegetation led to an immediate increase in beetle numbers (Kielhorn et al., 1999). The small-sized Carabidae species with high adaptive abilities and high ecological plasticity dominate reclaimed lands (Kędzior, 2018). The "niche-breadth" hypothesis sensu Brown (1984) was applied to explain the local communities' structure in non-reclaimed mining sites (Brandle et al., 2003). Investigations of niche characteristics of the terrestrial arthropods are still rare (Entling et al., 2007), and there is no evidence about species response within terrestrial invertebrates groups to the effect of the environmental gradients of the reclaimed lands.

Drylands include arid, semi-arid, and dry subhumid areas. Water is the main limiting factor affecting the dynamic of the ecological processes in many semi-arid grassland ecosystems (Chase et al., 2000). In dryland areas, precipitation is in deficit and exhibits a high temporal variability, and is mostly unpredictable (Reynolds et al., 2007). Water, soil nutrients, and plant biomass are the essential resources in arid or semi-arid conditions, for which long periods of low abundance are interrupted by relatively short periods of high abundance. Rainfall events usually trigger short periods of high resource availability until moisture is depleted (Schwinnig et al., 2004). Ecological processes in arid lands are often described by the pulse-reserve paradigm (Collins et al., 2014). High air temperatures, low humidity, and abundant solar radiation result in high potential evapotranspiration, leading to water deficit. Many dryland soils contain small amounts of organic matter and have low aggregate strength (Reynolds et al., 2007). These ecological conditions are the most extreme within anthropogenically transformed territories, including reclimation ecosystems (Hendrychova, 2008).

In this work, various approaches were applied to assess the soil moisture optimum and tolerance of the ecological niche and temporal projection of terrestrial invertebrates within an experimental polygon set up to investigate the dynamics of reclamations processes after deep underground mining of hard rock in steppe arid regions of Ukraine. Our main objective was to determine whether the variability of water content in technosols affects the temporal dynamics of terrestrial invertebrates and find the properties of species response curves on soil water content.

Materials and methods

Experimental polygon

The research was carried out at the Research Centre of the Dnipro State Agrarian and Economic University in Pokrov city. The experimental polygon for the study of optimal regimes of agricultural recultivation was established in 1968–1970. Sampling was carried out on a variant of artificial soil (technosols) formed on loess-like loam, red-brown clay, green-grey clay, the technological mixture of the rocks, and formed on loess-like loam with a humus-rich 70 cm topsoil layer (the geographic coordinates of the experimental polygon are 47°38'55.24"N, 34°08'33.30"E). According to WRB 2007 (IUSS Working group WRB, 2007), the examined soils belong to the RSG Technosols. The examined profile also satisfies the prefix qualifier Spolic, having 20 percent or more artifacts (consisting of 35 percent or more of mine spoil) in the upper 100 cm from the soil surface. From 1995 to 2003, a long-term legume-cereal agrophytocenosis grew on the site, after which the process of naturalization of the vegetation began (Yorkina et al., 1999; Pontegnie et al., 2005) and emptied each 7–9 days. In 2013 pitfalls were placed on April 9. The pitfalls were emptied 26 times each year. The exact sampling dates were as follows in 2013: April 15, April 22, April 30, May 7, May 14, May 21, May 28, June 4, June 11, June 18, June 25, July 2, July 9, July 17, July 24, July 31, August 7, August 15, August 23, August 30, September 7, September 15, September 23, September 30, October 7, October 14. In 2014 pitfalls were placed on April 5. The exact sampling dates were as follows in 2014: April 13, April 20, April 28, May 5, May 13, May 21, May 30, June 8, June 16, June 24, July 1, July 8, July 17, July 25, August 1, August 9, August 18, August 26, September 4, September 11, September 18, September 26, October 5, October 12, October 19, October 27. In 2015 pitfalls were placed on April 6. The exact sampling dates were as follows in 2015: April 14, April 21, April 28, May 5, May 13, May 20, May 27, June 3, June 10, June 17, June 24, July 1, July 9, July 17, July 24, August 2, August 11, August 18, August 26, September 4, September 11, September 20, September 28, October 6, October 15, October 22. The pitfall traps were made of 1-liter glass cups (10 cm in diameter) buried in the soil with the cup's rim with the soil surface. The cups were filled with 250 ml of preservative fluid made from concentrated NaCl solution, which seems to be efficient for collecting invertebrates and is relatively non-toxic to non-target organisms. Traps were covered by roofs made of a polyethylene sheet to prevent flooding during heavy rain. Roofs were placed 5 cm above the cups' openings using four long rods that penetrated the corners of the roofs.

Vascular plant species lists were recorded in 2015 for each site, and visual estimates of species cover using the nine-degree Braun-Blanquet scale (Westhoff, van der Maarel, 1978). The projective cover of plant species was recorded at ground level. We were able to make species-level identification for all sites. Based on geobotanical descriptions, phytoindicative assessment of soil humidity, according to Didukh (2011), was made. Phytoindicational evaluation of environmental factors was performed by the ideal indicator method of Buzuk (2017).
The readily available water for plants (RAW, mm in 1 m soil layer), precipitation (mm), wind speed (m/s), atmospheric temperature (daily minimum, daily maximum, daily mean, °C), atmospheric humidity (%) and atmospheric pressure (gPa) were used as environmental predictors (Kunah et al., 2019). The Nikopol Meteorological Station data were used. The readily available water was calculated as follows. The daily evapotranspiration was computed based on the weather data on maximum and minimum temperature, solar radiation, relative humidity, and wind speed at 2 m height (Monteith, 1965; Allen et al., 1994a, b; Allen et al., 1998; Evett et al., 2011). This approach has been tested and validated (Pereira et al., 2003; Popova et al., 2006; Jabloun, Sahli, 2008). The FAO Penman-Monteith reference evapotranspiration equation is the primary tool to calculate evapotranspiration from meteorological data (Penman, 1948; Allen et al., 1998). The Penman-Monteith equation has been revealed to be reliable in various environments (Hess, 1996). The first step of actual evapotranspiration estimation includes calculating potential evapotranspiration from meteorological data using equations based on the aerodynamic theory and energy balance (Penman, 1948; Monteith, 1965). The potential evapotranspiration is then used to estimate actual evapotranspiration after applying a soil water reduction factor based on available or extractable soil water (Slabbers, 1980). Reference crop evapotranspiration is the evapotranspiration from a crop with specific characteristics. FAO-56 method sets the specific characteristics of a reference crop with a certain height (0.12 m), surface resistance (70 s m⁻¹), and albedo (0.23) and then determines the reference evapotranspiration using the Penman-Monteith equation (Allen et al., 1998). The soil water balance is performed in ISAREG (Teixeira & Pereira, 1992) with a daily time step as:

\[ SW_i = SW_{i-1} + P_i + l_i + G_i - E_i - D_i \]  

where \( SW_i \) and \( SW_{i-1} \) are respectively the soil water storage (mm) in the soil layer zone at the end of day \( i \) and of the previous day, \( i-1 \); \( P_i \) is the precipitation; \( l_i \) is the net irrigation depth; \( G_i \) is the capillary rise; \( E_i \) is the actual evapotranspiration, and \( D_i \) is the deep percolation out of the root zone, all referring to day \( i \). All units but for \( SW \) are in mm d⁻¹. \( l_i \), \( G_i \), and \( D_i \) were neglected in this application. Readily available water was found as:

\[ RAW = SW - PWP, \]  

where \( RAW \) – readily available water (mm), \( SW_i \) is the soil water storage (mm), \( PWP \) – permanent wilting point (mm). PWP data were used from the article Zhukov & Maslilkova (2018). For each time pitfall exposition interval, the following environmental parameters were calculated: total precipitation sum and mean per period value of the readily available water, wind speed, atmospheric temperature, atmospheric humidity, atmospheric pressure.

Ecological structures and processes occur across multiple spatiotemporal scales (Nash et al., 2014). Ecological systems are hierarchically structured and may be decomposed into structural and process components (La Notte et al., 2017). These components can be defined over a range of spatial and temporal scales (Allen et al., 2014). The elucidation of the mechanisms underlying observed spatial and temporal patterns and different scale levels is a key to prediction and understanding to the development of management principles (Levin, 1992). The importance of stochastic processes was found to be dependent on the size of the study area (Bonsall, Hastings, 2004). The complexity has to be accounted for to model temporal patterns in time series (Baho et al., 2015). The neighbor matrix method (PCNM)’s principal coordinates are a powerful approach that can detect temporal structures of varying scales in time series data (Borcard, Legendre, 2002; Borcard et al., 2004; Dray et al., 2006). PCNM, after conversion of the time vector, produces a series of temporal variables with distinct sine-wave properties. This approach has become famous for assessing hierarchical dynamics and resilience of ecological systems (Angeler et al., 2011; Spanbauer et al., 2014). The PCNM-based approach has been further elaborated in distance-based Moran’s eigenvector maps (MEM) (Dray et al., 2006; Baho et al., 2015). MEM can increase the proportion of explained variation compared to the original PCNM approach and are more robust in dealing with correlation structures (Dray et al., 2006; Baho et al., 2015).

The two-dimension geographic coordinates of sampling locations were used to generate a set of orthogonal eigenvector-based spatial variables (Spatial dbMEMs), each representing a particular scale pattern within the extent of the sampling area. Likewise, the one dimension time series of sampling dates were used to generate a set of orthogonal eigenvector-based temporal variables (Time dbMEMs), each representing a particular scale pattern within the extent of the investigated period (Borcard, Legendre, 2002). The spatial and time dbMEMs-variables were used as spatial and temporal predictors of the species responses. Weighted averaging can be used to estimate the species’ optimal value along the ecological gradient in cases of the symmetric bell-shaped response curves (ter Braak & Looman, 1986). The weighted averaging method is reasonably fair when samples cover the whole range of species distribution. The species abundances were used as the weights in calculating the ecological factor average (Šmilauer & Lepš, 2014):

\[ WA = \frac{\sum_i Env_i \times Abund_i}{\sum_i Abund_i}, \]

where \( Env_i \) is the value of an environmental variable in the \( i \)-th sample, and \( Abund_i \) is the abundance of the species in the \( i \)-th sample.

The species tolerance presented by the width of the bell-shaped curve can be calculated as the square root of the weighted mean of the squared differences between the species optimum and the actual value in the sample. The value is analogous to the standard deviation (Šmilauer & Lepš, 2014):

\[ SD = \sqrt{\frac{\sum_i (Env_i - WA)^2 \times Abund_i}{\sum_i Abund_i}}, \]

If the complete range of species distribution is covered and species response is symmetrically bell-shaped, then the weighted average estimation is correct. In contrast, the estimate is biased if only a part of the range is covered. In this case, the estimate is shifted concerning real value in the tail’s direction that is not truncated. The number of species with truncated distribution will increase if the covered portion of the gradient is short, and as a consequence, the optimum estimates will be biased. The longer the environmental gradient, the more species will have their optima estimated correctly (Šmilauer & Lepš, 2014).
Generalized linear mixed models were used for modeling species-environment relationships. This approach is useful in the unimodal symmetric species response (Jamil, ter Braak, 2013). In the environment R the model can be fitted by (Oksanen, 2004):

```r
mod <- glm(y ~ x + I(x^2), family=poisson)
b <- coef(mod)
va <- b[2]/2/b[3]
```

where \( y \) – species response, \( x \) – environmental factor, \( u \) – species optimal value, \( t \) – species tolerance.

The use of symmetric Gaussian response functions in gradient analysis is not a universal approach due to systematic deviation of the real data from the symmetric response (Austin, 1976, 1999, 2013; Austin et al., 1999). Huisman, Olff, and Fresco’s (1993) hierarchical models (HOF) and symmetric response also include a skewed response. Apart from the five HOF-model two bimodal (skewed and symmetric) response shapes, they were included to cope with species restricted to gradient extremes due to competition (Jansen, Oksanen, 2013; Michaelis, Diekmann, 2017). The Huisman-Olff-Fresco models expanded by Jansen-Oksanen (HOFO) are ranked according to the increasing complexity of biological information contained (Huisman et al., 1993; Jansen & Oksanen, 2013). The model I: no significant trend in space or time:

\[
y = M \frac{1}{1 + e^{a+b x}}
\]

Model II: an increasing or decreasing trend where the maximum is equal to the upper bound M:

\[
y = M \frac{1}{1 + e^{a+b x}}
\]

Model III: an increasing or decreasing trend where the maximum is below the upper bound M:

\[
y = M \frac{1}{1 + e^{a+b x}}
\]

Model IV: increase and decrease by the same rate – symmetrical response curve:

\[
y = M \frac{1}{1 + e^{a+b x}}
\]

Model V: increase and decrease by different rates – skewed response curve:

\[
y = M \frac{1}{1 + e^{a+b x}}
\]

Model VI: bimodal symmetric responses:

\[
y = M \frac{1}{1 + e^{a+b x}}
\]

Model VII: bimodal skewed responses:

\[
y = M \frac{1}{1 + e^{a+b x}}
\]

where \( y \) and \( x \) are the response and the explanatory variable respectively, \( a, b, c, \) and \( d \) the parameters to be estimated (\( b \) and \( d \) have opposite signs), and \( M \) – a constant which equals the maximal value which can be attained (for relative frequencies \( M = 1 \), for percentages \( M = 100 \)), \( L \) – a constant which equals the maximal value for minor extreme value.

Huisman-Olff-Fresco models were fitted in the R statistical program (v. 3.3.1) (R Developmental Core Team, 2019) using the package "eHOF" (Jansen & Oksanen, 2013, version 3.2.2). The stability of model choice was double-checked by bootstrapping to improve modeling results even for small data sets (100 samplings, default package setting) to ensure model robustness. The Akaike information criterion corrected for small data sets (AICc) was used (Burnham, Anderson, 2002). When the two procedures differed in their choice for the best model type, the bootstrapping model was preferred (Michaelis, Diekmann, 2017). Optimum value and relative species tolerance limits, or the Central Borders, were calculated as implemented in the eHOF package (Jansen & Oksanen, 2013). These are specified fractions of the curve maxima (max * e^0.5) and are calculated separately for the left (LowCB) and right (UppCB) hand side of the optimum (Heegaard, 2002).

We applied multivariate ordination techniques to analyze the spatiotemporal variation in the species composition of invertebrate assemblages. For ecological data with many zeros, the Hellinger transformation is useful (Rao, 1995). In this regard, before analyses, species data were Hellinger-transformed (Legendre, Gallagher, 2001). There were no rare species represented by less than seven sites; that is why all species were included in the analyses, resulting in a final dataset of 202 species or parataxonomic units. We subjected the Hellinger-transformed abundance matrix of species to correspondence analysis (CA), constrained correspondence analysis (CCA), and constrained redundancy analysis (RDA) to extract the significant patterns of variation (Legendre, Birks, 2012; ter Braak, Smilauer, 2015). The environmental factors were fitted onto a CA-ordination by enfit function from the vegan library (Oksanen et al., 2018). The constrained ordination approaches (correspondence or redundancy analysis) allowed assessing the effects of the soil moisture as an explanatory variable on the invertebrate community with temporal, spatial, environmental factors, and technosol type as condition variables. The detrended correspondence analysis (DCA) was used to discriminate whether species responses are primarily monotonic or primarily unimodal and the length of the first significant gradient of variation in community data was estimated (ter Braak, Prentice, 1988). If gradient length is more than two standard deviations, constrained correspondence analysis (CCA) must be selected as an ordination approach. Otherwise constrained redundancy analysis (RDA) is most suitable. Species scores generated by the CA or CCA indicate the center of the species’ distribution in a unimodal model. Therefore, species scores represent the niche position (optimum) of species along the extracted axes (ter Braak, Smilauer, 2002). The standard deviations of species scores quantify the niche width. If the ordination axes are correlated with environmental gradients, the scores and the associated standard deviations can be used to characterize the niche properties of each species concerning that gradient (Entling et al., 2007). For the statistical analyses, we used the appropriate procedures of Statistica (Version 5.5, StatSoft Inc., http://www.statsoft.com) or R (version 3.5.2; R Core Team, 2018).
Results and discussion

Precipitation falls very unevenly in time on the investigated area. In 2013, the duration of rainless periods was 259 days; in 2014 – 264 days, in 2015 – 261 days. The maximum daily rainfall varied within 18–49 mm. The highest amount of precipitation fell in the year 2015 (506.8 mm), and the lowest – in 2014 (328.9 mm) (Table 1).

Table 1. Descriptive statistics of the daily meteorological characteristics based on observations at Nikopol meteorological stations and readily available water content in the technosols (2013–2015)

| Parameters                          | 2013     | 2014     | 2015     |
|-------------------------------------|----------|----------|----------|
| Precipitation, mm                   | 0.95±0.15| 0.90±0.15| 1.39±0.23|
| Wind speed, m/s                     | 2.39±0.06| 2.28±0.07| 2.00±0.08|
| Temperature, ºC                      | 11.33±0.50| 10.89±0.56| 11.21±0.51|
| $T_{min}$, ºC                        | 7.49±0.43| 6.47±0.49| 6.43±0.45|
| $T_{max}$, ºC                        | 15.10±0.57| 15.17±0.62| 15.88±0.58|
| Atmospheric humidity, %             | 73.37±0.78| 69.12±0.91| 69.59±0.83|
| Atmospheric pressure, gPa            | 1014.85±0.39| 1014.67±0.40| 1009.26±0.43|
| Readily available water, mm in 1 m soil layer | 99.44±3.30| 49.14±1.34| 105.14±2.53|

In 2013 precipitation was 345.6 mm. The intensity of the rainfall varies throughout the year. The highest rainfall usually occurs in June and the lowest – in August. There are significant interannual differences in the intensity of rainfall. Minimum total annual precipitation in 2014 was due to decreased atypical rainfall in late winter and early winter. The average annual temperature was 11.14 ± 0.30ºC and was not statistically significantly different between years during the study period ($F = 0.19$, $p = 0.82$). The temperature range was from –23.4 to + 37.8ºC during the study period. The average wind speed was statistically significantly different from year to year ($F = 8.72$, $p <0.001$).

Table 2. Plant species number and phytoindicator estimation of the soil humidity condition

| Site | Technosols                  | Species number | Soil humidity indicator value | Phytoindicator estimation of the readily available water content, mm | The favorable eco-groups by the relation of plant species to soil water regime* |
|------|-----------------------------|----------------|------------------------------|---------------------------------------------------------------|--------------------------------------------------------------------------------|
| 1    |                             | 23             | 12.6                         | 123.2                                                         | Mesophytes                                                                     |
| 2    |                             | 17             | 10.3                         | 87.4                                                          | Sub-mesophytes                                                                |
| 3    | Loess-like loam             | 21             | 12.6                         | 124.0                                                         | Mesophytes                                                                     |
| 4    |                             | 15             | 12.0                         | 113.0                                                         | Mesophytes                                                                     |
| 5    |                             | 20             | 12.2                         | 116.6                                                         | Mesophytes                                                                     |
| 6    |                             | 26             | 11.9                         | 110.6                                                         | Mesophytes                                                                     |
| 7    |                             | 11             | 12.8                         | 127.3                                                         | Mesophytes                                                                     |
| 8    | Red-brown clay              | 31             | 8.4                          | 65.7                                                          | Sub-xerophytes                                                               |
| 9    |                             | 30             | 8.6                          | 67.7                                                          | Sub-xerophytes                                                               |
| 10   |                             | 13             | 12.6                         | 124.0                                                         | Mesophytes                                                                     |
| 11   |                             | 23             | 9.1                          | 72.9                                                          | Sub-mesophytes                                                                |
| 12   | Green-grey clay             | 23             | 8.7                          | 68.5                                                          | Sub-xerophytes                                                               |
| 13   | Humus-rich 70 cm top soil layer formed on loess-like loam | 13             | 7.4                          | 56.7                                                          | Sub-xerophytes                                                               |
| 14   |                             | 16             | 11.3                         | 102.0                                                         | Mesophytes                                                                     |
| 15   |                             | 26             | 12.1                         | 114.7                                                         | Mesophytes                                                                     |
| 16   |                             | 16             | 12.7                         | 124.8                                                         | Mesophytes                                                                     |
| 17   |                             | 15             | 6.9                          | 52.8                                                          | Xerophytes                                                                    |
| 18   | Technological mixture of rocks | 16             | 9.0                          | 71.5                                                          | Sub-xerophytes                                                               |
| 19   |                             | 19             | 9.0                          | 71.6                                                          | Sub-xerophytes                                                               |
| 20   |                             | 20             | 6.7                          | 51.2                                                          | Xerophytes                                                                    |

Note: * – according to Didukh scale (2011).

The highest wind speed was observed in 2013, and the lowest – in 2015. The average atmospheric humidity is statistically significantly varied from year to year ($F = 7.67$, $p <0.001$). The highest humidity was found in 2013, differences in humidity in
2014 and 2015 were not significant. The mean atmospheric pressure is statistically significantly varied from year to year \( (F = 60.22, p < 0.001) \). The lowest atmospheric pressure was observed in 2015. The difference between 2013 and 2014 was not statistically significant.

In total, 35 species of herbaceous plants were found within the experimental polygon. At each site, the number of plant species varied from 11 to 31 (Table 2).

The phytodiversity estimation of soil moisture was in the range 6.7-12.8 according to 23-point of the Didukh (2011) soil moisture scale. The phytodiversity values can be converted to the value of the readily available water content, which can be estimated to be in the range of 51.2-127.3 mm per meter depth of the soil. The soil humidity range can be evaluated qualitatively as favorable for plants range from xerophytes to mesophytes.

In total, 257,437 invertebrate (Arthropoda and Mollusca) individuals of 6 classes, 13 orders, 50 families, and 202 species or parataxonomic units were recorded. Diplopoda was the most abundant taxonomic group, though it was represented only one species *Rossiulus kessleri* (Lohmander, 1927). This species took up 49.4 % of the total community abundance. Coleoptera and Araneae were the considerably numerous taxonomic groups, which took up 22.4 and 18.2 % of the total community abundance. These taxa were the most species-rich. One hundred twenty-two species represented Coleoptera, and 67 species represented Araneae. Seventy-one species among the investigated 202 species occurred 50 times or less. The distribution of plant-available moisture weighted by species abundances is multimodal and can be presented as a mixture of three normal distributions \( (\text{Kolmogorov-Smirnov } d = 0.037, p = 0.93) \). The plant available moisture has a 0.18 chance of the following distribution with a mean of 30.5 mm and variance of 11.5. The chance is 0.66 that plant-available moisture distribution has a mean of 87.2 mm and a variance of 29.3. A distribution mixture component with a mean of 144.5 mm and variance of 6.4 has a 0.16 chance. The distribution of species tolerance to soil water content assessed based on the variance of the plant available moisture weighted by species abundances is bimodal and can be presented as a mixture of two normal distributions \( (\text{Kolmogorov-Smirnov } d = 0.027, p = 0.99) \). A mixture distribution component with a mean of 12.6 and variance of 3.0 has a 0.17 chance. The next component with a mean of 33.5 mm and variance has a 0.83 chance, respectively.

HOFJO-approach provides opportunities for species response modeling using more alternative models. Model IV often is the most optimal model of the species response to a soil moisture gradient. Somewhat less often, the optimal models are III, V, and VII. Thus, species response to the soil's moisture content's influence can be described by Gaussian curve (model IV) or close to its asymmetric bell-shaped curve (model V). For ordination procedure of the community with monotonic patterns, the more appropriate is RDA, and with unimodal patterns, the more appropriate procedure is CCA (ter Braak, Prentice, 1988). DCA revealed that the length of the first axis gradient is more than two standard deviations \( (2.49 \text{ in our case}) \), which points to the appropriateness of applying the correspondence analysis (CA) and the constrained correspondence analysis (CCA) as an ordination procedure.

CA-ordination axes may be explained by the environmental factors, time and spatial variables, and the technosol type. The first four CA-axes can explain 19.1 % of the total community variation. Axis 1 is the most correlated with soil moisture (RAW), but a significant correlation is found between soil moisture and CA-axis 4. Temperature and atmospheric pressure are considerable predictors of the invertebrate community variation. The significance of the time predictors is strongly ranged in descending order of the temporal dBMEMs-variables. This result reveals that the importance of the time oscillations decreases with the increasing of their frequencies. The spatial predictors considerably affect the community variation. The HOFJO-approach showed that species responses to soil moisture gradient are mainly bell-shaped (for which CCA is the best ordination procedure). Nevertheless, many species' responses are monotonic (for which the RDA is the best ordination procedure). The fractioning of the animal community variability concerning meteorological data, spatial and time predictors, and the technosol type was performed based on the CCA and RDA.

The variation partitioning of the community based on CCA-approach points to the major role of the complex factors that result from the interaction of temporal and spatial factors and the temporal, spatial, and meteorological factors. The results based on RDA-approach also indicate the significant role of interaction between weather and time factors. These results point to the need to extract the role of soil moisture factor interactions with other factors to precisely assess the influence of soil moisture on the dynamics of invertebrate communities. The soil moisture influence with considering the conditional effect of the time, space, technosol type, and meteorological factors can explain 2.1% of the community variation \( (F = 44.04, p < 0.001) \). CCA-axis statistically significantly correlates with readily available water \( (r = -0.53, p < 0.001) \). The statistically significant correlation of this axis with meteorological parameters, time, and space variables was not found. There is no statistically significant effect of the technosol type on this axis \( (F = 0.04, p = 0.99) \). The species scores on CCA-axis allow us to estimate the species optimum within the humidity gradient, and the mean-square deviation of the species scores are indicators of tolerance to moisture conditions. Correlation analysis of the species' optimal values obtained by a set of methods shows a different level of relationship between these estimations. The weighted average and optimum values estimations derived from the regression model are the most correlated \( (r = 0.97, p < 0.001) \). The regression-derived estimation corresponds to model IV, which is applied to all distributions because it provides the basic component of the correlation between the regression evaluation of the species optimums and HOFJO-assessment of the optima. Other models that consider the different levels of asymmetry or bimodality of the species' responses to soil moisture gradient provide the optimum assessments that are significantly different from the regression approach and the weighted mean method. According to the regression approach and CCA-approach, the largest deviations from the monotonous dependence between optimum estimations are seen in the central zone of the soil humidity gradient for models IV and V. Thus, considering the impact of other factors of environment, time, space, and interaction between species in a community by using the constrained ordination procedure provides a wider range of estimates of species optimum. These optimums were found in the central part of the soil humidity range (the correlation between estimation according to HOF- and CCA-approaches depending on the types of optimum models has a similar character, so it is not shown). According to CCA-
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approaches, K-mean cluster analysis of the animal species on the basis of the optimum and tolerance estimations allowed us to establish the presence of four isolated clusters. The optimal values and tolerance to the soil humidity are statistically significant different for the clusters ($F = 444.6, p < 0.001$ and $F = 20.7, p < 0.001$ respectively).

Clusters can be meaningfully interpreted as ecological groups of animals characterized by specific relation to the soil moisture conditions. The most major ecological groups based on the relationship to moisture conditions usually are designated as hygrophilous, mesophilous, and xerophilous. Phytoindication evaluation of humidity conditions on the studied technosols indicates a range from mesophilous to xerophilous. Therefore, the species group that is the most demanding of moisture conditions we identified as mesophilous. Accordingly, the species group that can survive with water deficit conditions was identified as xerophilous. Species also show differences in tolerance to moisture conditions. Among the mesophilous species investigated are stenotopic, while xerophilous species are eurytopic. Within the humidity range between the xerophilous and the mesophilous species, two homogeneous ecological groups are transitional in terms of preferences for humidity conditions and are different concerning the tolerance. Thus, we identified stenotopic and eurytopic xeromesophilous. The total number of mesophilous species is 66 (32.7% of the total), xerophilous – 19 (9.4%), stenotopic xeromesophilous – 48 (23.8%), and eurytopic xeromesophilous – 69 (34.2%).

The optimum and tolerance estimations of ecological groups by CCA-approach are significantly different from the assessments by other methods. The main difference is that the soil humidity optimum assessment without considering the effect of the other environmental factors, spatial and temporal variables, and technosol type for xerophilous and xerophilous is biased in the direction of the more humid conditions. The assessment for stenotopic xeromesophilous species is biased in the direction of more dry conditions. Assessing tolerance of the species is also somewhat displaced. So, by the method of weighted mean, xerophilous species are the most stenotopic, and the level of the mesophilous tolerance is almost the same as eurytopic xeromesophilous.

The correspondence between ecological groups and the best response models from the HOFJO-list was revealed. The xerophilous species' response to humidity gradient conditions is often best described by the model V. The response of the eurytopic xeromesophilous species is best described by model VII the response of the stenotopic xeromesophilous species is best described by model IV. For the mesophilous species, the best models are III and VI. Model II is usually best for mesophilous and stenotopic xeromesophilous species. The trophic groups also have their characteristics, typical responses to the influence of the soil moisture. For predators (116 species, 57.4 % of the total), the most typical response models are III and IV (23.3 and 36.2 % of the predators' total). For herbivores (116 species, 35.2 % of the total), the most typical are V and VII (19.9 and 28.2 % of herbivores). For saprophages (15 species, 7.4 % of the total, the most typical are III and IV (33.3 and 40.0 % of saprophages).

Conclusion

Moisture content in technosols is the most crucial factor determining the temporal dynamics of the terrestrial invertebrate community under semi-arid climate conditions and in an ecosystem formed from reclamation. The ecological features of the species niche, such as optimality and resilience to environmental factors, explain the community structure's temporal dynamics. The weighted mean as a way of assessing the species optimum provides only a general indication of a species' ecological preferences in the case of a unimodal model of species response. Similarly, the situation is the same for other indicators that do not include the effects of factor combinations and interspecific interactions. The combination of factors and interspecific interactions can significantly adjust the estimate of the optimum value for several species. The deviation of species responses to soil moisture effects from the symmetrical bell-shape in different parts of the soil moisture range is due to different factors. Abiotic factors dominate asymmetric responses to the physiologically more stressful ends of soil moisture gradients. Biotic factors are predominant at the extremes of the range when moisture conditions are sufficient. The failure to account for soil moisture interactions with other factors leads to a shift in species optimal estimates to the central part of the moisture range.

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