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Application of Regression Models in Bird Population Data: An Example of Haçlı Lake

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ABSTRACT: In this study, the effects of habitat, ordo, UTM frame, seasons and number of species on bird populations and distribution in Haçlı Lake were investigated. Bird population data were obtained using point counts and transect observation methods. Poisson regression is typically used in such data sets. The basic principle of Poisson regression assumes that the variance is equal to the mean. Failure to achieve this equality causes incorrect parameter estimates and standard errors. In practice, the variance is often higher than the mean (variance > mean). This is called over-dispersion, where the value of over-dispersion is greater than 1.0. The population status of the data set used in the study was over-dispersed. Negative binomial regression is the most common method used to eliminate the over-dispersion value in the Poisson regression was considerably greater than 1.0 (54.937) while the over-dispersion value was very close to 1.0 (1.588) in the negative binomial regression. The results indicated that the use of negative binomial regression method is more appropriate. Therefore, parameter estimations were interpreted according to negative binomial regression method. Herein, climatic factors including temperature and humidity exhibited significant impacts on population density and number of species.

Keywords: Bird population, haçlı lake, negative binomial, over-dispersion, poisson regression

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INTRODUCTION

Every living species on earth has a particular environment and habitat (Adızel et al., 2010). Habitats are the main determinants of the distribution and abundance of organisms and form the fundamental of the conservation studies (O'Hara and Kotze, 2010; Boyce et al., 2016). The living organisms in the biosphere are in direct or indirect interaction with each other. Living organisms sharing the same food web and interacting with each other maintain their life span in equilibrium. Therefore, a living group in each step of the food chain possess great roles for the proper and sustainable ecosystem.

Birds, which are the biological indicators of the natural ecosystem, contribute to biodiversity with their habitat preferences and population densities (Kiziroğlu, 2008). Therefore, determining the species and population status of birds in an area is important to reveal the naturalness and ecological importance of the area. Searching for appropriate habitat and modelling the distribution of organisms are increasingly becoming important for ecology and conservation biology. Correlative species distribution models (SDM) assess the relationship between species distribution data and environmental factors, as well as determining habitat suitability for a focus species in a given area (Guisan and Zimmermann, 2000; Graham et al., 2004). Therefore, assessing the effects of these environmental factors on the distribution of living species is very important (Aksan et al., 2014).

Several ecological and environmental factors including vegetation structure (Clark and Shutler 1999; Milsom et al., 2000), human activities (Milsom et al., 2000; Yuan et al., 2014), temperature, precipitation, humidity (Girma et al., 2017), geographical structure (Li and Martin, 1991) and nutrient diversity (Beerens et al., 2011) affect the habitat preferences and geographical distribution of birds.

The most common technique used to determine the bird populations is the counting method (Bibby and Burgess, 2000). Counting studies are typically used to identify changes occurred in the number of organisms in wildlife (Knape et al., 2018). Numerous statistical methods are used in the analysis of the data obtained by counting. Independent data obtained by counting can show Poisson distribution (PD) and are analyzed by Poisson regression (PR) (Ridout, 1998).

The basic principle of Poisson distribution is that the mean and variance are equal (Yeşilova and Denizhan, 2016). However, the counting data, in actual applications, does not always support Poission distribution (Muthen and Muthen, 2006). Over-dispersion is often encountered where the variance is greater than mean (Ver Hoef and Boveng, 2007). In order to eliminate the over-dispersion, negative binomial regression binomial is of the used regression tools (Agresti, 1997; Hilbe, 2007). There are examples of negative binomial regression in many different studies. The negative binomial regression model have been used in many studies including the determination of species richness (O'Hara, 2005), parasite determination in birds (Rekasi et al., 1997), bird population densities (Durmuş et al., 2018; Çelik and Durmuş, 2020), identifying the environmental variables affecting bird migration (Lindén and Mantyniemi, 2011) and estimating the direction and abundance parameters of water seal (Small et al., 2003; Kery et al., 2005).

In this study, Poisson regression and negative binomial regression were used in order to determine the effects of habitat, ordo, UTM frame, seasons and number of species on bird populations and distribution in Haçlı Lake (Muş, Turkey).

MATERIALS AND METHODS

The materials of this study are the Haçlı Lake located in Euphrates-Tigris River Basin of Turkey (38 S 265 952 N 4322968 E) and the birds that use the delta for living purpose. The observations on population density were conducted by 15 day-period in each month, covering 4 seasons between April 2016 - September 2017 and the numerical data obtained were obtained by 18 months of field studies.

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Observations depending on seasonal conditions started with sunrise and ended with sunset. The study area has been divided into 54 UTM squares of $1x1 \text{ km}^2$ (Figure 1). Three observation points were used, at least 300 m far from each other and representing the habitats in each UTM square.

Observation along a Line (line transect) and Point Observation Method (point counts) were used in observations to determine population density and number of individuals (Bibby and Burgess, 2000). Bird species and population numbers identified in the point and surrounding area and UTM coordinates were recorded on field observation cards. The point records obtained in the study were then assigned to UTM squares and subsequently used. This process has enabled the possibility of analysis on a grid basis in addition to the point scale (Onmuş, 2008)

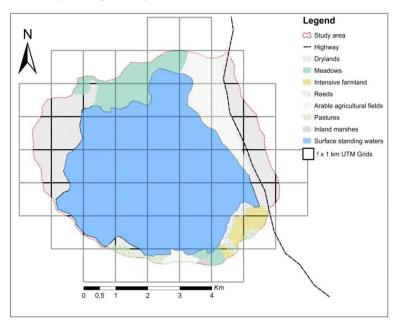


Figure 1. Location of the study area, 1 x 1 km UTM squares, and the major habitats

Statistical Analysis

Poisson regression: Poisson regression analysis is of the used statistical methods in assumption of dependent variable that is observed event number (y_i) and exhibits a Poisson distribution. Poisson mean logarithm (μ) is deemed to be a linear function of the independent variables (Equation 1) (Yeşilova et al., 2016). Herewith the Poisson regression analysis, maximum likelihood estimation (ML) method is used for parameter estimation. Likelihood function for nonlinear Poisson regression model can be written as follow;

$$\ln L = \sum_{i=1}^{n} \left[-\lambda_{i} + y_{i} x_{i}^{\dagger} \beta - \ln y_{i} \right] = \sum_{i=1}^{n} \left[-e^{x_{i} \beta} + y_{i} x_{i}^{\dagger} \beta - \ln y_{i} \right]$$
(Equation 1)

Negative binomial regression: Negative binomial regression model is represented as follows (Hilbe, 2007). For the equation, the auxiliary parameter indicates over-dispersion degree and k is considered as a positive value (Equation 2).

$$P(Y = y \mid X_1, X_2, X_3, k) = \frac{\Gamma(y+k)}{\Gamma(k)\Gamma(y+1)} \left(\frac{k}{k+\mu}\right)^k \left(\frac{\mu}{k+\mu}\right)^y \qquad y = 0, 1, 2, \dots \quad \text{(Equation 2)}$$

Bird population data in Haçlı Lake were used as a model dependent variable. In addition, seasons, ordo and frames were modelled as independent variables and Poisson and negative binomial regressions

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were applied respectively. Necessary statistical analyses were performed using SAS 9.1.1.4 statistical software program.

RESULTS AND DISCUSSION

The results of study carried out for approximately two years indicated the existence of 117 species and 1 subspecies belonging to 14 ordo and 34 families in Haçlı Lake. Of these species, 35.6% (n: 42) were native, 52.5% (n: 62) were migrants, 10.2% (n: 12) were winter visitors and 1.7% (n: 2) were transit migrants. The observations conducted during the reproduction period indicated that 13 species certainly breeding, 12 species probably breeding based on the observations such as courtship behaviour and existence of male and female individuals and 93 species do not produce in the study area. Seven habitat types based on geographical, topographic and floristic characteristics were identified within the boundaries of the study area. The habitats and population status of the ordo groups present in the area were determined by periodical observations during the four seasons (Table 1).

Ordo	Number of species	Population density	Season	Habitat type	Square
201	2	106	11	2	8
201	7	128	11	4	5
201	1	20	11 7		2
201	5	98	22	4	3
201	1	7	22	7	2
201	3	141	22	8	4
201	2	11	33	3	2
201	5	84	33	8	3
201	2	126	44	7	1
202	1	8	22	3	1
203	1	1	11	1	1
203	1	45	22	2	1
204	3	298	11	2	6
204	4	108	11	4	3
204	5	264	11	7	4
204	1	770	11	8	3
204	2	24	22	2	1
204	3	43	22	4	2
204	2	963	22	5	3
204	8	1609	22	7	3
204	1	184	22	8	1
204	1	13	33	2	2
204	1	5	33	4	1
204	2	201	33	5	6
204	6	337	33	7	5
204	1	32	44	4	1
205	2	70	11	2	1
205	1	19	22	2	1
205	1	43	22	5	1
205	1	18	33	5	1
206	2	91	22	3	3
206	4	124	33	3	3
206	1	164	44	7	2
207	1	27	11	1	1
207	1	5	11	2	1
207	2	226	22	1	2
207	1	4	22	2	1
207	2	54	33	1	2
207	1	1	33	2	1

Table 1. Population densities, number of UTM square used and seasonal habitat distributions of ordos defined in Haçlı Lake

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Table 1. (cont.)

Ordo	Number of species	Population density	Season	Habitat type	Square
208	1	3	11	2	1
208	2	3	11	7	2
208	6	29	22	2	1
208	2	3	22	4	1
208	1	4	22	7	1
208	4	19	33	2	1
208	1	4	33	4	1
208	1	4	33	8	1
208	2	5	44	7	2
209	1	60	11	4	2
209	1	320	11	7	4
209	1	1352	22	8	2
209	1	480	33	8	6
210	3	166	11	1	2
210	2	3	11	2	1
210	10	132	11	3	1
210	7	66	11	4	1
210	11	451	22	1	3
210	8	484	22	2	3
210	17	709	22	3	3
210	13	63	22	4	1
210	3	406	22	5	2
210	4	82	22	7	1
210	4	74	33	1	1
210	14	103	33	2	2
210	16	350	33	3	6
210	5	23	33	4	1
210	2	6	33	7	1
210	14	523	44	7	6
211	5	95	11	2	1
211	1	20	11	4	1
211	3	27	22	2	1
211	4	135	22	7	1
211	2	4	33	6	1
211	2	50	33	7	2
211	1	2	33	8	1
211	1	4	44	7	1
212	1	35	11	4	1
212	3	915	22	8	2
212	3	303	33	8	4
213	1	2	22	1	1
214 214	1	7 18	11 22	8 4	2

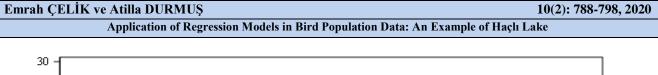
Ordo: 201-Anseriformes, 202-Bucerotiformes, 203-Caprimulgiformes, 204-Charadriiformes, 205-Ciconiiformes, 206-Columbiformes, 207-Coraciiformes, 208-Falconiformes, 209-Gruiformes, 210-Passeriformes, 211-Pelecaniformes, 212-Podicipediformes, 213-Strigiformes, 214-Suliformes Season: 11-Spring, 22-Summer, 33-Autumn, 44-Winter

Habitat type: 1-Drylands, 2-Meadows, 3-Intensive farmland, 4-Reeds, 5-Arable agricultural fields, 6-Pastures, 7-Marshes, 8-Surface standing waters

Numerical data obtained from habitat based observations were arranged for statistical analysis. The population density of the birds in the ordo the number of UTM squares used, as well as the population densities varying depending on the seasons and habitat structures were determined.

Statistical Model

Bird population data obtained by counting in the study area were used as dependent variable of the model. Seasons, ordo, habitat and squares were modelled as independent variables and Poisson and negative binomial regressions were applied respectively. The bird counts used as dependent variable of the model was represented in Figure 2. As seen from Figure 2, the distribution regarding the data of the present study is skewed right. The extreme skewness to the right in such data does not change much despite the transformations (Agresti, 1997; Cameron et al., 1998).



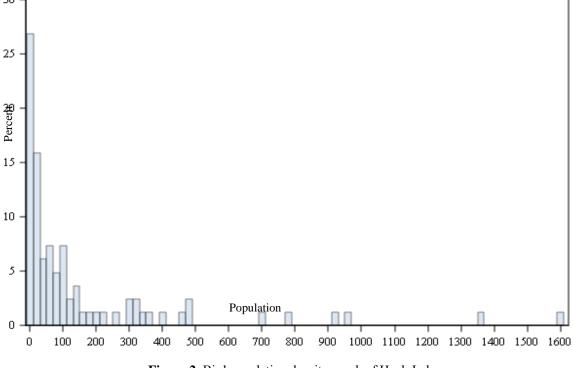


Figure 2. Bird population density graph of Haçlı Lake

Compliance criteria for Poisson and negative binomial regression were given in Table 2. The overdispersion value was obtained through division of deviance statistics using their degrees of freedom. The over-dispersion value in Poisson regression was considerably greater than 1.0 (54.937) while overdispersion value in negative binomial regression was very close to 1.0 (1.588). According to deviance criterion, dependent variable exhibited a large over-dispersion. Thus, negative binomial regression was used. Parameter estimation values and standard errors of Poisson and negative binomial regressions were given in Table 3-4. The effect of the over-dispersion values was reflected in both regression methods. Estimates obtained for independent variables were quite different for Poisson regression and negative binomial regression, indicating the effects of over-dispersion on parameter estimations.

Models	$\mathbf{D}\mathbf{f}^*$	Devians statistics	Over-dispersion ^{**}
Poisson regression	56	3076.4920	54.937
Negative binomial regression	56	88.9479	1.588

Table 2. Compliance criteria for Poisson and negative binomial regression models

*Df= degree of freedom, **Devians statistics /Sd

The variation in the populations of Bucerotiformes and Caprimulgiformes ordo groups, according to the reference parameter of the Anseriformes ordo, were not statistically significant. However, the increase and decrease in the populations of Ciconiiformes (p < 0.05), Falconiformes, Strigiformes, Suliformes, Charadriiformes, Coraciiformes, Columbiformes, Gruiformes, Passeriformes Pelecaniformes and Podicipediformes ordo groups were statistically significant (p < 0.01) (Table 3). The population changes in spring, summer and autumn seasons, according to winter reference parameter, were statistically significant (p < 0.01). The population changes in arid land, meadow, cultivated field, irrigated field, swamp and open water surface habitats, according to the reed area reference parameter, were statistically significant (p < 0.01), whereas the difference in the pasture habitat was not significant.

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The increase in the number of species caused a 1.224-fold increase in population density and a one-unit increase in the number of UTM squares used by the species caused a 1.212-fold increase in population density and the increases were statistically significant (p<0.01) (Table 3).

Negative binomial regression results showed that population changes in 7 ordo groups were not significant according to the Anseriformes ordo reference parameter. However, the population decreases and increases in Charadriiformes, Columbiformes and Podicipediformes ordo groups (p < 0.05) and Falconiformes, Strigiformes and Gruiformes ordo groups (p < 0.01) were statistically significant (Table 4). The population changes in the spring and autumn seasons were not statistically significant (p > 0.05) according to the winter parameter, whereas summer population changes were significant. The population variations in arid area, irrigated field and open water surface were significant (p < 0.01) according to reed area the reference parameter. Population changes in meadows and marshes were found to be statistically significant (p < 0.05), however, population changes in cultivated farmland and pasture areas were not significant. The increase in the number of species and one unit increase in the number of UTM squares caused a significant increase in population density.

The over-dispersion value of the dependent variable was quite high (Table 3-4) and subsequently caused different parameter estimation values and standard errors in both regression models. The interpretation of parameter estimates for both regression models exhibits differences from that of linear regression. The Poisson regression and the negative binomial regression models should be linearized using the log link function to estimate how much each individual variable has effect on the dependent variable.

In regression models based on generalized linear models such as Poisson and Negative binomial, one level of each independent variable is typically used as reference category (Luo and Qu 2015; Yeşilova et al., 2016). Therefore, Anseriformes for ordo, winter for seasons, reeds for habitats, number of species and UTM square numbers were considered as reference levels.

The inflence of independent variables on bird populations were analysed and the results were given in Table 3 and Table 4. However, independent variables, ordo (team) (Anseriformes, Bucerotiformes, Caprimulgiformes, Charadriiformes, Ciconiiformes, Coraciiformes, Cuculiformes, Falconiformes, Gruiformes, Passeriformes, Pelecaniformes, Podicipediformes, Strigiformes, Suliformes), the number of species, the number of UTM used, seasons (summer, autumn and winter) and habitats (open water surface, swamps, reeds, arid lands, irrigated lands, cultivated lands, pastures and meadows) have different levels. Therefore, each independent variable level should be tested independently to assess the importance of individual effect on bird populations.

Since negative binomial regression was the best regression model according to the compliance criterion, only the results of binomial regression were interpreted (Table 4). Accordingly, it was found that population decreases and increases were statistically significant in Charadriiformes, Columbiformes and Podicipediformes ordo groups (p < 0.05), Falconiformes, Strigiformes and Gruiformes ordo groups (p < 0.01). One unit increase in the number of species and UTM square led to the population increase by 1.215 (p < 0.01) and 1.056 (p < 0.05) times, respectively. Population changes between seasons and habitats were statistically significant.

The population in the summer season significantly (p < 0.01) increased compared to the reference parameter of winter while the population increase and decrease in spring and autumn were not significant. The population changes only in cultivated lands and pasture areas, according to the habitat reference parameter were not significant.

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				%95 Wald	confidence	Wald Khi-square		
Parameters	Df	Estimate	Standard error	inte	rval	value	p-value	Exp.
Intercept	1	0.1372	0.1114	-0.0811	0.3555	1.52	0.2182	1.147
Bucerotiformes	1	-0.6196	0.3647	-1.3343	0.0952	2.89	0.0893	0.538
Caprimulgiformes	1	-0.1780	0.1617	-0.4949	0.1390	1.21	0.2710	0.836
Charadriiformes	1	1.4737	0.0462	1.3832	1.5642	1019.55	0.0001**	4.365
Ciconiiformes	1	-0.2142	0.1021	-0.4143	-0.0141	4.40	0.0359*	0.807
Columbiformes	1	1.7610	0.1016	1.5619	1.9601	300.47	0.0001**	5.818
Coraciiformes	1	0.7283	0.0873	0.5571	0.8994	69.56	0.0001**	2.071
Falconiformes	1	-1.1951	0.1287	-1.4474	-0.9428	86.22	0.0001**	0.302
Gruiformes	1	2.1545	0.0465	2.0634	2.2456	2150.54	0.0001**	8.623
Passeriformes	1	0.4004	0.0652	0.2726	0.5282	37.69	0.0001**	1.492
Pelecaniformes	1	0.4179	0.0786	0.2638	0.5720	28.25	0.0001**	1.518
Podicipediformes	1	1.5518	0.0517	1.4505	1.6531	901.21	0.0001**	4.719
Strigiformes	1	-2.9976	0.7112	-4.3915	-1.6037	17.77	0.0001**	0.049
Suliformes	1	-1.3981	0.2060	-1.8019	-0.9943	46.06	0.0001**	0.247
Number of species	1	0.2024	0.0056	0.1914	0.2134	1305.16	0.0001**	1.224
Spring	1	1.5288	0.0750	1.3817	1.6758	415.32	0.0001**	4.612
Summer	1	1.8864	0.0708	1.7476	2.0252	709.34	0.0001**	6.595
Autumn	1	0.4110	0.0670	0.2796	0.5424	37.60	0.0001**	1.508
Dry lands	1	1.2725	0.0601	1.1548	1.3902	448.89	0.0001**	3.569
Meadows	1	0.8751	0.0521	0.7730	0.9772	282.29	0.0001**	2.399
Intensive farmland	1	0.2807	0.0648	0.1537	0.4077	18.78	0.0001**	1.324
Arable agricultural fields	1	2.2872	0.0617	2.1663	2.4081	1375.29	0.0001**	9.847
Pastures	1	-0.1769	0.5053	-1.1674	0.8135	0.12	0.7263	0.837
Inland marshes	1	1.3669	0.0458	1.2772	1.4567	890.98	0.0001**	3.923
Surface standing waters	1	2.2464	0.0508	2.1467	2.3460	1953.32	0.0001**	9.453
UTM	1	0.1923	0.0099	0.1729	0.2118	373.91	0.0001**	1.212

Table 3. Poisson regression model parameter estimation values and standard error obtained for Haçlı Lake (sd. error)

Df= Degrees of freedom *p<0.05, **p<0.01

 Table 4. Negative binomial regression model parameter estimation values and standard error obtained for Haçlı Lake (sd. error)

			Standard			Wald Khi-squa	re	
Parameters	Df	Estimate	error	%95 Wald co	nfidence interval	value	p-value	Exp.
Intercept	1	1.4190	0.4849	0.4686	2.3694	8.56	0.0034	4.132
Bucerotiformes	1	-1.1081	1.0007	-3.0695	0.8533	1.23	0.2682	0.330
Caprimulgiformes	1	-0.5850	0.6976	-1.9523	0.7823	0.70	0.4017	0.110
Charadriiformes	1	0.8891	0.3648	0.1741	1.6040	5.94	0.0148*	2.432
Ciconiiformes	1	-0.2123	0.5728	-1.3350	0.9105	0.14	0.7110	0.808
Columbiformes	1	1.1797	0.5980	0.0076	2.3518	3.89	0.0485*	3.253
Coraciiformes	1	-0.9688	0.5865	-2.1183	0.1808	2.73	0.0986	0.379
Falconiformes	1	-1.4593	0.4608	-2.3624	-0.5562	10.03	0.0015**	0.232
Gruiformes	1	1.8858	0.4980	0.9097	2.8619	14.34	0.0002**	6.591
Passeriformes	1	-0.3444	0.4440	-1.2146	0.5257	0.60	0.4379	0.708
Pelecaniformes	1	-0.0062	0.4629	-0.9136	0.9011	0.00	0.9893	0.993
Podicipediformes	1	1.2663	0.5544	0.1796	2.3530	5.22	0.0224*	3.547
Strigiformes	1	-4.4282	1.1786	-6.7383	-2.1181	14.12	0.0002**	0.011
Suliformes	1	-0.7869	0.6762	-2.1121	0.5384	1.35	0.2445	0.455
Number of species	1	0.2352	0.0487	0.1397	0.3306	23.33	0.0001**	1.265
Spring	1	0.7717	0.4828	-0.1745	1.7179	2.56	0.1099	2.163
Summer	1	1.1831	0.4432	0.3145	2.0518	7.13	0.0076**	3.264
Autumn	1	0.0459	0.4619	-0.8594	0.9512	0.01	0.9209	1.046
Dry lands	1	2.0966	0.5029	1.1110	3.0822	17.38	0.0001**	8.138
Meadows	1	0.6568	0.3284	0.0130	1.3005	4.00	0.0455*	1.928
Intensive farmland	1	0.1628	0.4940	-0.8054	1.1310	0.11	0.7417	1.176
Arable agricultural fields	1	1.9419	0.4893	0.9829	2.9010	15.75	0.0001**	6.971
Pastures	1	-0.7302	0.9806	-2.6522	1.1918	0.55	0.4565	0.481
Inland marshes	1	0.8473	0.3306	0.1994	1.4953	6.57	0.0104*	2.333
Surface standing waters	1	1.5878	0.3588	0.8847	2.2910	19.59	0.0001**	4.892
UTM	1	0.1874	0.0839	0.0230	0.3518	4.99	0.0254*	1.206

Df= Degrees of freedom *p<0.05, **p<0.01

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CONCLUSION

The result of deviance statistics revealed a significant over-dispersion of the dependent variable. Therefore, the use of negative binomial regression was considered more appropriate to evaluate the parameters. Parameter estimates were quite different in both Poisson and negative binomial regressions.

The categorical independent variables of Haçlı Lake model are Anseriformes ordo, reeds and winter season which were considered as references. The significance levels of these independent variables on population density were given as follows. Comparision of Anseriformes ordo with the other ordos revealed that the populations of 9 ordos decreased, however the decrease in population was only statistically significant (p < 0.01) for the Falconiformes ordo. However, the population densities of the 4 ordo increased compared to the reference. The population was increased in Charadriiformes (p < 0.05), Columbiformes (p < 0.05), Podicipediformes (p < 0.05) and Gruiformes (p < 0.01) ordos.

Population differences between ordos were attributed to the contribution of Armenian gull (Larus armenicus), Rock pigeon (Columba livia), Common coot (Fulica atra) and Little Grebe (Tachybaptus *ruficollis*) which are seen in all seasons in the area. The intense presence of species such as Redshank (Tringa totanus), Common tern (Sterna hirundo), Ruddy shelduck (Tadorna ferruginea) and Lapwing (Vanellus vanellus) in the study area during the reproduction period cause the differences between populations. The differences in feeding area of species are another reason for the increase and decrease in population density. The wetlands around the lake are suitable feeding areas for the shore birds, while the coastal reeds and the lake mirror are suitable feeding areas for diving or surface feeding birds. However, in some cases, different species have been observed feeding in the same feeding areas. Therefore, similarities and differences in feeding areas either facilitated for some species to find food or made it difficult for others and caused them to leave. This resulted in an increase in population densities of some ordos and a decrease in others. Guisan et al. (2007) stated that the population status of birds differs from species to species, and population trends may change by vital activities and habitat preferences. Beerens et al. (2011) reported that seasonal and annual nutritional diversity in the reproduction period plays an active role in the population mobility of many bird species. They also stated that the bird species in extremely lively ecosystems such as wetlands have different population densities and habitat preferences, which vary according to vital activities such as finding and using food resources.

Comparison of reeds considered the reference habitat, with other habitat types, indicated that population increases in arid lands, irrigated lands and on open water surface were statistically significant at p < 0.01 level while the population increase at meadows and swamp areas were statistically significant at p < 0.05 level. Human activities around the lake are the main causes of the differences in bird populations among habitats. Reed cutting is intensively performed around the lake in spring-summer (after reproduction period) seasons. Reed cutting during the reproduction period has a negative effect on the birds living in this habitat. Reed cutting has endangered the vital activities of birds such as hiding and shelter. In this case, the birds scattered to different habitats to continue their activities such as feeding and hiding. Consequently, population density in the reeds area decreased compared to the other habitats. Similarly, Milsom et al. (2000) and Yuan et al. (2014) emphasized that human activities have negative impacts on the distribution, density and reproductive success of birds.

Comparing winter that is the reference season with other seasons, the population increase in spring and autumn seasons was not significant, however, the population increase in summer was statistically significant (p < 0.01). The population increase in the summer compared to the winter season is related to the contribution of the hatching individuals with the arrival of migratory species to the number of populations. The hatching individuals of Ruddy shelduck (*Tadorna ferruginea*), Lapwing (*Vanellus*)

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vanellus), Common tern (*Sterna hirundo*) and Redshank (*Tringa totanus*) species which nest in colonies in the area during spring season significantly increase the population density in the summer season. These species are migratory and leaving the area before the winter season, which explains the population differences between the seasons. Johnston et al. (2015) also showed that the distribution and density of birds vary depending on species and seasons.

The temperature rise during the summer season increased the living activity in the area, which had a positive effect on the population density and caused to the difference in population density between summer and winter seasons. The effect of temperature variable on population density and number of species was demonstrated by Gonçalves et al. (2017) who emphasized that climatic factors such as temperature and humidity have a significant effect on the number of species and population density.

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