

Anais da Academia Brasileira de Ciências (2016) 88(4): 2121-2136 (Annals of the Brazilian Academy of Sciences) Printed version ISSN 0001-3765 / Online version ISSN 1678-2690 http://dx.doi.org/10.1590/0001-3765201620150103 www.scielo.br/aabc

Geostatistical Approach for Spatial Interpolation of Meteorological Data

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Manuscript received on February 2, 2015; accepted for publication on March 1, 2016

ABSTRACT

Meteorological data are used in many studies, especially in planning, disaster management, water resources management, hydrology, agriculture and environment. Analyzing changes in meteorological variables is very important to understand a climate system and minimize the adverse effects of the climate changes. One of the main issues in meteorological analysis is the interpolation of spatial data. In recent years, with the developments in Geographical Information System (GIS) technology, the statistical methods have been integrated with GIS and geostatistical methods have constituted a strong alternative to deterministic methods in the interpolation and analysis of the spatial data. In this study; spatial distribution of precipitation and temperature of the Aegean Region in Turkey for years 1975, 1980, 1985, 1990, 1995, 2000, 2005 and 2010 were obtained by the Ordinary Kriging method which is one of the geostatistical interpolation methods, the changes realized in 5-year periods were determined and the results were statistically examined using cell and multivariate statistics. The results of this study show that it is necessary to pay attention to climate change in the precipitation regime of the Aegean Region. This study also demonstrates the usefulness of the geostatistical approach in meteorological studies.

Key words: geostatistical interpolation, geographic information system, ordinary kriging, meteorological data.

INTRODUCTION

Measurement and evaluation of the spatially distributed meteorological data have become important in connection with climate-change impact studies, determination of water budgets at different temporal and spatial scales, as well as validation of atmospheric and hydrological models. Meteorological data are usually available from a limited number of meteorological stations (Hofierka et al. 2002), mostly because it is not economically and technically possible to obtain meteorological data throughout the entire surface. For this reason, spatial interpolation of the meteorological variables obtained from the certain sample points is performed in order to create a model for the entire surface.

Spatial interpolation is the procedure of estimating the value of unsampled points using existing observations (Waters 1997). Methods for spatial interpolation can be classified into two main categories

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as deterministic and geostatistical (Burrough and McDonnell 1998, Matthews 2002). Deterministic interpolation techniques calculate the values of unsampled points and create surfaces from measured points, based on either the extent of similarity or the degree of smoothing (Matthews 2002). Deterministic methods do not use probability theory (Waters 1997). Geostatistical interpolation techniques use the statistical properties of the measured points, quantify the spatial autocorrelation among the measured points and account for the spatial configuration of the sample points around the estimation location (Matthews 2002).

Kriging is a geostatistical technique for optimal spatial estimation (Waller and Gotway 2004). Kriging provides a solution to the problem of estimation based on a continuous model of stochastic spatial variation and takes the variogram model (Webster and Oliver 2007). Today, with the developments in computer and Geographical Information System (GIS) technologies, the statistical methods have been integrated with GIS and the geostatistical methods have constituted a strong alternative to deterministic methods in the interpolation of the spatial data. In addition, statistical methods to analyze the interpolated layers have allowed a better understanding of the changes occurred in the specific time period.

Climate change is one of the biggest threats for the entire globe (Kropp 2015). Climate changes affect the natural balance of the earth and ecosystems and whole life is disrupted (National Academy of Sciences 2009) Climate change is most often measured by changes in primary climate variables, such as temperature and precipitation. These variables are the main drivers of climate changes (Sheffield and Wood 2012). For this reason, to understand and monitor the changes and their causes and effects accurately, changes should be determined both spatially and quantitatively and the results should be evaluated in detail.

In this study it is aimed to investigate the spatial distribution of precipitation and temperature of the Aegean Region in Turkey for years 1975, 1980, 1985, 1990, 1995, 2000, 2005 and 2010 by the Ordinary Kriging method and statistically examine the results using cell statistics and multivariate statistics to understand the changes. This study demonstrates the usefulness of the geostatistical approach for both interpolation of meteorological data and analysis and comparison of the results.

MATERIALS AND METHODS

The Aegean Region is one of Turkey's seven geographical regions. It is surrounded by the Aegean Sea on the west and takes its name from the Aegean Sea (Ozcaglar 2014). In this study, the area comprising eight provinces located in the Aegean Region has been analyzed. The total area is approximately 90,000 km² (Figure 1). The coastal areas of the Aegean Region has a Mediterranean climate. The effects of the Mediterranean climate extend up to 100-150 km inland from the coast. In coastal areas, winters are mild and summers are very hot and dry. The interior side of the region is affected by the continental climate (Sensoy et al. 2008).

In the present study, the time series of monthly precipitation and temperature data from 98 meteorological stations for the years 1975, 1980, 1985, 1990, 1995, 2000, 2005 and 2010 were used. Spatial distributions of the stations are shown in Figure 1. The geospatial interpolation of temperature and precipitation data and all statistical analyses of the precipitation and temperature layers were performed using ArcGIS 10.0 software (Esri, Redlands, CA). The method of creating an estimation surface layer with the Ordinary Kriging is explained in Section "Creating An Estimation Surface Layer with the Ordinary Kriging" and statistical analyses of layers is presented in Section "Statistical Analyses of Layers".



Figure 1 - The location of the study area (The Aegean Region, Turkey) and spatial distributions of the meteorological stations.

CREATING AN ESTIMATION SURFACE LAYER WITH THE ORDINARY KRIGING

Estimation with the Kriging interpolation method has a two-step process: *(i) fitting a model:* creation of the variograms and covariance functions to estimate the statistical dependence (spatial autocorrelation) values that depend on the model of autocorrelation and *(ii) making an estimation:* estimation of the unknown values (ESRI 2014a).

The first step in the Ordinary Kriging is to create a semivariogram from the scatter point set to be interpolated. A semivariogram consists of *(i) an empirical semivariogram* (experimental variogram) and *(ii) a model semivariogram* (GMS User Manuel 2012). Semivariogram is a mathematical model of the semivariance as a function of *lag* and displays the statistical correlation of nearby points (Prasad et al. 2007). Spatial autocorrelation (means feature similarity) is based on both feature locations and feature values simultaneously (not only based on feature locations or attribute values alone). Given a set of features and an associated attribute, it evaluates whether the pattern expressed is clustered, dispersed, or random (Matthews 2002). Empirical semivariogram, computed by (Eq.1) for all pairs of locations separated by distance h (ESRI 2014a):

The formula involves calculating the difference squared between the values of the paired locations. Figure 2 shows the pairing of one point (the red point) with all other measured locations. This process continues for each measured point (ESRI 2014a).

Often, each pair of locations has a unique distance, and there are often many pairs of points. To plot all pairs quickly becomes unmanageable. Instead of plotting each pair, the pairs are grouped into *lag bins*. The empirical semivariogram is a graph of the averaged semivariogram values on the y-axis and the distance (or *lag*) on the x-axis (Figure 3) (ESRI 2014a).

When two locations are close to each other (far left on the x-axis of the semivariogram cloud), then they are expected to be similar (low on the y-axis of the semivariogram cloud) (ESRI 2014a, Prasad et al. 2007). "As pairs of locations become farther apart (moving to the right on the x-axis of the semivariogram cloud), they should become more dissimilar and have a higher squared difference (moving up on the y-axis of the semivariogram cloud)" (ESRI 2014a).

Once the empirical variogram is obtained, the next step is to define a model semivariogram (GMS User Manuel 2012). Semivariogram modeling is a main step between spatial description and spatial estimation. The empirical semivariogram provides information on the spatial autocorrelation of datasets, however does not supply information for all possible directions and distances. For this reason, it is necessary to fit



Figure 2 - Calculation of the difference squared between the paired locations.



Figure 3 - Empirical semivariogram graph example.

a model (a continuous function or curve) to the empirical semivariogram (ESRI 2014a). There are many semivariogram models. Some of the most common are linear, circular, spherical, exponential, and Gaussian model (Li and Heap 2008). The selected model influences the estimation of the unknown values and each model is designed to fit different types of phenomena more accurately (ESRI 2014a).

Once the model variogram is obtained, it is used to calculate the weights used in Kriging (GMS User Manuel 2012). The basic equation used in the Ordinary Kriging is as (Eq.2) (ESRI 2014a, GMS User Manuel 2012, Borga and Vizzaccaro 1996):

$$\hat{Z}(s_0) = \sum_{i=1}^{N} \lambda_i Z(s_i)$$
⁽²⁾

Where;

 $Z(s_i)$: the measured value at the *i*th location

 λ_i : an unknown weight for the measured value at the *i*th location

 (s_0) : the estimation location

N: the number of measured values

With Kriging method, the value $Z(s_0)$ at the point s_0 , where the true unknown value is $Z(s_0)$, is estimated by a linear combination of the values at N surrounding data points (Borga and Vizzaccaro 1996).

In the Ordinary Kriging, the weight, λ_i , depends on a fitted model to the measured points, the distance to the estimation point, and the spatial relationships among the measured values around the estimation location (ESRI 2014a) and the Kriging weights are calculated by minimizing the variance (Li and Heap 2008). The Ordinary Kriging is the most widely used Kriging method (Wackernagel 2003) and this method assumes that the data set has a stationary variance but also a non-stationary mean value within the search radius. The Ordinary Kriging is highly reliable and recommended for most data sets (Vertical Mapper User Guide 2008).

STATISTICAL ANALYSES OF LAYERS

CELL STATISTICS

In a local function, the value at each location on the output raster is a function of the input values at that location. When computing a local function, input rasters can be combined and a statistic can be calculated. In ArcGIS software, several cell statistics can be calculated for raster layers: *(i) MEAN:* Calculates the mean (average) of the inputs, *(ii) MAXIMUM:* Determines the maximum (largest value) of the inputs, *(iii) MEDIAN:* Calculates the median of the inputs, *(iv) MINIMUM:* Determines the minimum (smallest value) of the inputs, *(v) RANGE:* Calculates the range (difference between largest and smallest value) of the inputs, *(vi) STD:* Calculates the standard deviation of the inputs (ESRI 2014b).

Multivariate Statistics

The multivariate statistics allow exploration of relationships between many different data layers or types of attributes. In band collection function, main statistical measures (minimum, maximum, mean and standard deviation) can be calculated for every layer and in addition to these standard statistics, the covariance and correlation matrices can also be determined (ESRI 2014b).

RESULTS AND DISCUSSION

The time series of monthly precipitation and temperature data for the years 1975, 1980, 1985, 1990, 1995, 2000, 2005 and 2010 were used for preparing spatial distribution layers of precipitation and temperature of the Aegean Region, Turkey. The Ordinary Kriging interpolation was applied for each month and a total of 192 interpolations were performed (96 for precipitation and 96 for temperature) and grid layers with 250-meter pixel size were formed. The Ordinary Kriging interpolation results of the precipitation and temperature data for January are shown in Figures 4 and 5, respectively.

Based on the multivariate statistics (band collection), spatial analyses were applied for the monthly precipitation and temperature layer series which calculated by the Ordinary Kriging. Table I (for precipitation) and Table II (for temperature) represent the main statistics, including the minimum, maximum, mean and standard deviation values. In addition, the correlation coefficients were calculated with these analyses (Tables III and IV).

When examining Table I, it was seen that the highest "average precipitation" and the highest "precipitation" values were in December 1990. Table II shows that both highest "average temperature" and highest "temperature" values were in August 2010.

According to Table III, correlation coefficients for precipitation are between -0.04847 and 0.92382 for January, 0.18609 and 0.92908 for February, -0.57255 and 0.74793 for March, -0.43776 and 0.76531 for April, -0.62850 and 0.86128 for May, -0.26385 and 0.79177 for June, -0.22994 and 0.74375 for July, 0.01302 and 0.80480 for August, -0.20455 and 0.69922 for September, -0.68396 and 0.74704 for October,



Figure 4 - The Ordinary Kriging interpolation results of the precipitation data for the month of January (For the years of 1975, 1980, 1985, 1990, 1995, 2000, 2005, and 2010).



Figure 5 - The Ordinary Kriging interpolation results of the temperature data for the month of January (For the years of 1975, 1980, 1985, 1990, 1995, 2000, 2005, and 2010).

0.29150 and 0.85486 for November, 0.22520 and 0.90808 for December. According to Table IV, correlation coefficients for temperature are between 0.95152 and 0.99524 for January, 0.95804 and 0.99591 for February, 0.91016 and 0.99459 for March, 0.94823 and 0.99361 for April, 0.96744 and 0.99414 for May, 0.94961 and 0.99084 for June, 0.93391 and 0.99041 for July, 0.94175 and 0.99293 for August, 0.96714 and 0.99254 for September, 0.97214 and 0.99542 for October, 0.93409 and 0.99328 for November, 0.96537 and 0.99665 for December.

Correlations above 0.80 generally are accepted as high correlations. Correlations between 0.50 and 0.80 are usually considered as medium (moderate) correlations and correlations below 0.50 are typically regarded as low correlations (Wang et al. 1990). Accordingly, very high correlation values were observed between temperature values of years 1975, 1980, 1985, 1990, 1995, 2000, 2005 and 2010 for all months (Table IV). But, the correlations between layers of precipitation were examined, both high and low correlation values were observed. For precipitation layers, the highest correlation was observed between the year of 1975 and 2010 for January; 2005 and 2010 for February; 1985 and 1995 for March; 2005 and 2010 for April; 1990 and 2000 for May; 1990 and 1995 for June; 1975 and 2010 for July; 1975 and 2000 for August; 1980 and 2000 for September; 2000 and 2010 for October; 2005 and 2010 for November; 1980 and 1990 for December.

By calculating cell statistics, a statistic for each cell in an output raster can be calculated based on the values of multiple input rasters (ESRI 2014b). In this study; maximum, minimum, mean, median, range and standard deviation layers were produced by using precipitation and temperature layers for years 1975, 1980, 1985, 1990, 1995, 2000, 2005 and 2010 for all months. Totally 144 statistical layers were obtained

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	STATISTICS of INDIVIDUAL LAYERS - PRECIPITATION											
Layer(Yea	r)	MIN	MAX	MEAN	STD	Layer(Year)	MIN	MAX	MEAN	STD	
January	1975	15.0962	290.5551	105.2915	68.9066	July	1975	0.0000	11.4688	3.8647	3.2307	
·	1980	57.4809	301.0310	144.8527	41.7206	·	1980	0.0000	6.4559	1.5206	0.9994	
	1985	46.7396	325.7432	160.8042	58.6881		1985	0.0000	7.6697	0.7246	0.7361	
	1990	0.0000	35.0969	12.1264	5.1998		1990	0.0000	39.4809	5.0383	5.5698	
	1995	18.9700	272.5263	120.9055	59.4265		1995	0.0000	189.5152	34.9476	30.6981	
	2000	42.8687	132.2835	69.6099	16.4874		2000	0.0000	23.2685	3.9946	4.4414	
	2005	14.4151	305.8113	70.4621	46.7614		2005	0.0000	85.0155	24.6473	19.6237	
	2010	34.0842	249.2391	114.5032	48.2253		2010	0.0173	12.8247	5.7179	3.1505	
February	1975	35.9389	164.3231	58.1565	23.4616	August	1975	0.0000	66.0855	12.1734	13.2959	
	1980	11.4949	100.1111	35.0220	14.2885		1980	0.0000	29.5959	4.3619	5.3946	
	1985	29.6399	149.1992	73.9867	26.1003		1985	0.0000	41.5747	10.0517	7.4810	
	1990	7.0911	143.2509	49.7590	25.1979		1990	0.4801	67.0008	10.8518	7.5046	
	1995	6.6114	117.6134	27.3189	15.0847		1995	0.2385	16.9660	7.1374	3.8129	
	2000	49.3334	109.7159	86.5541	15.9437		2000	0.0000	26.2174	6.2557	5.7277	
	2005	18.2838	330.9904	117.0124	68.6871		2005	0.4208	17.9898	8.6525	4.3138	
	2010	47.3964	343.1137	151.4875	62.9992		2010	0.0000	20.6179	7.7861	5.9152	
March	1975	26.2257	107.8506	64.0406	15.5155	September	1975	0.0000	49.7015	10.9055	10.0935	
	1980	49.5541	159.2850	95.6635	22.8128		1980	0.0000	55.4421	16.4988	14.7347	
	1985	27.8837	106.6610	59.4646	17.9716		1985	0.0000	6.5261	1.9362	1.3202	
	1990	6.9016	39.8629	23.6062	5.8286		1990	1.4236	38.1317	20.5715	8.0680	
	1995	61.6728	222.2994	126.3867	33.2200		1995	0.0000	46.1658	17.5078	11.7651	
	2000	38.4527	143.0037	88.8729	20.7716		2000	0.0000	22.9868	5.9550	5.8907	
	2005	33.6981	117.8575	79.2205	15.7518		2005	4.7207	35.6864	15.2393	6.0027	
	2010	2.0541	66.2231	30.9691	13.4006		2010	22.0556	44.4905	33.0731	4.0500	
April	1975	18.2819	98.7922	51.6425	12.2515	October	1975	9.5328	90.6642	33.0401	13.0100	
	1980	29.4372	69.5153	47.0104	7.3046		1980	4.3897	46.2036	22.3987	9.1381	
	1985	2.0838	44.1912	19.7078	9.1412		1985	11.1300	151.5788	42.3043	26.9946	
	1990	19.9863	90.5741	54.8812	12.3692		1990	7.2120	42.6564	24.6231	8.0837	
	1995	18.1657	89.0869	51.9969	14.4514		1995	6.0607	56.4756	34.5488	14.9560	
	2000	26.9695	152.4105	85.1713	26.0013		2000	3.6411	98.6216	34.3385	19.8806	
	2005	15.9945	90.1445	39.7732	13.9404		2005	11.2323	121.7506	33.1998	17.9017	
	2010	9.5546	83.4186	37.2400	14.8342		2010	49.5620	311.1042	116.5370	49.2287	
May	1975	39.9216	119.2508	71.7452	17.9338	November	1975	39.6952	239.7979	112.4898	38.7931	
	1980	10.6646	77.5801	36.4267	10.0239		1980	31.0222	174.3634	93.1593	26.5030	
	1985	1.4462	92.5336	33.3793	9.2920		1985	28.2310	180.9006	85.0800	32.2572	
	1990	0.0000	62.5613	23.3644	13.0269		1990	13.8738	74.1966	32.1743	11.6332	
	1995	0.0000	76.9429	27.1503	9.6845		1995	31.6046	177.6392	88.1167	27.7271	
	2000	1.0087	79.3443	31.0651	20.8717		2000	1.8513	201.2959	54.5331	44.2754	
	2005	6.2719	83.2165	44.0559	17.0640		2005	43.2833	254.8676	114.4541	41.6780	
	2010	12.1662	46.3693	26.6590	6.6865		2010	8.2926	51.3080	28.8157	9.8915	
June	1975	15.4860	100.2484	50.0390	17.3824	December	1975	39.7735	188.7916	88.9672	25.1946	
	1980	0.0000	80.8975	24.9794	12.4476		1980	28.2021	296.8829	147.5694	61.1975	
	1985	0.0000	45.2143	16.5010	11.0090		1985	19.9052	135.5423	51.3575	16.5321	
	1990	0.0000	54.2057	20.5578	14.1597		1990	43.9064	361.6299	165.4436	71.0610	
	1995	0.0021	47.5019	9.5297	10.6071		1995	24.6129	165.1980	71.9208	30.6091	
	2000	0.0000	37.4081	11.7682	8.3772		2000	14.4990	329.9915	52.2345	36.1115	
	2005	0.3222	58.3159	27.5195	12.8268		2005	11.9330	203.2546	65.3737	31.7432	
	2010	0.7233	141 3197	55 6312	24 9145		2010	59 0785	231 4372	119 4989	34 7257	

TABLE I

Main statistics for precipitation layers (Minimum, Maximum, Mean, Standard deviation).

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Main statistics for temperature layers (Minimum, Maximum, Mean, Standard deviation).

STATISTICS of INDIVIDUAL LAYERS - TEMPERATURE											
Layer(Yea	r)	MIN	MAX	MEAN	STD	Layer(Year)		MIN	MAX	MEAN	STD
January	1975	-1.6080	11.9795	3.7326	3.1861	July	1975	20.8040	28.7429	25.1050	2.1311
	1980	-2.6259	11.4972	3.0345	3.4320		1980	21.6467	29.0423	25.5491	1.7273
	1985	2.2792	13.7292	7.0922	2.9161		1985	18.8785	28.5773	24.1568	2.6798
	1990	-4.3366	11.9610	2.5856	3.4689		1990	20.9858	29.3868	25.8605	2.2016
	1995	1.8665	12.9328	6.2080	2.7793		1995	19.5084	29.1838	24.6306	2.9346
	2000	-5.0773	9.6267	1.1922	3.7585		2000	22.6168	30.2668	27.1807	2.0244
	2005	2.2003	12.6267	6.3250	2.6454		2005	21.7852	29.3529	26.1972	2.1397
	2010	1.5386	13.2897	6.6969	2.7832		2010	21.6157	30.2688	26.5241	2.1923
February	1975	-0.4336	11.9043	4.1513	3.0501	August	1975	20.1038	27.2521	23.9362	1.9830
	1980	-0.3540	11.8185	4.6182	2.9815		1980	20.3963	28.3603	24.5688	2.0282
	1985	-4.1206	10.0607	1.7981	3.7145		1985	21.4755	28.3776	25.3309	1.8210
	1990	0.0275	12.6155	5.6012	3.1362		1990	19.6488	27.6955	24.3883	2.2268
	1995	3.9147	13.8931	7.8196	2.5229		1995	20.3723	28.2541	24.5760	2.0307
	2000	-1.8539	11.9216	4.6025	3.3023		2000	20.0756	28.9870	25.3378	2.4911
	2005	1.0839	11.0291	5.1357	2.7220		2005	22.1073	28.9807	26.0388	1.9475
	2010	4.8857	14.0768	8.8109	2.4160		2010	25.2166	30.3014	28.1310	1.5564
March	1975	6.6072	15.0204	10.1282	2.2135	September	1975	16.5458	26.1924	20.8361	2.7081
	1980	3.7778	13.6001	7.4684	2.5244		1980	14.3647	24.4542	19.1395	2.5351
	1985	1.5080	13.6020	7.9737	2.9179		1985	15.2840	25.1916	20.5854	2.4419
	1990	5.8160	14.7953	9.8410	2.4662		1990	15.4788	25.2687	19.8176	2.4584
	1995	4.9791	14.0615	8.4624	2.4898		1995	17.0612	25.4172	21.0791	2.1735
	2000	3.4803	12.7911	7.1545	2.6171		2000	16.3220	25.8142	21.1166	2.4484
	2005	5.0723	13.9870	8.8848	2.4185		2005	16.4752	25.7482	20.7710	2.5786
	2010	5.8425	16.4715	10.4517	2.3320		2010	18.3427	25.9256	21.7630	2.1259
April	1975	11.4027	17.6734	13.9034	1.7218	October	1975	11.0174	21.0976	14.8899	2.6129
	1980	8.4474	15.4275	11.4915	2.0839		1980	12.3536	21.1985	16.2249	2.3637
	1985	10.4556	17.2842	13.9654	2.0125		1985	8.7044	19.1918	13.0156	2.5467
	1990	9.9327	17.4434	13.1470	2.2247		1990	11.7813	21.7325	16.0564	2.4525
	1995	8.2844	15.9269	11.5358	2.2356		1995	9.2743	19.8509	13.8540	2.5798
	2000	11.7082	17.5848	14.3388	1.8857		2000	10.5590	20.9450	14.9599	2.6980
	2005	9.9162	17.1927	13.1155	2.1300		2005	9.7082	20.0835	13.8896	2.6384
	2010	9.9424	18.6159	13.8762	2.4406		2010	10.2576	21.7663	14.9154	2.9379
May	1975	13.7103	20.7852	17.0399	2.3309	November	1975	5.2976	16.6203	9.3378	2.6591
	1980	14.1786	20.2122	16.8566	1.7605		1980	7.4070	18.1094	11.4330	2.6440
	1985	15.8079	22.5850	19.2854	2.1122		1985	8.4017	18.0259	12.0326	2.3988
	1990	15.3384	21.6440	1/.3523	2.50/4		1990	/.61/9	18.0775	12.0642	2.6051
	1995	15.4864	21.1/62	18.3836	1.6126		1995	1.5022	13.8388	6.//09	3.0380
	2000	13.92//	22.0098	18.0709	2.5511		2000	/.1828	18.5339	11.8420	2.50/5
	2005	14.9138	21.6488	18.4084	2.1///		2005	5.1913	15./904	9.106/	2.5619
T	2010	14.5255	22.5254	19.3349	2.2301	Develop	2010	9.6838	19.5951	13.9344	2.3800
June	19/5	1/./122	24.6564	21.2/20	2.2485	December	19/5	-3.2/8/	13.1689	3.8009	3.83/0
	1980	18.1393	25.9512	21./894	2.0404		1980	2.1515	13.9134	0./185	2.9002
	1900	18.0790	20.3428	22.1907	2.3044		1900	2.0959	14.5184	0.0195	3.0085
	1990	1/.50/9	23.9///	21.8186	2.0/33		1990	2.4093	14.2393	7.3080	3.04// 2.1052
	2000	19.0111	21.1813	24.1438	2.4040		2000	1.004/	14./038	1.1/49	3.1032
	2000	10.0133	21.3/08	23.1208	2.0309		2000	3 0240	17.4294	0.3028	2 8025
	2003	12.00/3	23.9203	22.0338	2.4329		2003	5 2610	15.0/30	0.0007	2.0023
	2010	10.0412	20.3030	22.1002	2.4//9		2010	5.2019	13.0/38	7.007/	2.0403

TABLE III											
	Cor	relation mat	rix for precip	oitation layers	5.						
ON MATRIX (P	recipitation-J	anuary)									
1975	1980	1985	1990	1995							

CORRELATION	CORRELATION MATRIX (Precipitation-January)											
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010				
1975	1.00000	0.78846	0.78386	0.12940	0.87703	0.75682	0.77189	0.92382				
1980	0.78846	1.00000	0.86062	0.17125	0.81056	0.69908	0.65659	0.77735				
1985	0.78386	0.86062	1.00000	0.16711	0.69942	0.63916	0.60556	0.81675				
1990	0.12940	0.17125	0.16711	1.00000	-0.04847	0.39700	0.01441	0.06176				
1995	0.87703	0.81056	0.69942	-0.04847	1.00000	0.55791	0.78129	0.86215				
2000	0.75682	0.69908	0.63916	0.39700	0.55791	1.00000	0.61256	0.74714				
2005	0.77189	0.65659	0.60556	0.01441	0.78129	0.61256	1.00000	0.81718				
2010	0.92382	0.77735	0.81675	0.06176	0.86215	0.74714	0.81718	1.00000				
CORRELATION	MATRIX (Precipitation-F	ebruary)									
Laver(Year)	1975	1980	1985	1990	1995	2000	2005	2010				
1975	1.00000	0.77978	0.71037	0.75465	0.78511	0.32408	0.62331	0.59694				
1980	0.77978	1.00000	0.70680	0.65260	0.75107	0.18609	0.51814	0.44678				
1985	0.71037	0.70680	1.00000	0.67623	0.59647	0.25324	0.33989	0.25617				
1990	0.75465	0.65260	0.67623	1.00000	0.79150	0.64538	0.842.64	0.77847				
1995	0.78511	0.75107	0 59647	0 79150	1 00000	0.50060	0.71130	0.68328				
2000	0 32408	0 18609	0 25324	0.64538	0.50060	1 00000	0 53958	0.66191				
2005	0.62331	0 51814	0 33989	0.84264	0.71130	0 53958	1 00000	0.92908				
2010	0 59694	0.44678	0.25617	0.77847	0.68328	0.66191	0.92908	1 00000				
2010 0.59094 0.44078 0.25017 0.77847 0.08528 0.00191 0.92908 1.00000 CORRELATION MATRIX (Precipitation-March)												
Laver(Year)	1975	1980	1985	1990	1995	2000	2005	2010				
1975	1.00000	0.43577	0.43600	0.54000	0.42494	0.60580	0.65125	0.24777				
1980	0.43577	1.00000	0.48775	0.59721	0.63492	0.69036	0.10031	-0.57255				
1985	0.43600	0.48775	1.00000	0.24778	0.74793	0.23000	-0.00553	-0.38343				
1990	0.54000	0.59721	0.24778	1.00000	0.34694	0.57528	0.44113	0.00851				
1995	0.42494	0.63492	0.74793	0.34694	1.00000	0.51069	-0.00559	-0.43359				
2000	0.60580	0.69036	0.23000	0.57528	0.51069	1.00000	0.39025	-0.20299				
2005	0.65125	0.10031	-0.00553	0.44113	-0.00559	0.39025	1.00000	0.61318				
2010	0.24777	-0.57255	-0.38343	0.00851	-0.43359	-0.20299	0.61318	1.00000				
CORRELATION	MATRIX (Precipitation-A	(pril)									
Laver(Year)	1975	1980	1985	1990	1995	2000	2005	2010				
1975	1.00000	0.04649	-0.29210	0.22152	0.08140	-0.09371	-0.43776	-0.29386				
1980	0.04649	1.00000	0.20565	0.08084	0.25487	-0.06780	0.46926	0.44613				
1985	-0.29210	0.20565	1.00000	0.11292	0.31988	0.64021	0.74893	0.63354				
1990	0.22152	0.08084	0.11292	1.00000	0.53602	0.27626	-0.03302	-0.19927				
1995	0.08140	0.25487	0.31988	0.53602	1.00000	0.62172	0.26567	-0.06769				
2000	-0.09371	-0.06780	0.64021	0.27626	0.62172	1.00000	0.33247	0.06834				
2005	-0.43776	0.46926	0.74893	-0.03302	0.26567	0.33247	1.00000	0.76531				
2010	-0.29386	0.44613	0.63354	-0.19927	-0.06769	0.06834	0.76531	1.00000				
CORRELATION	MATRIX (Precipitation-N	/lay)									
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010				
1975	1.00000	0.00014	0.51378	0.40567	0.04118	0.58152	-0.03261	-0.35965				
1980	0.00014	1.00000	-0.14016	0.09716	0.11290	0.16423	-0.14988	-0.11414				
1985	0.51378	-0.14016	1.00000	-0.22438	-0.12253	-0.13165	-0.22834	0.13157				
1990	0.40567	0.09716	-0.22438	1.00000	0.28300	0.86128	0.64427	-0.51083				
1995	0.04118	0.11290	-0.12253	0.28300	1.00000	0.19170	0.21629	0.00804				
2000	0.58152	0.16423	-0.13165	0.86128	0.19170	1.00000	0.29651	-0.62850				
2005	-0.03261	-0.14988	-0.22834	0.64427	0.21629	0.29651	1.00000	0.05753				
2010	-0.35965	-0.11414	0.13157	-0.51083	0.00804	-0.62850	0.05753	1.00000				

				(,					
CORRELATIC	ON MATRIX (I	Precipitation-J	une)							
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1.00000	0.20265	0.20593	0.28976	0.32791	0.31104	0.10884	-0.22344		
1980	0.20265	1.00000	-0.25895	0.25642	-0.01494	-0.26385	0.30535	0.13046		
1985	0.20593	-0.25895	1.00000	0.61296	0.63628	0.78552	0.42126	0.32951		
1990	0.28976	0.25642	0.61296	1.00000	0.79177	0.60613	0.66090	0.48602		
1995	0.32791	-0.01494	0.63628	0.79177	1.00000	0.76352	0.57111	0.30044		
2000	0.31104	-0.26385	0.78552	0.60613	0.76352	1.00000	0.27430	0.25948		
2005	0.10884	0.30535	0.42126	0.66090	0.57111	0.27430	1.00000	0.48441		
2010	-0.22344	0.13046	0.32951	0.48602	0.30044	0.25948	0.48441	1.00000		
CORRELATIO	ON MATRIX (I	Precipitation-J	uly)							
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1.00000	0.22850	-0.22994	0.71580	0.46294	0.37854	0.62662	0.74375		
1980	0.22850	1.00000	0.20738	0.21785	0.23528	0.29285	0.10786	0.54332		
1985	-0.22994	0.20738	1.00000	0.11996	0.04063	0.20946	0.10074	0.05492		
1990	0.71580	0.21785	0.11996	1.00000	0.29884	0.37246	0.57299	0.65655		
1995	0.46294	0.23528	0.04063	0.29884	1.00000	0.04565	0.39870	0.56466		
2000	0.37854	0.29285	0.20946	0.37246	0.04565	1.00000	0.55639	0.27101		
2005	0.62662	0.10786	0.10074	0.57299	0.39870	0.55639	1.00000	0.40386		
2010	0.74375	0.54332	0.05492	0.65655	0.56466	0.27101	0.40386	1.00000		
CORRELATION MATRIX (Precipitation-August)										
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1.00000	0.62665	0.45484	0.33267	0.52327	0.80480	0.47424	0.60626		
1980	0.62665	1.00000	0.63235	0.23099	0.34192	0.67653	0.49201	0.68952		
1985	0.45484	0.63235	1.00000	0.27238	0.49517	0.47897	0.36196	0.71590		
1990	0.33267	0.23099	0.27238	1.00000	0.11132	0.34003	0.01302	0.18870		
1995	0.52327	0.34192	0.49517	0.11132	1.00000	0.35925	0.67713	0.59460		
2000	0.80480	0.67653	0.47897	0.34003	0.35925	1.00000	0.51379	0.66336		
2005	0.47424	0.49201	0.36196	0.01302	0.67713	0.51379	1.00000	0.55531		
2010	0.60626	0.68952	0.71590	0.18870	0.59460	0.66336	0.55531	1.00000		
CORRELATIO	ON MATRIX (I	Precipitation-S	September)							
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1.00000	0.37163	0.19678	0.59756	0.53491	0.04033	0.02862	0.39440		
1980	0.37163	1.00000	0.57914	0.66755	-0.08048	0.69922	0.38872	0.20890		
1985	0.19678	0.57914	1.00000	0.36662	-0.10389	0.58103	0.36137	0.23253		
1990	0.59756	0.66755	0.36662	1.00000	0.37907	0.41984	0.03516	0.24139		
1995	0.53491	-0.08048	-0.10389	0.37907	1.00000	-0.20455	-0.19451	0.65751		
2000	0.04033	0.69922	0.58103	0.41984	-0.20455	1.00000	0.52987	0.17506		
2005	0.02862	0.38872	0.36137	0.03516	-0.19451	0.52987	1.00000	0.31895		
2010	0.39440	0.20890	0.23253	0.24139	0.65751	0.17506	0.31895	1.00000		
CORRELATIC	ON MATRIX (I	Precipitation-(October)							
Laver(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1 00000	0.06983	0.66618	-0 34307	-0.21368	0 26494	0.62937	0 25635		
1980	0.06983	1 00000	0.49524	0.12709	0.66087	-0.57686	0.02957	-0.60944		
1085	0.66618	0.40524	1 00000	0.08372	0.13023	-0.37080	0.61832	0.06324		
1000	0.00010	0.12700	0.00272	1 00000	0.13923	0.01497	0.01032	-0.00324		
1990	-0.3430/	0.12/09	-0.00372	0.50000	1.00000	-0.10393	-0.209/8	-0.04311		
2000	-0.21368	0.57(0)	0.13923	0.50000	1.00000	-0.08396	0.00114	-0.03589		
2000	0.26494	-0.5/686	0.01497	-0.16595	-0.68396	1.00000	-0.10283	0./4/04		
2005	0.62937	0.44764	0.61832	-0.20978	0.06114	-0.10283	1.00000	-0.27208		
2010	0.25635	-0.60944	-0.06324	-0.04511	-0.63589	0.74704	-0.27208	1.00000		

 TABLE III (continuation)

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CORRELATION	CORRELATION MATRIX (Precipitation-November)											
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010				
1975	1.00000	0.79762	0.65602	0.45331	0.41620	0.62822	0.79794	0.72336				
1980	0.79762	1.00000	0.77541	0.57684	0.59180	0.52085	0.72144	0.68886				
1985	0.65602	0.77541	1.00000	0.63039	0.56208	0.64761	0.72962	0.73200				
1990	0.45331	0.57684	0.63039	1.00000	0.34262	0.47725	0.34958	0.29150				
1995	0.41620	0.59180	0.56208	0.34262	1.00000	0.45066	0.43236	0.34430				
2000	0.62822	0.52085	0.64761	0.47725	0.45066	1.00000	0.64668	0.55346				
2005	0.79794	0.72144	0.72962	0.34958	0.43236	0.64668	1.00000	0.85486				
2010	0.72336	0.68886	0.73200	0.29150	0.34430	0.55346	0.85486	1.00000				
CORRELATION MATRIX (Precipitation-December)												
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010				
1975	1.00000	0.79379	0.25259	0.81574	0.78905	0.50197	0.86966	0.64845				
1980	0.79379	1.00000	0.31704	0.90808	0.90569	0.39818	0.78649	0.80757				
1985	0.25259	0.31704	1.00000	0.22520	0.48945	0.76838	0.31490	0.26742				
1990	0.81574	0.90808	0.22520	1.00000	0.85097	0.42111	0.74077	0.78907				
1995	0.78905	0.90569	0.48945	0.85097	1.00000	0.56697	0.77087	0.77156				
2000	0.50197	0.39818	0.76838	0.42111	0.56697	1.00000	0.52801	0.38538				
2005	0.86966	0.78649	0.31490	0.74077	0.77087	0.52801	1.00000	0.65563				
2010	0.64845	0.80757	0.26742	0.78907	0.77156	0.38538	0.65563	1.00000				

TABLE III (continuation)

 TABLE IV

 Correlation matrix for temperature layers.

CORRELATION MATRIX (Temperature-January)											
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010			
1975	1.00000	0.99001	0.98543	0.99064	0.98531	0.99384	0.99098	0.97266			
1980	0.99001	1.00000	0.99524	0.98171	0.99200	0.99217	0.99463	0.97417			
1985	0.98543	0.99524	1.00000	0.97166	0.99482	0.98869	0.99503	0.97539			
1990	0.99064	0.98171	0.97166	1.00000	0.97136	0.98582	0.97629	0.95152			
1995	0.98531	0.99200	0.99482	0.97136	1.00000	0.98977	0.99353	0.97975			
2000	0.99384	0.99217	0.98869	0.98582	0.98977	1.00000	0.99041	0.97014			
2005	0.99098	0.99463	0.99503	0.97629	0.99353	0.99041	1.00000	0.98280			
2010	0.97266	0.97417	0.97539	0.95152	0.97975	0.97014	0.98280	1.00000			
CORRELATION MATRIX (Temperature-February)											
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010			
1975	1.00000	0.99420	0.99281	0.98748	0.98275	0.96240	0.98388	0.97305			
1980	0.99420	1.00000	0.99591	0.99286	0.98342	0.97515	0.98280	0.96708			
1985	0.99281	0.99591	1.00000	0.99162	0.98447	0.97700	0.98697	0.97381			
1990	0.98748	0.99286	0.99162	1.00000	0.98965	0.98378	0.98806	0.97410			
1995	0.98275	0.98342	0.98447	0.98965	1.00000	0.98594	0.98194	0.98254			
2000	0.96240	0.97515	0.97700	0.98378	0.98594	1.00000	0.96256	0.95804			
2005	0.98388	0.98280	0.98697	0.98806	0.98194	0.96256	1.00000	0.98685			
2010	0.97305	0.96708	0.97381	0.97410	0.98254	0.95804	0.98685	1.00000			
CORRELATION	MATRIX (T	emperature-M	larch)								
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010			
1975	1.00000	0.96776	0.93293	0.96541	0.98026	0.97802	0.96322	0.91016			
1980	0.96776	1.00000	0.97682	0.97817	0.98798	0.98687	0.99252	0.96439			
1985	0.93293	0.97682	1.00000	0.97683	0.95402	0.95607	0.96738	0.93709			
1990	0.96541	0.97817	0.97683	1.00000	0.97132	0.97380	0.97428	0.93204			
1995	0.98026	0.98798	0.95402	0.97132	1.00000	0.99459	0.98740	0.93968			
2000	0.97802	0.98687	0.95607	0.97380	0.99459	1.00000	0.98191	0.93228			
2005	0.96322	0.99252	0.96738	0.97428	0.98740	0.98191	1.00000	0.96909			
2010	0.91016	0.96439	0.93709	0.93204	0.93968	0.93228	0.96909	1.00000			

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CORRELATION MATRIX (Temperature-April)										
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1.00000	0.97750	0.97998	0.97378	0.97661	0.96197	0.96881	0.94823		
1980	0.97750	1.00000	0.98631	0.98860	0.99361	0.96976	0.98274	0.97421		
1985	0.97998	0.98631	1.00000	0.98679	0.98503	0.97772	0.98419	0.96170		
1990	0.97378	0.98860	0.98679	1.00000	0.99253	0.98757	0.99332	0.98398		
1995	0.97661	0.99361	0.98503	0.99253	1.00000	0.97755	0.98891	0.97271		
2000	0.96197	0.96976	0.97772	0.98757	0.97755	1.00000	0.98612	0.96112		
2005	0.96881	0.98274	0.98419	0.99332	0.98891	0.98612	1.00000	0.97864		
2010	0.94823	0.97421	0.96170	0.98398	0.97271	0.96112	0.97864	1.00000		
CORRELATION MATRIX (Temperature-May)										
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1.00000	0.98754	0.99057	0.97503	0.96946	0.98393	0.98151	0.97436		
1980	0.98754	1.00000	0.99004	0.97957	0.98300	0.98179	0.98569	0.97928		
1985	0.99057	0.99004	1.00000	0.97169	0.98487	0.97795	0.97774	0.97943		
1990	0.97503	0.97957	0.97169	1.00000	0.97585	0.99414	0.98976	0.96744		
1995	0.96946	0.98300	0.98487	0.97585	1.00000	0.97536	0.97691	0.97110		
2000	0.98393	0.98179	0.97795	0.99414	0.97536	1.00000	0.98905	0.97046		
2005	0.98151	0.98569	0.97774	0.98976	0.97691	0.98905	1.00000	0.96891		
2010	0.97436	0.97928	0.97943	0.96744	0.97110	0.97046	0.96891	1.00000		
CORRELATION	N MATRIX (T	emperature-I		0.20711	0.97110	0.57010	0.90091	1.00000		
Laver(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1.00000	0.95346	0.98094	0.97460	0.97601	0.97138	0.96753	0.97943		
1980	0.95346	1 00000	0.98204	0.98186	0.95395	0.97538	0.98243	0 94961		
1985	0.98094	0.98204	1 00000	0.99084	0.97684	0.99073	0.99018	0.97364		
1900	0.97460	0.98186	0.99084	1.00000	0.98152	0.98635	0.98838	0.97215		
1995	0.97601	0.95395	0.97684	0.98152	1 00000	0.96700	0.96597	0.95207		
2000	0.97138	0.97538	0.99073	0.98635	0.96700	1.00000	0.98993	0.96894		
2000	0.96753	0.98243	0.99018	0.98838	0.96597	0.98993	1 00000	0.96989		
2003	0.97943	0.94961	0.97364	0.97215	0.95207	0.96894	0.96989	1.00000		
CORRELATION	MATRIX (T	emperature_I	11v)	0.97215	0.75207	0.70074	0.90909	1.00000		
Lavor(Voar)	1975	1080	1085	1990	1005	2000	2005	2010		
1075	1973	0.07647	0 08807	0.990/1	0.07678	0.98756	0.08871	0 07028		
1975	0.07647	1,00000	0.96587	0.97782	0.93301	0.96840	0.96744	0.97928		
1985	0.98897	0.96587	1 00000	0.98248	0.93591	0.90340	0.98407	0.93022		
1905	0.900/1	0.90387	0.08248	1 00000	0.95961	0.97507	0.98969	0.97413		
1995	0.99041	0.97782	0.98248	0.05061	1 00000	0.96200	0.98009	0.97109		
2000	0.97078	0.95591	0.96309	0.93901	0.06200	1.00000	0.98130	0.90555		
2000	0.98730	0.90840	0.97307	0.98519	0.90299	0.08577	1 00000	0.90078		
2003	0.98871	0.90744	0.96407	0.98009	0.98130	0.96577	0.06754	1.00000		
	0.9/920	0.93022	0.9/413	0.97109	0.90333	0.90078	0.90734	1.00000		
	1075		1095	1000	1005	2000	2005	2010		
Layer(Year)	1 00000	0.09710	0.09710	0.09092	0.02110	0.09917	0.09214	2010		
1973	0.00710	1.00000	0.98/10	0.20202	0.20112	0.9001/	0.76214	0.90300		
1005	0.98/10	0.00525	1.00000	0.98/10	0.9/92/	0.98//4	0.9/024	0.53483		
1905	0.98/10	0.98525	1.00000	1.00000	0.98400	0.99000	0.98430	0.93388		
1990	0.98982	0.98/10	0.98983	0.00000	0.98274	0.99293	0.98298	0.90/03		
1995	0.98119	0.9/92/	0.98460	0.98274	1.00000	0.98340	0.98//0	0.941/3		
2000	0.9881/	0.98//4	0.99000	0.99293	0.98340	1.00000	1.00000	0.9/248		
2005	0.98214	0.9/624	0.98430	0.98298	0.98//0	0.99239	1.00000	1.00000		
2010	0.90380	0.93483	0.93388	0.90/03	0.941/3	0.9/248	0.90030	1.00000		

 TABLE IV (continuation)

CORRELATION MATRIX (Temperature-September)										
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1.00000	0.98432	0.98128	0.98609	0.97975	0.98318	0.98589	0.99125		
1980	0.98432	1.00000	0.99155	0.98712	0.98378	0.98371	0.97888	0.97036		
1985	0.98128	0.99155	1.00000	0.98127	0.97791	0.98163	0.97309	0.96714		
1990	0.98609	0.98712	0.98127	1.00000	0.98041	0.97355	0.97007	0.98024		
1995	0.97975	0.98378	0.97791	0.98041	1.00000	0.99027	0.98815	0.96733		
2000	0.98318	0.98371	0.98163	0.97355	0.99027	1.00000	0.99254	0.97132		
2005	0.98589	0.97888	0.97309	0.97007	0.98815	0.99254	1.00000	0.97438		
2010	0.99125	0.97036	0.96714	0.98024	0.96733	0.97132	0.97438	1.00000		
CORRELATION MATRIX (Temperature-October)										
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1.00000	0.98886	0.99071	0.98472	0.99004	0.99062	0.99542	0.98544		
1980	0.98886	1.00000	0.97972	0.98269	0.98005	0.98243	0.98307	0.97214		
1985	0.99071	0.97972	1.00000	0.98881	0.99341	0.99234	0.98544	0.97534		
1990	0.98472	0.98269	0.98881	1.00000	0.98608	0.98709	0.97632	0.97474		
1995	0.99004	0.98005	0.99341	0.98608	1.00000	0.99196	0.98567	0.97342		
2000	0.99062	0.98243	0.99234	0.98709	0.99196	1.00000	0.98698	0.98191		
2005	0.99542	0.98307	0.98544	0.97632	0.98567	0.98698	1.00000	0.98408		
2010	0.98544	0.97214	0.97534	0.97474	0.97342	0.98191	0.98408	1.00000		
CORRELATION MATRIX (Temperature-November)										
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1.00000	0.98838	0.99165	0.97041	0.98973	0.96630	0.99328	0.94459		
1980	0.98838	1.00000	0.98876	0.98825	0.97959	0.97029	0.98830	0.96251		
1985	0.99165	0.98876	1.00000	0.97865	0.98934	0.97822	0.98729	0.95602		
1990	0.97041	0.98825	0.97865	1.00000	0.95958	0.97309	0.97731	0.97502		
1995	0.98973	0.97959	0.98934	0.95958	1.00000	0.96471	0.98812	0.93409		
2000	0.96630	0.97029	0.97822	0.97309	0.96471	1.00000	0.96795	0.96403		
2005	0.99328	0.98830	0.98729	0.97731	0.98812	0.96795	1.00000	0.95475		
2010	0.94459	0.96251	0.95602	0.97502	0.93409	0.96403	0.95475	1.00000		
CORRELATION	MATRIX (Te	emperature-D	ecember)							
Layer(Year)	1975	1980	1985	1990	1995	2000	2005	2010		
1975	1.00000	0.98203	0.98128	0.98008	0.98588	0.98097	0.96599	0.96537		
1980	0.98203	1.00000	0.99205	0.99665	0.98918	0.99443	0.99171	0.97762		
1985	0.98128	0.99205	1.00000	0.98926	0.98818	0.99459	0.98883	0.98108		
1990	0.98008	0.99665	0.98926	1.00000	0.98819	0.99248	0.99313	0.98019		
1995	0.98588	0.98918	0.98818	0.98819	1.00000	0.98769	0.98406	0.98327		
2000	0.98097	0.99443	0.99459	0.99248	0.98769	1.00000	0.98867	0.97993		
2005	0.96599	0.99171	0.98883	0.99313	0.98406	0.98867	1.00000	0.98158		
2010	0.96537	0.97762	0.98108	0.98019	0.98327	0.97993	0.98158	1.00000		

TABLE IV (continuation)

(72 for precipitation and 72 for temperature). Figures 6 and 7 show the cell statistics of the precipitation and temperature for the month of January.

Here, range layers give the most important information. Range layers indicate the difference between the largest and the smallest value of the inputs. When examining the range layers, it was understood that the month with largest changes is January for precipitation and November for temperature.



Figure 6 - Cell statistics of the precipitation for the month of January.



Figure 7 - Cell statistics of the temperature for the month of January.

CONCLUSIONS

Meteorological data are required in many fields such as environment, agriculture and management of natural disasters where spatial data are used. But meteorological data are generally available from a limited number of stations. For this reason, interpolation techniques are used to obtain complete surface information. In recent years, depending on the technological developments in computer and GIS, geostatistical methods are used in order to determine the spatial distribution of meteorological data and the Ordinary Kriging method is nowadays a preferable option in the literature. Unlike the deterministic methods, geostatistical interpolation techniques also utilize the statistical properties of the measured points. In geostatistical techniques the autocorrelation among the measured points is determined and spatial configuration of the sampling points around the estimation point is taken into consideration.

In this study, spatial distributions of precipitation and temperature of the Aegean Region in Turkey for years 1975, 1980, 1985, 1990, 1995, 2000, 2005 and 2010 in 5-year periods were determined by the Ordinary

Kriging method. The time series of monthly precipitation and temperature data from 98 meteorological stations were used for the Ordinary Kriging. To evaluate and interpret the results, multivariate statistics (band collection) and cell statistics were applied for the monthly precipitation and temperature layer series. The results revealed that a significant change in precipitation regime in the Aegean Region was occurred. It is necessary to pay attention to this change because of multiple environmental effects of the climate changes. In the following studies, prediction of the future trends and determination of the effects of these changes on nature and human health are required.

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