

Spatio-temporal analysis of diurnal air temperature parameterization in Weather Stations over Iran

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Received: 28 December 2017; Received in revised form: 14 March 2018; Accepted: 25 April 2018

Abstract

Diurnal air temperature modeling is a beneficial experimental and mathematical approach which can be used in many fields related to Geosciences. The modeling and spatio-temporal analysis of air Diurnal Temperature Cycle (DTC) was conducted using data obtained from 105 synoptic stations in Iran during the years 2013-2014 for the first time; the key variable for controlling the cosine term in DTC modeling known as β was analyzed and considered both as monthly and annual parameter. The effect of environmental variables of humidity, pressure, diurnal air temperature range, and wind speed were analyzed on β . The results showed that there is no significant difference between considering β as monthly (dynamic) or annual (constant) parameter through the year. The RMSE of approach with dynamic β was 2.1 °C and with constant 2.2 °C at 95% percent of whole data in all stations. The analysis of environmental variables showed that humidity had an indirect effect on β . Low pressure areas showed higher β values but high pressure areas showed higher variability in β and lower mean values. In areas with high air diurnal temperature range, lower β values with less standard deviation were observed. High wind areas showed positive effect on β values.

Keywords: Air temperature; DTC model; Humidity; Pressure, Wind, Spatio-temporal

1. Introduction

Air temperature is an important variable in many fields like climate (Hansen and Lebedeff, 1987; Jones *et al.*, 1999), biology (Peng *et al.*, 2004; Tanja *et al.*, 2003), agriculture (Walthall *et al.*, 2013), hydrosphere (Haas *et al.*, 2013; Lofgren, 2014), atmosphere (Philandras *et al.*, 2015), health (Lowen *et al.*, 2007) and biosphere (De Kauwe *et al.*, 2017). The spatio-temporal behaviors of air temperature depend on many environmental factors like wind speed, soil moisture and surface roughness (Geiger *et al.*, 2009; Prihodko and Goward, 1997; Prince

et al., 1998). According to (Benali *et al.*, 2012), solar radiation as the main energy influx is controlled by many factors like latitude, cloud cover, and particle matter. The influence of surface energy budget components control the air temperature cycles (Stensrud, 2009). In most of climate studies daily mean temperature is used which cannot explain the complex behavior of diurnal air temperature. In fact, mean temperature is mainly affected by maximum, and minimum diurnal temperatures (Sun *et al.*, 2006). Qu *et al.* (2014) studied the diurnal temperature range variations of continental United States. Yue and Hashino (2003) analyzed the annual, seasonal, and monthly mean temperatures in Japan during 1900-1996. Tabari and Talaei (2011) investigated the trends of annual, seasonal, and monthly maximum and minimum air

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temperature time series in western half of Iran. Türkeş and Sümer (2004) studied the spatio-temporal behavior of maximum and minimum temperatures and diurnal temperature range in Turkey. Gadgil and Dhorde (2005) analyzed air temperature changes in Pune city of India. The relation of environmental variables with air temperature were investigated in some studies (Aikawa *et al.*, 2008; Bonnardot *et al.*, 2017; Ephrath *et al.*, 1996; Nojarov, 2014; Péré *et al.*, 2014; Xia *et al.*, 2016; Zhang *et al.*, 2014).

Study of air temperature in continuous temporal space is possible by energy-budget and empirical models. Energy budget models require a large number of input parameters. Empirical models like DTC models are sinusoidal models which only require maximum and minimum temperature and some parameters as input (Parton and Logan, 1981). These parameters are calculated based on latitude, day of year and maximum and minimum air temperatures. Leuning *et al.* (1995) proposed a Diurnal Temperature Cycle (DTC) for air temperature. One of these parameters in this model, known as β in this research, controls the half with cosine term in DTC model. In some researches (Leuning *et al.*, 1995; Parton and Logan, 1981; Stisen *et al.*, 2007), this parameter was considered as a constant parameter derived experimentally. No methodology was proposed

for calculation of these parameter and its relation to environmental variables in these studies.

The main purposes of this study were: 1) applying a DTC model for air temperature over the study area for the first time, 2) calculating and analyzing the β value in this area, and 3) analyzing the relationships between environmental variables and β . The outcomes of present study can pave the way for boosting modeling temporal observations of air temperature in the study area. As result more accurate results can be achieved from climate and meteorological researches.

2. Materials and Methods

2.1. Study area

Iran, located between 25° - 40° N and 45° - 60°E, has diverse climate regimes due to wide latitude, Alborze and Zagros mountain chains in north and west, vast deserts (the Kavir and Lut Deserts). North of Iran is limited by Caspian Sea and south of Iran is constrained by Persian Gulf and Oman Sea. According to (De Martonne, 1926, the main climate zones include: arid, semi-arid, Mediterranean, Semi-arid, humid, very humid, and extremely humid.

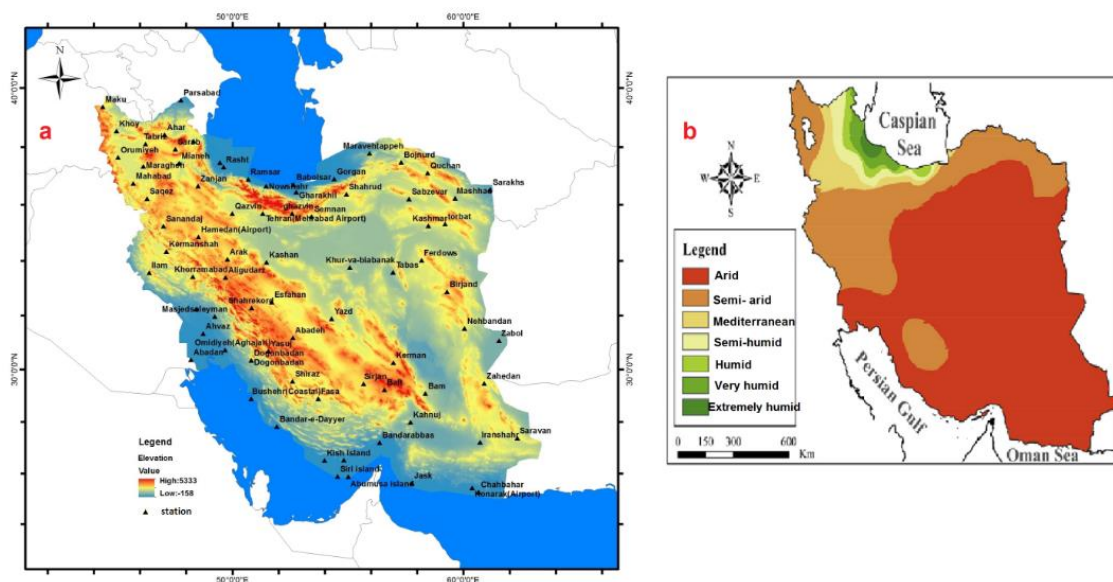


Fig. 1. a) Topographic map and weather stations of the study area. b) the climate classification of Iran based on De Martonne climate type (1966- 2005) (Tabari *et al.*, 2014)

2.2. Data and Methodology

Data including Air temperature, relative humidity, and pressure in three hourly intervals were obtained from 105 synoptic stations of

Islamic Republic of Iran Meteorology Organization (IRIMO) for the studied area during 2013 to 2014. Figure (1) shows the spatial distribution of weather stations. The acquired data were checked by constructed time

series and missed data were excluded from dataset. The applied air DTC model in this

$$T_{a\text{-day}}(t) = T_{\min} + (T_{\max} - T_{\min}) \cos\left(\frac{\pi}{\omega_{\text{air}}}(t - t_m)\right) \quad (1)$$

$$T_{a\text{-nigth}}(t) = T_{\min} + \left[(T_{\max} - T_{\min}) \cos\left(\frac{\pi}{\omega_{\text{air}}}(t_s - t_m)\right) \right] e^{-\frac{(t-t_s)}{k}} \quad t > t_s \quad (2)$$

$$\omega_{\text{air}} = \omega + 2\beta \quad (3)$$

$$k = \frac{\omega_{\text{air}}}{\pi} \tan^{-1}\left(\frac{\pi}{\omega_{\text{air}}}(t_s - t_m)\right) \quad (4)$$

Where, T_a is the air temperature ($^{\circ}\text{C}$) at time t ; T_{\min} and T_{\max} are two variables of minimum and maximum diurnal air temperatures; t_m is the time when maximum diurnal air temperature occur, t_s is the turning point of DTC model, ω is day duration, and ω_{air} is modified day duration indicating the width over the half period of the cosine term. β is a coefficient which was used to tune the period of cosine term with air temperature changes. In order to calculate β , the values of 105 synoptic stations the proposed methodology in (Gholamnia *et al.*, 2017) was applied and the range of β was defined between 0 and 10.

Day duration was calculated by (Duffie and Beckman, 1980):

$$\omega = \frac{2}{15} \arccos(-\tan\phi \tan\delta) \quad (5)$$

ϕ is latitude and δ is solar declination dependent on day of year (DOY) based on ELAGIB* *et al.* (1999):

$$\delta = 23.45 \sin\left(\frac{360}{365}(284 + \text{DOY})\right) \quad (6)$$

The behavior and sensitivity of β to environmental variables of humidity, diurnal temperature range, pressure and wind speed were analyzed; in the first step the spatial behavior of β values for 12 months were shown over study area by using Kriging interpolation method. After that variation bars (mean values with standard deviation) were applied to show the behavior of β in contrast with each environmental variable in each month. Finally,

study is based on (Gholamnia *et al.*, 2017):

the monthly average of environmental variables was plotted with β values to analyze the general trends in one year period for all weather stations in the study area and their effects on β were discussed.

3. Results

3.1. Estimation of β in the study area

The estimated β values were shown in figure (2) for each month in the study area. In January the maximum β was 2.6 in North West of Iran (Figure 2a). In February (Figure 2b), β was decreased in North West but in other areas it had slight increase. In March (Figure 2 C), β had slight decrease except for south east of the study area. In April (Figure 2d), β increased significantly specially at south part of Iran with 9.2. The spatial extent of high β values in south, south west, south east and central part of Iran was increased and reached maximum in May, June and July (Figure 2 e, f, and g). In July the values of β were decreased up to December (Figure 2 h, i, j, k, and l).

3.2. The accuracy of air temperature modeling

For modeling of air DTC, β was considered both as constant parameter through the year and as a dynamic variable which may changes every month. The accuracy of both modeling approaches are shown in Figure (3) for all weather stations in the study area. Figure (3a and b) shows the RMSE, kurtosis, and skewness of air temperature modeling with fix and monthly dynamic β , respectively. Both histograms do not show significant differences. The dynamic β approach just had better performance (0.1 $^{\circ}\text{C}$ error in air DTC model) in 95% percent of data.

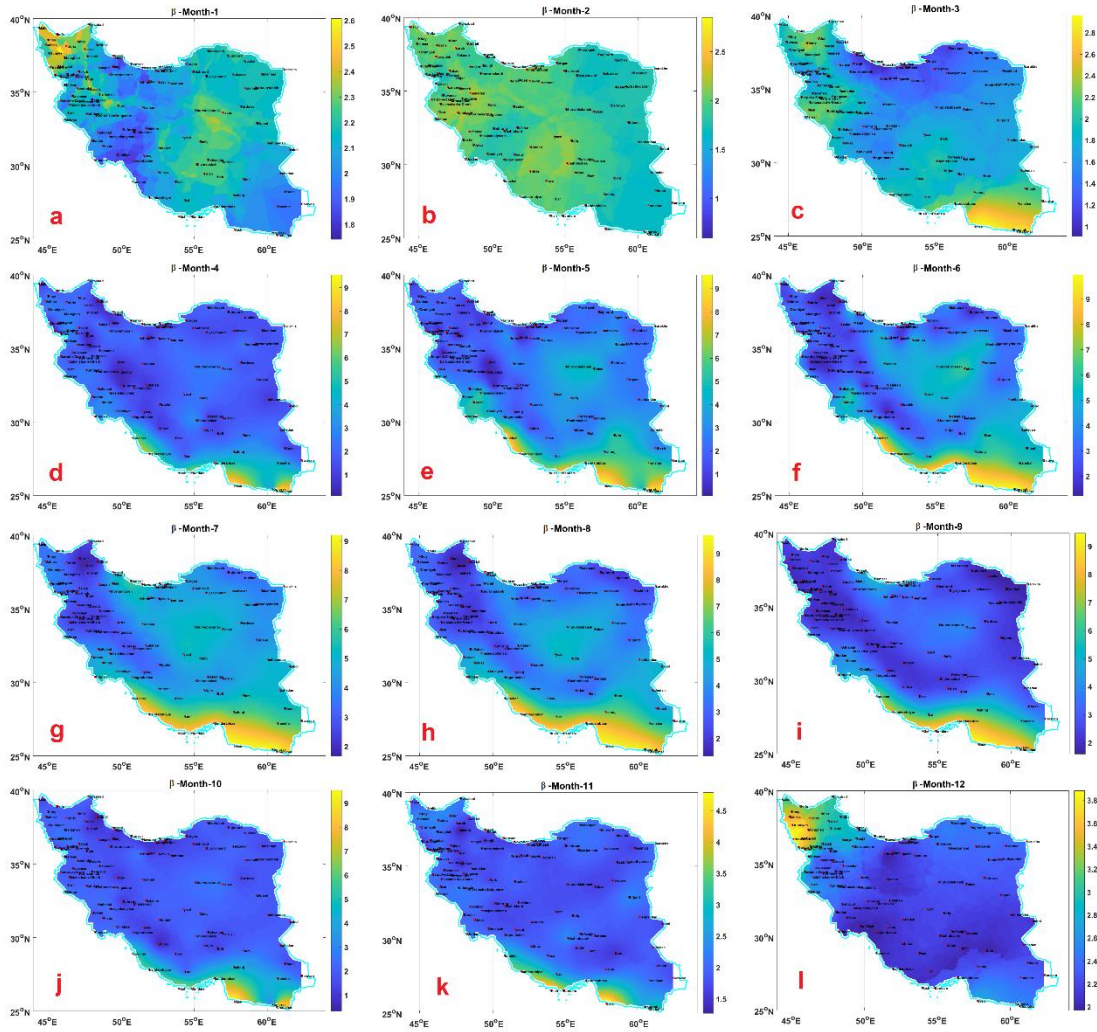


Fig. 2. The spatio-temporal behavior of β in the study area; a) January, b) February, c) March, d) April, e) May, f) June, g) July, h) August, i) September, j) October, k) November, and l) December

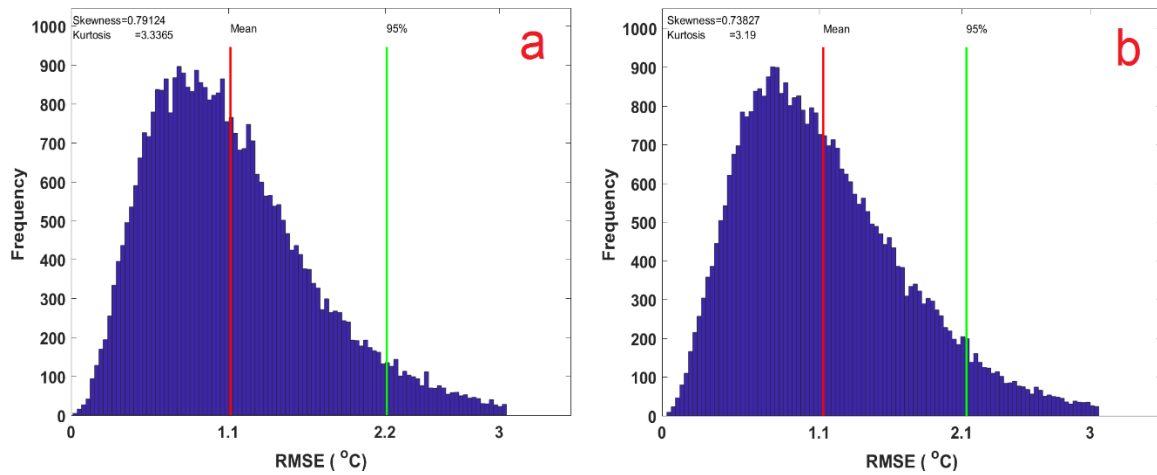


Fig. 3. Histogram of RMSE of air DTC modeling along with kurtosis and skewness. a) considers β as a constant throughout the year, b) considers β as dynamic variable which may change each month

3.3. Annual analysis of environmental variables and their relation to β

The humidity pattern in the study area was shown in Figure (4) for each month. In January (figure 4 a) in west, south west, north and north east of Iran, the relative humidity is more than 60 %. In south part (near coast stations) the humidity was almost 55%. The central part, east and south east had relative humidity less than 55%. In February, March, April, May, June, July, August, September, and October (Figure 4 b, c, d, e, f, g, h, i, and j) the humidity in north of Iran was more than other parts (more than 60%). The humidity in May was increased slightly in northwest and west of Iran which can be attributed to the activity of convective

systems during this month of the year (Alijani and Harman, 1985). In south part in near coast stations, the relative humidity was higher in warm periods of year due to the effect of high irradiation and sea breeze. In November and December, the relative humidity was increased in Northwest, west, and north east again.

The annual trend showed that in central and eastern parts of Iran, the relative humidity was decreased in warm period of year. The temporal variations in north of Iran had the lowest variation of humidity (minimum humidity range was 4% in Jirandeh in north of Iran). In most of weather stations the humidity was decreased in warm seasons. West of the study area showed high humidity variations (maximum variation was 56.67% in Qasreshirin).

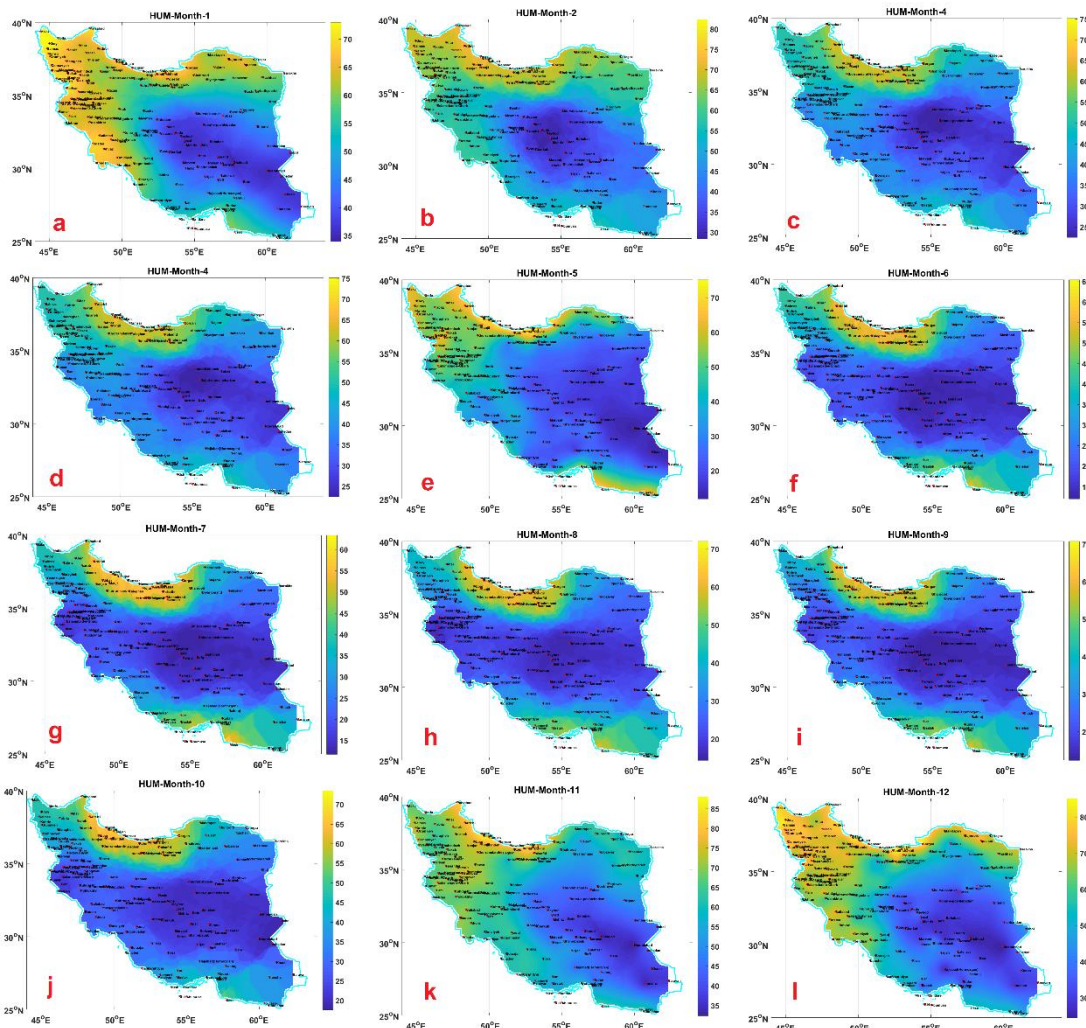


Fig. 4. The spatio-temporal behavior of humidity through a year in the study area; a) January, b) February) March, d) April, e) May, f) June, g) July, h) August, i) September, j) October, k) November, and l) December

Figure (5) shows the variation bars of mean humidity and its standard deviation against β values for every month (from January to

December). The humidity values were divided to four classes based on the annual maximum observed range. In January, mean β changed

between 1.9 and 2.3 (Figure 5 a). In February and March, mean β values had slight decrease below 2 (Figure 5 b and c). The pattern of mean β variations in these three months are almost the same; in class three of humidity a slight decrease is observed. After that, the mean β range was increased and in April, except for class two of humidity, mean β had an almost increasing trend (Figure 5d). In May (Figure 5 e), β had a decreasing trend up to class three of humidity and then increased. In June the range of β range was increased and the maximum value (5.8) was observed at fourth level of humidity (Figure 5 f). In July (Figure 5 g), the β range still grew and except for group two it has an increasing trend. In August (Figure 5 h), the

same trend as July was almost observed and the β range was decreased slightly. In September (Figure 5 i) the β reached its maximum value (6.3) in the third group of humidity. In October (Figure 5 j), the range of β decreased suddenly in group three of humidity and has the maximum value of 3.4. In November (Figure 5 k) the curve of plot is reverse in contrast with other plots and has the maximum value in the second class of humidity. In December (Figure 5 l) the trend of plot has an increasing trend from first to fourth class of humidity. The general analysis of all variation bars showed no specific trend for humidity and mean β . It is clear that mean β had an increasing trend after March and then decreased after October.

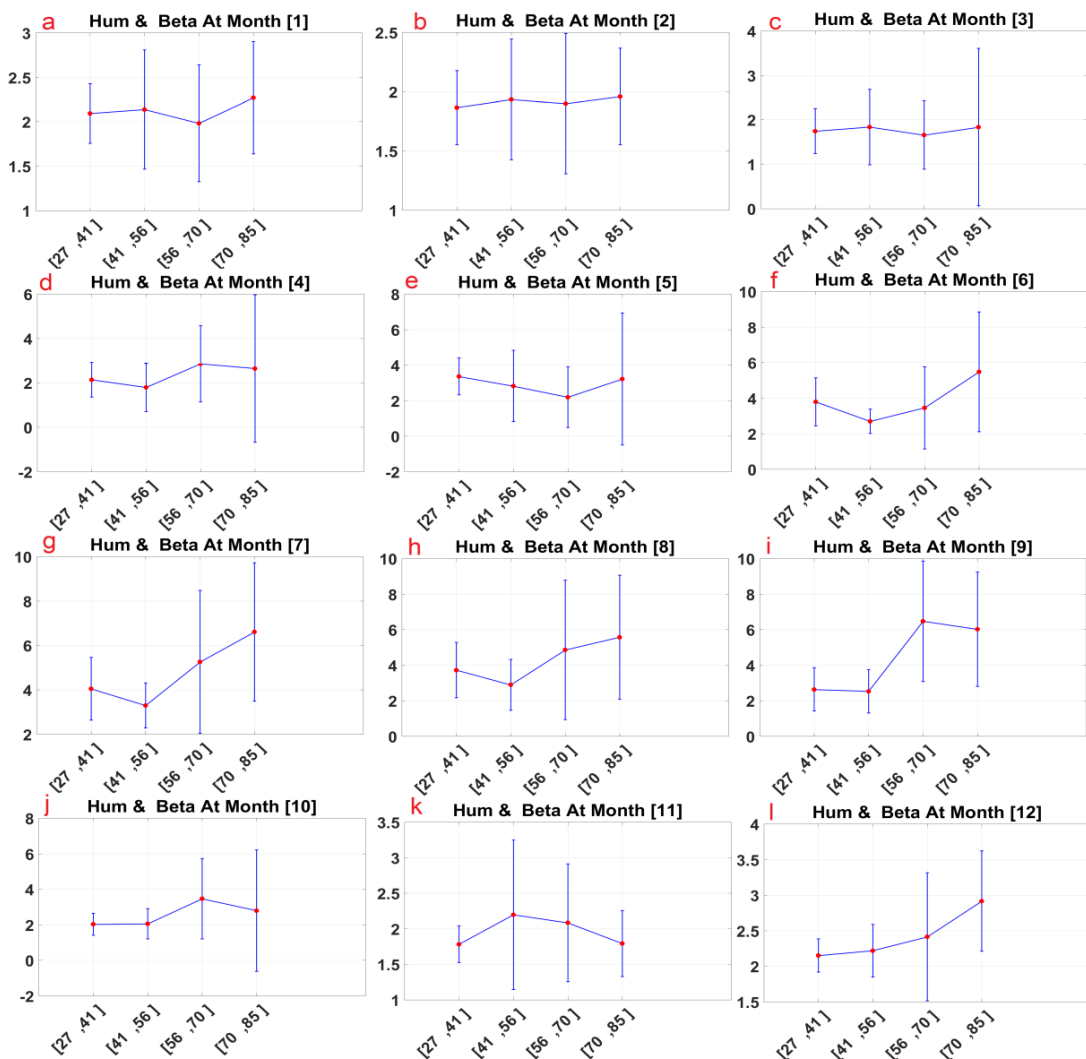


Fig. 5. Variation bars of mean and standard deviation of relative humidity against different β classes; a) January, b) February) March, d) April, e) May, f) June, g) July, h) August, i) September, j) October, k) November, and l) December

In figure (6) the monthly patterns of pressure in the study area were shown. The results (Figure 6 a, b, c, d, e, f, g, h, I, j, k, and l)

showed that most of weather stations had small pressure variations during a year. In North, south west, south, and south east parts (near

coast) which has low elevation almost 20 m AMSL had highest pressures in the study area (more than 960 hPa). In mountainous regions, west and north west, the pressure was lower in

contrast with other parts. One should note that since no weather station is present in Alborz Mountains in north of Iran, the pressure values at this part were not interpolated properly.

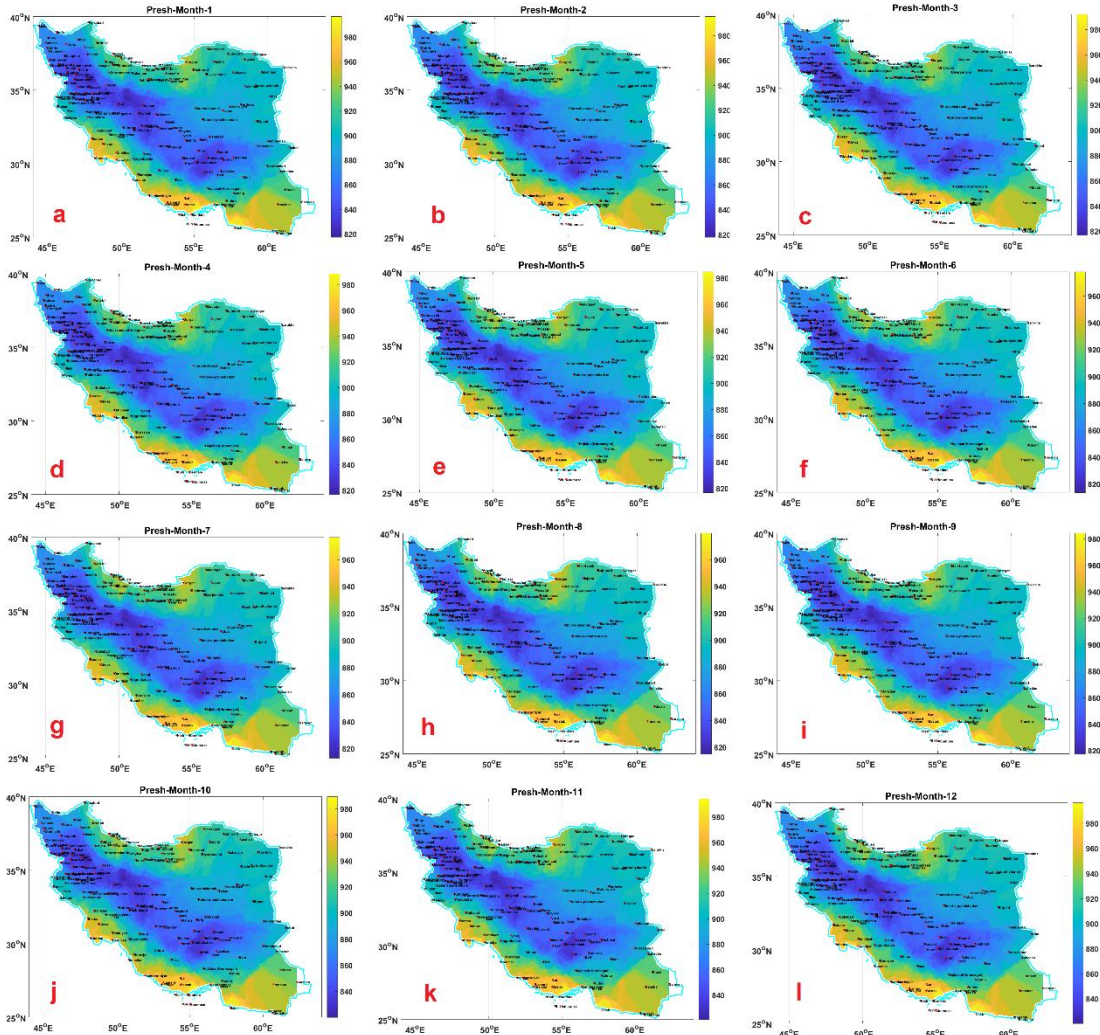


Fig. 6. The spatio-temporal behavior of pressure in the study area; a) January, b) February, c) March, d) April, e) May, f) June, g) July, h) August, i) September, j) October, k) November, and l) December

The variation bars of mean β and mean pressure for each month is shown in figure (7). Based on annual maximum observed pressure range in weather stations, four classes were defined. In January, February, and March (Figure 7 a, b and c), almost the same pattern

was observed. After that from April to November β showed an increasing trend from the first to fourth class of pressure (Figure 7 d, e, f, g, h, i, j, and k). In December (Figure 7 l) the trend of plot is almost like the bars belonged to January to March.

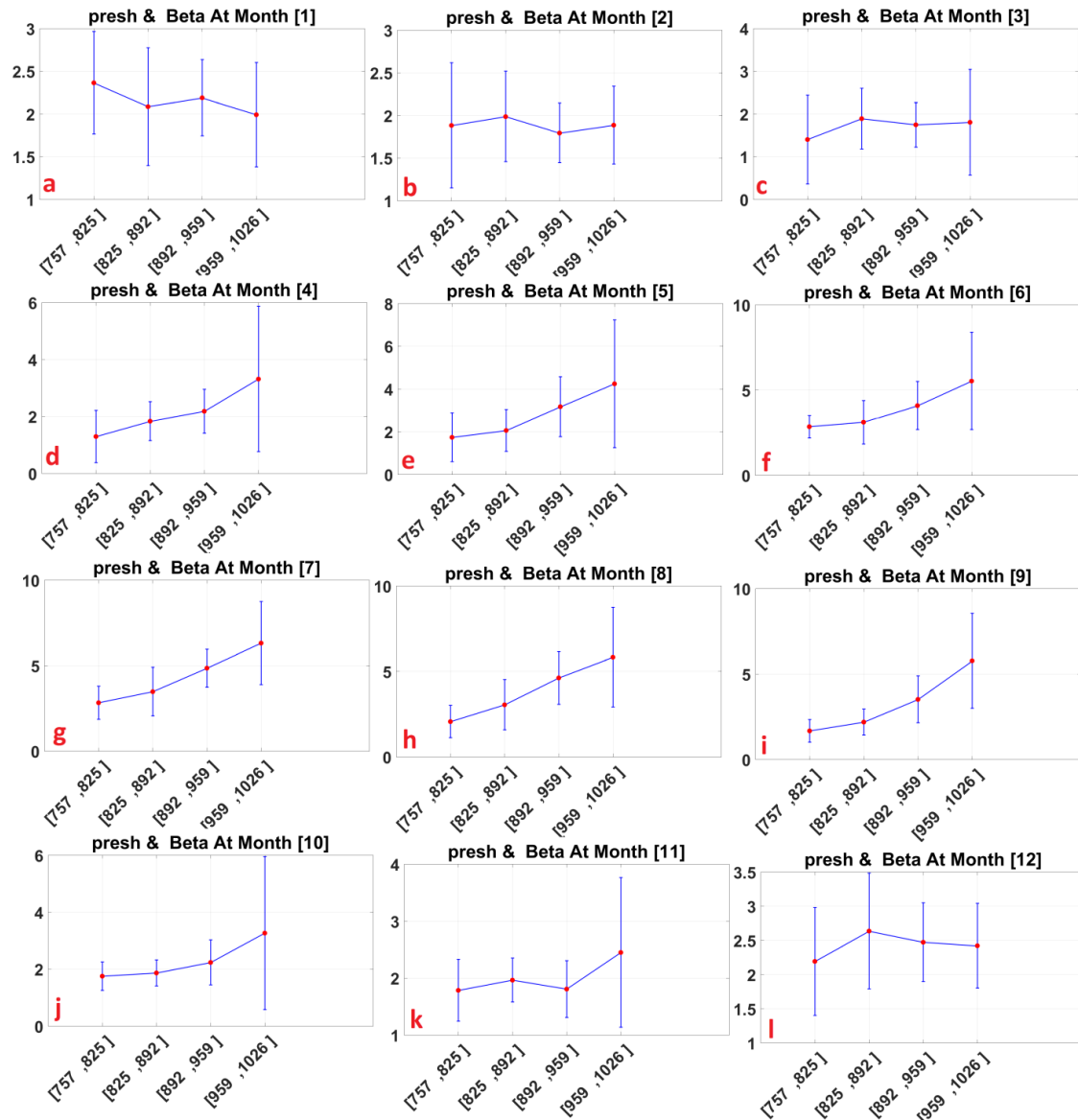


Fig. 7. Variation bars of mean and standard deviation of relative pressure against different β classes; a) January, b) February, c) March, d) April, e) May, f) June, g) July, h) August, i) September, j) October, k) November, and l) December

The mean of temperature range at each month were shown in Figure (8) in the study area. The spatial pattern of temperatures at each month did not show a specific behavior. This showed that spatial behavior of mean temperature range in the study area is complex

and is related to many variables. But in all maps it is clear that the mean temperature range was lower in north and south of Iran which are close to sea; the temperature ranges in south and south east in May, June, July, August, and September were decreased.

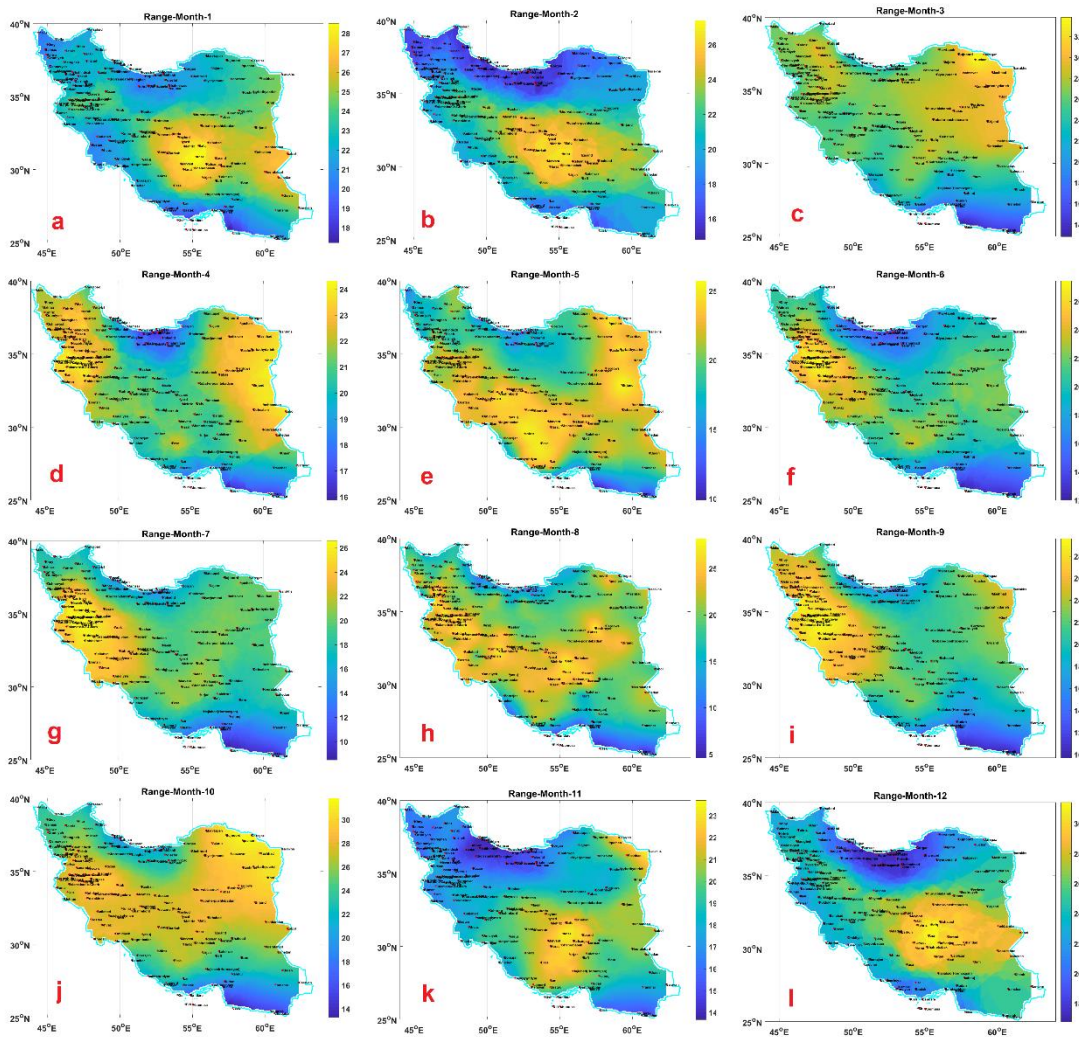


Fig. 8. The spatio-temporal behavior of temperature range in the study area; a) January, b) February, c) March, d) April, e) May, f) June, g) July, h) August, i) September, j) October, k) November, and l) December.

In Figure (9) the variation bars of mean diurnal temperature range and β values at each month are shown. The diurnal temperature range was divided to four classes based on maximum observed mean diurnal temperature range in weather stations. In January, February, and March β did not change significantly in four classes (Figure 9 a, b, c). From April to November (Figure 9 e, f, g, h, I, j, and k), β experienced significant ranges almost between 2 and 8.9 (in July it had the maximum value). The

general trend in these months shows that increasing the mean diurnal temperature range is associated with the decrease of β . It is to be noted that during this time period the variation of bars at each diurnal temperature range was decreased and approached the mean value. In March (Figure 9 c), the fourth class does not exist since no observation in this range was available. In November and December, the β range decreased significantly (Figure 9 i and l).

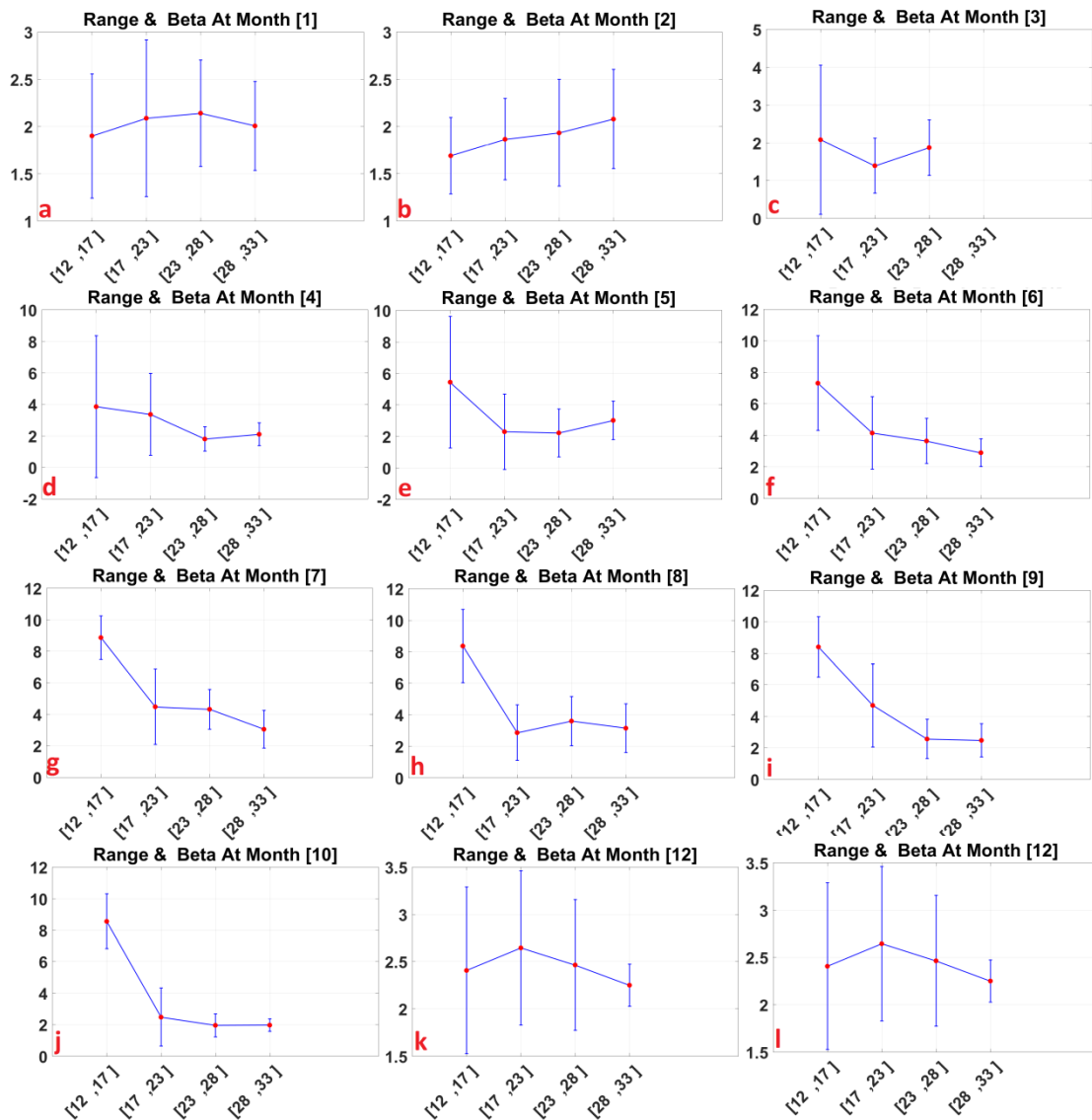


Fig. 9. Variation bars of mean and standard deviation of mean diurnal air temperature against different β classes; a) January, b) February) March, d) April, e) May, f) June, g) July, h) August, i) September, j) October, k) November, and l) December

In figure (10) the mean of wind speed were shown at each month in the study area. In January, February and March (figure 10 a, b and c) the pattern of wind speed is almost the same and central part of study area had higher values and the wind speed increased (between 3.2 and 3.8). In April, May, June, and July (Figure 10 d, e, f, and g) the mean wind speed was increased specially in eastern and south eastern parts

(maximum 6 m/s). This can be attributed to dominance of wind of 120 days in eastern and south eastern parts of Iran between May and October. In rest of months (Figure 10 h, I, j, k, and l), the mean wind speed decreased. Generally, west, south west and north of Iran experienced less wind speeds during study period time.

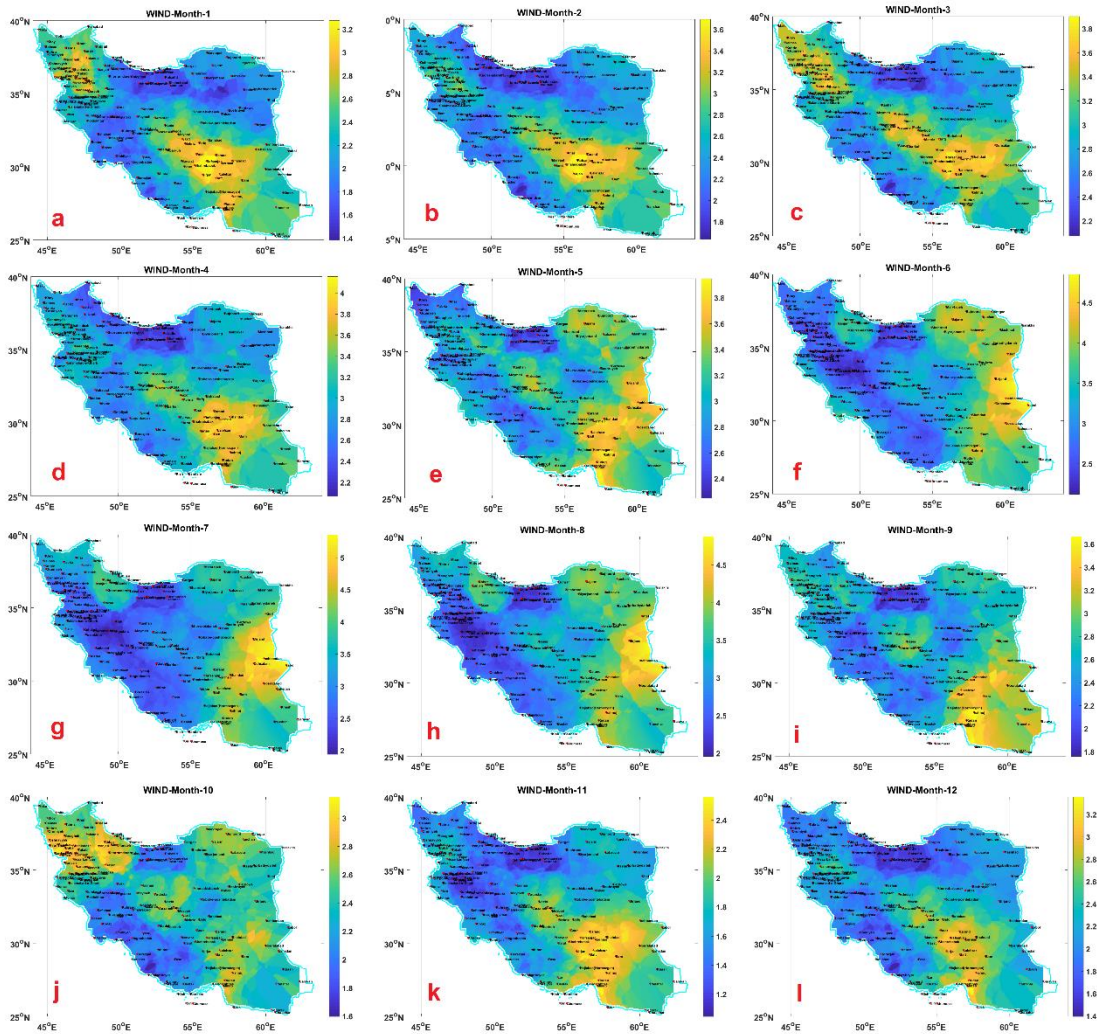


Fig. 10. The spatio-temporal behavior of wind in the study area; a) January, b) February) March, d) April, e) May, f) June, g) July, h) August, i) September, j) October, k) November, and l) December

Figure (11) shows the variations of mean wind speed in contrast with β for each month. The wind speed was classified based on the maximum observed range. From April to August the range of β increased and reached its maximum in July (5.7) (Figure 11 c, d, e, f, and

g). In January and February no general trend can be found in four classes but from March to December an increasing trend can be observed (Figure 11 h, I, j, k and l); the more the mean wind speed increased, the more the β increased.

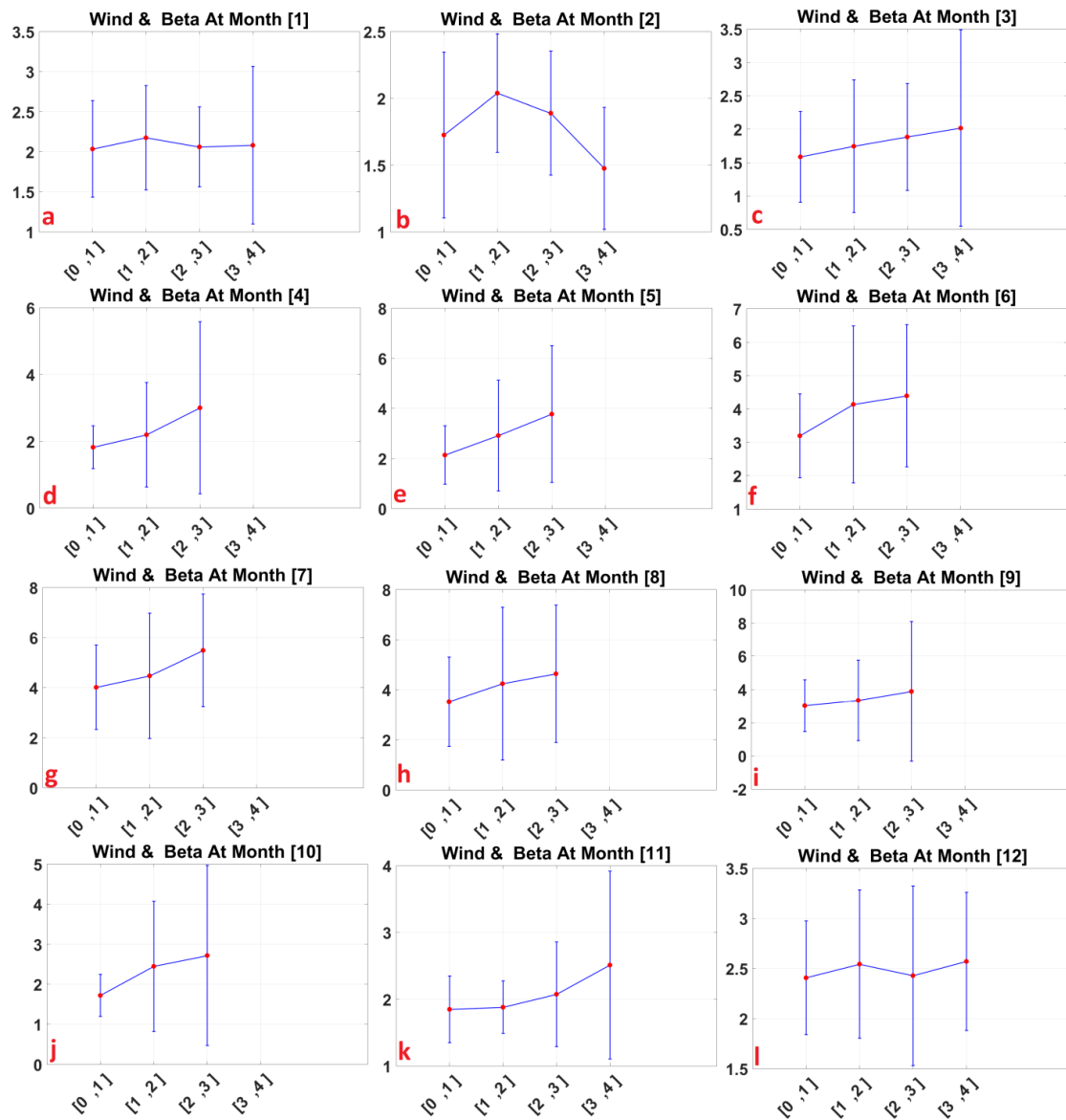


Fig. 11. Variation bars of mean and standard deviation of wind against different β classes; a) January, b) February, c) March, d) April, e) May, f) June, g) July, h) August, i) September, j) October, k) November, and l) December

In order to analyze the annual variations of β , the mean values of all environmental variables were calculated for one year and the variation bars were shown (Figure 12). In figure (12 a), the variation bar of β values against humidity classes were shown. In class two of humidity (30 to 50) an abrupt decrease in β is observed and an increasing trend is happened but the total range of β did not change significantly. The maximum variation in β values was observed at fourth class of humidity. Based on the results from figure (5) it can be concluded that the humidity can have a direct effect on β . Figure (12 b) shows the variation of β against with pressure classification. As the variation bar shows as far as pressure increases, the β increased. Also, the variations of β at each

classes increased by increasing pressure. It can be concluded that in high pressure areas in the study area like coast lines, the mean β values were higher and had higher oscillations. In figure (12 c) the behavior of β was analyzed in contrast with the mean diurnal temperature range. The general trend shows that the more the mean diurnal temperature increased, the more the β values and their variation at each class decreased; in Figure (13) the behavior of β for three values and their fitting RMSE for three weather stations are shown in 8 September of 2015. In Figure (13 a) the DTC model for Abali station in north of the study area is shown. Since the amplitude of diurnal temperature range is high the β value is small. In Figure (13 b) the DTC model in Boushehr station in southwest of

the study area has high β value and lower diurnal temperature range. In Figure (3 c) the β is the largest in comparison to last stations. The diurnal temperature range in Jask station was the lowest (less than 2 °C) and the RMSE of fitting is the largest (1.37 °C). Figure (12 d)

shows the β values against mean wind speeds. By mean wind speed increase, the β values increased. Also, the variations of β around mean were increased from the first to the third class but in the fourth class it decreased significantly.



Fig. 12. The variation bars of mean and standard deviation of relative mean diurnal air temperature against different β classes in one year; a) humidity, b) pressure) mean diurnal temperature range, d) wind

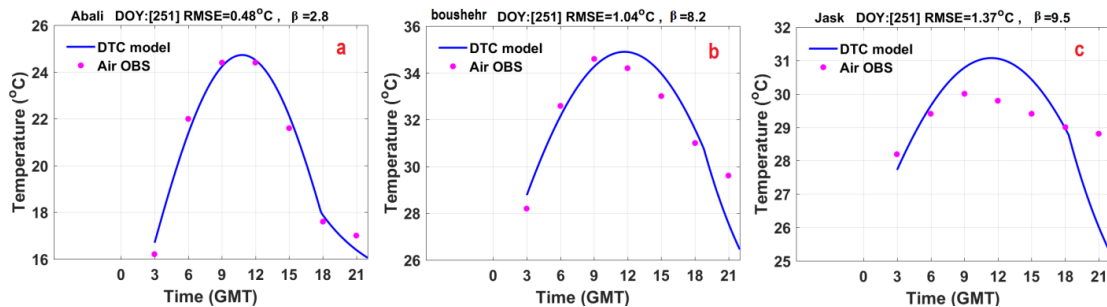


Fig. 13. The DTC model and air temperature observations in three sample weather stations; a) Abali, b) Boushehr, and c) Jask

5. Discussion and Conclusion

The diurnal modeling of air temperature by DTC model was implemented over Iran with RMSE of less than 2.1° C in 95% percent of data and the sensitivity of β coefficient as a dynamic variable in relation to environmental variables of humidity, pressure, diurnal temperature range, and wind speed were analyzed. The results showed that the accuracy of air DTC modeling did not have significant difference by considering β both as constant (for one year) or dynamic parameter (monthly) through the year. The variability of β showed significant monthly differences at each weather station and weather stations in south of Iran had the maximum values at most of months (Konarak station had maximum variation between 0.9 and 10). The relationships between β and environmental factors of humidity,

pressure, diurnal temperature range, and wind speed were analyzed. The monthly variations of humidity and β were analyzed by variation bars and no specific trend was found between these variables. The annual results suggested that humidity had an indirect effect on β parameter. The annual time series of pressure did not showed significant changes in all stations (Lalehzar in Kerman province was 4 hPa and Bostan in south west was 25 hPa). Results showed β had positive relation with pressure; in high pressure areas the values of β were lower but in low pressure areas a wide range of β were identified. In fact, it can be concluded that in high lands, β had low standard deviation (between 1.5 and 3.0) and in flat lands it had a wide range of values (between 1.5 and 6.5). This shows that in coast and desert areas this coefficient had different behaviors. The diurnal temperature range did not showed a specific

spatial pattern but the humid regions in north and south of the study area had lowest variations. Diurnal temperature range show negative relation with β ; as far as the diurnal temperature increases the standard deviation and values of β decreased; by increase of temperature range the β values experience lower standard deviations (between 1.8 and 2.5). This may shows that in desert areas with high diurnal temperature ranges, the β values were the lowest and in high lands they were highest. Wind variable showed positive effect on β ; the more the wind speed increased, the more mean of β values were increased. This can be concluded that in areas with higher wind values, the β values should increase to fit the mathematical model DTC to temperature observations. The results from this study can be applicable in many disciplines. It is necessary to conduct researches in future to analyze future the effect of environmental variables at each climate regime on β values. Also, it would be beneficial if the performance of different DTC approaches be compared to the study area.

Acknowledgments

The authors deeply thank the Iranian National Science Foundation and Iranian National Space Administration for their great support in conducting this research.

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