



Workshop on Quality Assurance Framework for Earth Observation (QA4EO) 29 September – 1 October, 2009, Antalya

The Importance of Traceability



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- The climate system
- Climatological Observations
 In-situ (Proxy, Historical, Conventional, AWOS, etc)
 Remote sensing (Radar, Lidar, Satellite, etc)
- Calibration and QC of observational data Laboratory check, limit check, internal consistency, relational and area control
- Metadata issue
- Climate data management Archive, access and retrieval, QC, homogeneity, statistical process, missing value estimation, data rescue, exchange
- Traceability of observation
 Time series, monitoring, indices, graphics, climate atlas
- Conclusion
- Acknowledgement



Climate system is comprised by the complicated interactions among the atmosphere, the ocean, the cryosphere, the surface lithosphere and the biosphere.





Energy from the sun drives the earth's weather and climate, and heats the earth's surface; in turn, the earth radiates energy back into space. The greenhouse effect is a necessary phenomenon. Without it Earth temperature would be -18°C. But the Greenhouse gases trap some of the outgoing energy and maintain Earth's temperature 15°C. However, too many greenhouse gases could increase in mean temperatures **Radiative flux balance**

 $\pi r^2 S_0(1-a) = 4\pi r^2 \sigma T^4$

=255K=-18°C

(1-0.3)×1.37×10



as the integration of discreet weather events and variables over time and space

(by J.W. Zillman, WMO Bulletin 48 No.2

Table: Summary of atmospheric scales in time and space

1 hour to 1 day

1/2 day to 1 week

3 days and longer

Mesoscale (regional scale) 1 minutes to 3 hours

Synoptic scale

Macroscale

Global scale

4/30

10 m to 1 km

100 m to 10 km

1 km to 100 km

1 km to 20 km

300 m to 30 km

3 km to 1000 km

300 km to globe

30 km to 10000 km



Conventional Climatological Observations and Instruments









Instrument and sensor calibration



Opening the Meteorological Calibration Center (KALMER) in TSMS, 24.09.2009



centres of excellence for processing satellite data and form an integral part of the distributed EUMETSAT Application Ground Segment.





Historical data & Data Rescue



Climate data archive waiting for rescue

Data Rescue is :

WMO **MEDARE** initiative The MEditerranean MEDARE climate DAta REscue Download PDF (5 MB) **Workshop Proceedings** (MEDARE) is an initiative, born under the auspice of the World Meteorological Organization, with the main • The MEDARE Initiative objective is being to develop, consolidate and progress climate data and metadata rescue activities across the Greater MEDARE Workshop outcomes Mediterranean Region (GMR)

Workshop MEDARE

WCDMP World Climate Data

and Monitoring

World

Climate

an ongoing process of preserving all data at risk of being lost due to deterioration of the medium, and the digitization of current and past data into computer compatible form for easy access. There are several on going data rescue program. **WMO MEDARE Initiative** is one of them.



1000 to 1861, N.Hemisphere, proxy data; 1861 to 2000 Global, instrumental; 2000 to 2100, SRES projections

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Data management system





Data quality control includes comparison to physical and climatological limits and inter-comparison of data elements.



Metadata (Data about data)



Station documentation is information about the data or data about the data: Metadata should reflect how, where, when and by whom information was collected. Ideally, complete metadata should register all the changes a station has undergone during its lifetime, composing what is called the station history. Additional information about instrument or exposure, can provide additional insights. Sometimes when the instruments change, the observations will show an artificial increase or decrease. Such jump cause an inhomogeneity and adjustments needed to these data. If a long-term time series is homogeneous, then all variability and change is due to the behavior of the atmosphere. (By Aguilar, E., et al.)

ETEOROLO





Even if the small changes in the site locations can cause big breakpoints in the time series.



By Aguilar, E., et al.





Homogeneity test

RIZE ANNUAL MIN



Homogeneity test result of annual minimum temperature for station Rize, Turkey. The discontinuity in 1995 is reflected in metadata which shows that the station relocated in this year.

Data homogeneity is assessed using R-based program, RHtest, developed at the Meteorological Service of Canada. It is based on two-phase regression model with a linear trend for the entire base series (Wang, 2003). There is also cost action in homogenization http://www.homogenisation.org



Statistical process, estimation of missing data, Normal (are calculated from 1971-2000)

	F8 •	fs =(1	=(F6+F11)/2 = (J7+J9)/2														
	A	В	C	0	E	F	G	н	1	J	K	L	M	N	0	P	
1	ISTASYON	- ELEMENT -	YIL -	OCK -	SBT +	MRT -	15 -	MAY -	HAZ -	1 14 -	AGST -	EYL -	екім 👻	KAS -	ARA -	YILLI -	100.001
2	ACIPAYAM	ORT-SIC	1971	5.1	3.3	6.6	10.0	15.3	19.9	22.9	22.3	18.8	11.4	7.3	1.1	12.1	= mean (D2:02)
3	ACIPAYAM	ORT-SIC	1972	0.8	1.7	7.6	13.2	16.0	19.7	22.9	22.8	19.7	12.2	6.2	1.1	12.0	
4	ACIPAYAM	ORT-SIC	1973	0.7	4.0	5.3	9.4	16.7	18.6	24.2	23.8	20.6	13.7	4.9	3.7	12.1	
5	ACIPAYAM	ORT-SIC	1974	-0.9	33	8.5	9.6	15.5	21.7	24.5	22.4	18.3	15.7	7.1	1.9	12.3	
6	ACIPAYAM	ORT-SIC	1975	1.7	1.7	7.2	11.8	14.5	18.8	24.3	23.1	19.2	13.1	6.5	1.7	12.0	
7	ACIPAYAM	ORT-SIC	1976	0.0	0.2	6.5	9.7	15.1	19.3	22.0	21.3	17.5	14.3	7.9	3.4	11.4	estimated
8	ACIPAYAM	ORT-SIC	1977	1.3	5.8	7.3	10.3	16.3	19.9	23.5	21.8	17.5	14.3	6.8	3.4	12.3	
9	ACIPAYAM	ORT-SIC	1978	1.8	5.7	6.3	10.8	17.4	20.4	25.0	22.3	17.4	14.2	5.7	3.4	12.5	data
10	ACIPAYAM	ORT-SIC	1979	2.2	5.6	8.0	10.8	14.5	19.6	23.0	22.9	19.9	14.1	7.7	3.3	12.6	data
11	ACIPAYAM	ORT-SIC	1980	0.3	2.6	5.3	9.7	15.2	21.0	25.0	24.1	18.1	14.2	8.5	3.8	12.3	
12	ACIPAYAM	ORT-SIC	1981	1.9	2.5	7.8	10.7	13.5	20.6	23.6	22.6	19.4	15.1	5.7	6.2	12.5	
13	ACIPAYAM	ORT-SIC	1982	3.2	0.8	5.0	10.6	15.1	19.9	21.4	23.1	19.2	12.8	5.4	3.1	11.6	Deferences
14	ACIPAYAM	ORT-SIC	1983	-1.1	0.7	5.8	11.5	15.7	18.0	21.2	21.5	18.2	11.8	8.5	4.5	11.4	Reference.
15	ACIPAYAM	ORT-SIC	1984	3.4	4.8	6.1	9.2	17.3	21.0	23.0	21.2	19.6	14.2	7.4	2.3	12.5	WMO No 100
16	ACIPAYAM	ORT-SIC	1985	4.3	-0.6	6.0	12.1	16.8	20.8	23.2	24.8	19.5	10.9	9.3	3.5	12.6	
17	ACIPAYAM	ORT-SIC	1986	3.8	4.7	7.9	13.3	13.8	19.9	25.0	24.5	19.1	12.3	4.7	1.7'	12.6	
18	ACIPAYAM	ORT-SIC	1987	3.3	4.3	2.4	9.0	15.1	20.4	24.9	23.4	21.0	12.4	6.5	3.9	12.2	
19	ACIPAYAM	ORT-SIC	1988	3.4	3.0	4.1	11.6	17.5	21.6	26.0	23.7	19.5	12.1	4.7	3.7	12.6	
20	ACIPAYAM	ORT-SIC	1989	-0.1	2.8	8.4	15.7	16.3	20.2	24.5	24.8	19.8	11.8	7.2	3.0'	12.9	
21	ACIPAYAM	ORT-SIC	1990	0.0	3.5	7.9	11.2	15.5	21.0	25.8	24.0	19.0	14.3	8.9	4.4	13.0	
22	ACIPAYAM	ORT-SIC	1991	1.6	3.2	9.2	10.9	13.9	22.5	24.0	23.8	19.7	14.2	7.2	0.5'	12.6	
23	ACIPAYAM	ORT-SIC	1992	-1.9	-0.7	4.3	11.3	15.3	20.3	22.3	24.3	18.7	16.3	6.8	0.5'	11.5	
24	ACIPAYAM	ORT-SIC	1993	-0.1	1.2	5.6	11.3	14.3	21.0	24.3	25.2	19.9	16.0	7.0	5.0'	12.6	
25	ACIPAYAM	ORT-SIC	1994	4.5	3.8	7.1	13.0	17.2	21.3	24.2	24.8	22.9	15.8	6.6	2.3	13.6	
26	ACIPAYAM	ORT-SIC	1995	3.8	5.6	6.2	9.7	16.8	22.9	23.3	24.5	20.6	12.8	4.9	4.7	13.0	
27	ACIPAYAM	ORT-SIC	1996	1.9	5.1	5.3	9.4	18.0	22.1	25.9	25.0	18.4	12.1	9.3	6.6	13.3	
28	ACIPAYAM	ORT-SIC	1997	3.6	2.6	43	7.0	17.8	21.4	24.2	21.7	18.5	13.6	8.7	4.8	12.4	
29	ACIPAYAM	ORT-SIC	1998	29	4.9	3.9	12.9	15.7	21.2	25.6	26.3	19.5	14.8	9.6	4.3	13.5	
30	ACIPAYAM	ORT-SIC	1999	4.3	3.8	7.3	12.0	18.4	20.8	24.8	24.6	19.7	15.4	9.0	6.0'	13.8	
31	ACIPAYAM	ORT-SIC	2000	-0.8	25	5.1	12.5	16.1	21.6	26.4	24.7	19.7	13.3	9.3	3.8	12.9	
32	ACIPAYAM	ORT-SIC	71-00	1.8	31	6.3	11.0	15.9	20.6	24.0	23.5	19.3	13.6	7.2	3.4	12.5	- = mean (P2:P31)
33	ACIPAYAM	ORT-SIC	stdev	2.0	1.8	1.6	1.7	1.3	1.1	1.3	1.3	1.1	1.5	1.5	1.6	0.6	= STD (P2-P31)
34	ACPAYAM	ORT-SIC	CV	106.1	60.0	25.4	15.4	8.2	5.4	5.6	5.6	5.9	10.7	20.9	46.8	49 -	
															CV	=(P33	(P32)*100

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Products-Monthly, seasonal and annual assessments

Anomaly detection is the traceability option of the data. **Anomaly** is the differences from normal and it is determining by using *Standardized Normal Distribution* which is formulated as Z=(X-X mean)/STD. If Z < -1 it means the data is below normal, if Z is in between -1 to 1, near normal, and if Z > 1 it means this value is over normal.



Products- Monitoring & Time series

Annual climate assessment

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Maps and Time series are very good tools to trace spatial and temporal variation of data. Below there is an assessment for the year 2008. 2008 mean temperature was 0.8°C above the 1961-1990 normal (13.6°C). Generally coastal area and western part of the country had temperatures above the mean. Positive temperature anomalies have been occurred since 1994 (except 1997) (Fig. 1). This kind of analyses have been contributed to the "Statement on the Status of the Climate " which published by WMO and NOAA, BAMS













Products- Climate Diagram



Climate Diagram is another traceability study which shows monthly variation on precipitation and temperature in the study area.



It's rainy in Winter but dry and hot in Summer (Mediterranean climate)



Products- Climate atlas



Climate atlas is another monitoring and traceability options. All the data used in the atlas have enough time length (at least 30 years) and must be quality controlled. Nowadays GIS techniques (Geostatistical analyze, Co-kriging, IDW, GWR) have been used For interpolation, modeling and monitoring the data.



20/30



Climate Indices

CANCEL

RClimDex produced on behalf of the ET by Xuebin Zhang from Met. Service of Canada

http://cccma.seos.uvic.ca/ET CCDMI/



OK

	APP	PENDIX A - List of the 27 core climate indices								
	No	ID	Indicator name	Definitions	UNITS					
	1	FDO	Frost days	Annual count when TN(daily minimum)<0°C	Days					
	2	SU25	Summer days	Annual count when TX(daily m aximum)>25°C	Days					
[3	IDO	I ce days	Annual count when TX(daily m axim um)<0℃	Days					
	4	TR20	Tropical nights	Annual count when TN(daily minimum)>20°C						
			Growing season Length	Annual (1 st Jan to 31ª Dec in NH, count between first span of						
				at least 6 days with TG>5 $^{\circ}\mathrm{C}$ and first span after July 1 of 6						
	- 5	GSL		days with TG<5℃	Days					
-	б	TXx	Max Tm ax	Monthly maximum value of daily maximum temp	ç					
	7	TNx	Max Tmin	Monthly maximum value of daily minimum temp	င					
	8	TXn	Min Tmax	Monthly minimum value of daily maximum temp	င္					
	9	TNn	Min Tmin	Monthly minimum value of daily minimum temp	င္					
	10	TN10p	Cool nights	Percentage of days when TN<10th percentile	Days					
	11	TX10p	Cool days	Percentage of days when TX<10th percentile	Days					
	12	TN90p	Warm nights	Percentage of days when TN>90th percentile	Days					
t	13	TX90p	Warm days	Percentage of days when TX>90th percentile	Days					
			Warm spell duration	Annual count of days with at least 6 consecutive days when						
	14	WSDI	indicator	TX>90th percentile	Days					
		aabi	Cold spell duration	Annual count of days with at least 6 consecutive days when	D					
╞	1		indicator	IN-IUth percentile	Days					
┟	10	DTR	Diumai temperature	Monthly mean difference between IX and IN	τ					
	17	RXI day	Max 1-day precipitation	Monthly maximum 1-day precipitation	mm					
	18	Rx5day	Max 5-day precipitation	Monthly maximum consecutive 5-day precipitation	mm					
			Simple daily intensity	Annual total precipitation divided by the number of wet days						
	19	SDII	index	(defined as PRCP>=1.0mm) in the year	mm/day					
	20	R10	Number of heavy prec.	Annual count of days when PRCP>=10mm	Days					
	21	R20	Number of very heavy P	Annual count of days when PRCP>=20mm	Days					
			Number of days above nn	Annual count of days when PRCP>=nn mm, nn is user						
	22	Rnn	mm	defined threshold	Days					
	23	CDD	Consecutive dry days	Maximum number of consecutive days with RR<1mm	Days					
	24	CWD	Consecutive wet days	Maximum number of consecutive days with RR>=1 mm	Days					
<	25	R95p	V ery wet days	Annual total PRCP when RR>95 th percentile	mm					
	26	R99p	Extrem ely wet days	Annual total PRCP when RR>99 th percentile	mm					
	27	PRCPTOT	Annual total wet-day	Annual total PRCP in wet days (RR>=1 mm)	mm					



Middle East Climate Indices Study





//www.agu.org/pubs/crossref/2005/2005JD006181.shtml

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JOURNAL OF GEOPHYSICAL RESEARCH, VOL. 110, D22104, doi:10.1029/2005JD006181, 2005

Trends in Middle East climate extreme indices from 1950 to 2003

Xuebin Zhang

SAGU

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http://www.agu.org/pubs/crossref/2005/2005JD006181.shtml

It shows that warm nights and warm days have been increasing in the Middle East.



Kendall's tau based slope estimator has been used to compute the trends since this method doesn't assume a distribution for the residuals and is robust to the effect of outliers in the series. If slope error greater than slope estimate we can't trust slope estimate.

This indices plot shows that frost days will be decreasing 89.9 days in 100 years in Gaziantep and this trend is statistically significant at 95% level of confidence because of P Value is less than 0.05.





Advantages of Indices versus Data

- Indices are information derived from data
- It represents the data
- More readily released than data
- Useful in a wide variety of climate change analyses
- Useful for Model observations comparisons
- Useful for analyses of extremes
- They can be more accessible than data (exchangeable)



Climate indices continued





Trends in Number of Summer Days from 1971 to 2004 [Tx > 25°C]

Numbers of Summer Days have been increasing all over Turkey.



90N

45N

45S

Trends were calculated only for the grid boxes with sufficient data (at least 40 years of data. Black lines enclose regions where trends are significant at the 95% confidence of level. The red curves on the plots are nonlinear trend estimates obtained by smoothing using a 21-term binomial filter.





Global Indices Analyses

By Alexander, L. et al

Trends in (a) cold nights (TN10p), (b) warm nights (TN90p), (c) cold days (TX10p) and (d) warm days (TX90p).

(d) Warm days





Projected surface temperature changes for the early and late 21st century. If the PDF is widespread, it means that too many extreme event could be occurred.





GIS isn't only monitoring software, it's capable to combine all kind of geographically referenced data, (vector, table, satellite, raster, proxy, etc) and to analyze, to select, to model and to monitor them.





✓ Climate system is comprised by the complicated interactions among the atmosphere, the ocean, the cryosphere, the surface lithosphere and the biosphere.

✓Long time reliable climate observation and information are necessary to detect climate variability and change and to help decision makers to mitigate and adaptation

✓ Climate data needs to undertake quality control, homogeneity assessment, data rescue and statistical process in order to produce reliable climate information and products.

 \checkmark There are still huge amount of historical climate data which need to rescue and digitization.

 \checkmark Climate indices also very useful tool to detect climate variability and to understand trends in the extreme.

✓GIS is also very useful tool for traceability. It's not only monitoring software, it's capable to combine all kind of geographically referenced data, (vector, table, satellite, raster, proxy, etc) and to analyze, to select, to model and to monitor them.





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