

Exercise for Guidance

TCC Training Seminar on Application of
Seasonal Forecast GPV Data to Seasonal
Forecast Products
18-21 January 2011

Objectives

- To understand how to make guidance for seasonal-mean temperature and precipitation.
- To find out effective predictors.
- To make more accuracy guidance.

Procedure

1. Single Regression Model
2. Multiple Regression Model
3. Probabilistic Forecast
4. Verification
5. Create a presentation
6. Presentation (5 minutes)

Today

Tomorrow

The day after tomorrow

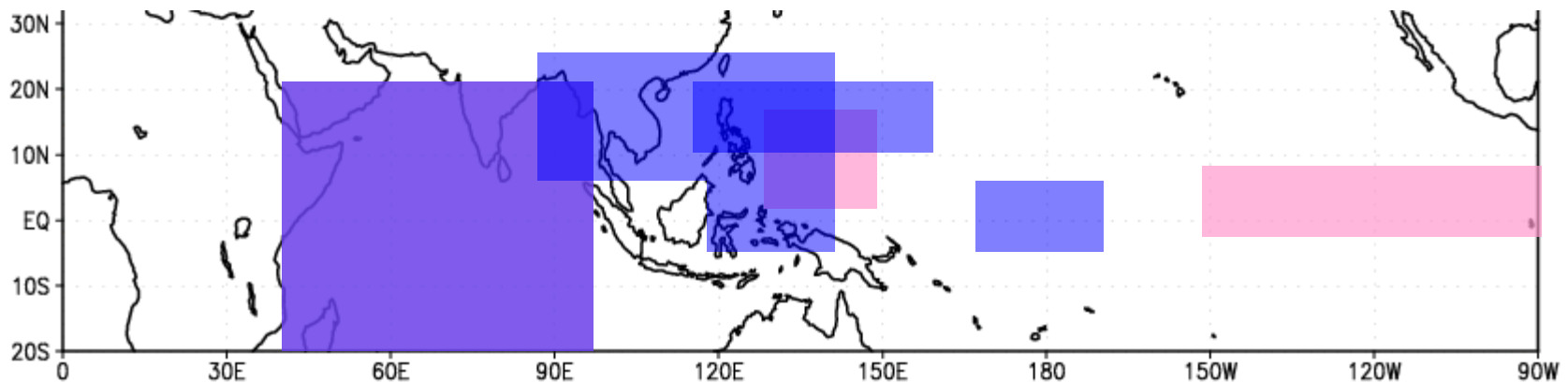


Preparations

- Observation data (yourself)
- Predictors (init. 1st May for JJA, 28th Oct for DJF)
 - GPV data over your stations 「*GPVdata.xls*」
 - Indices such as NINO3SST, Asia Monsoon, etc.
「*indices.xls*」
- Excel software for making guidance
「*ExerciseForGuidance.xls*」
- Textbook 「*Exercise for Guidance.doc*」

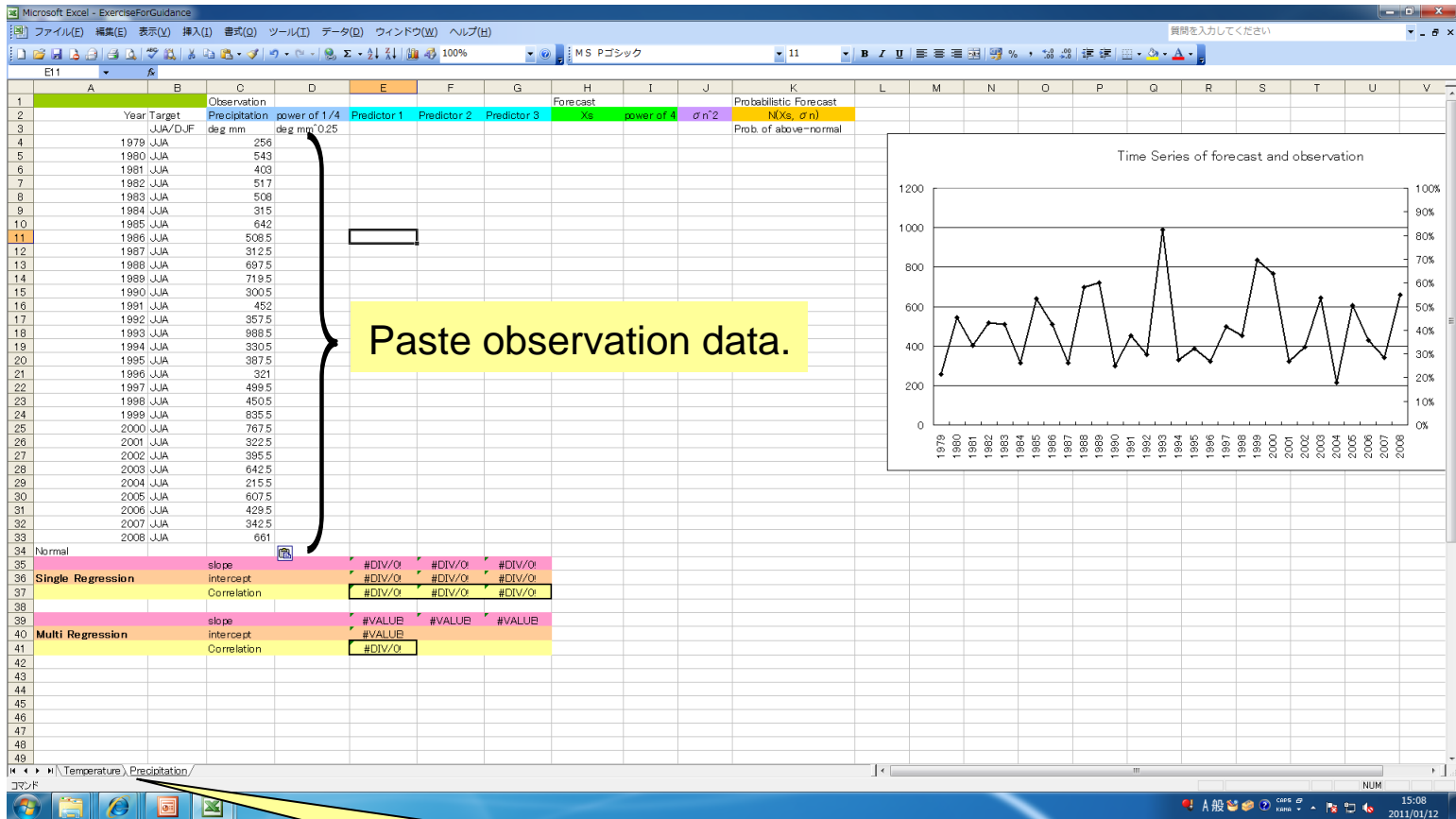
Indices

	indices	variables	areas
SST	NINO3 SST	SST	(150W-90W, 5S-5N)
	NINOWEST SST	SST	(130E-150E, EQ-15N)
	IOBW SST	SST	(40E-100E, 20S-20N)
RAIN	IOBW RAIN	RAIN	(40E-100E, 20S-20N)
	SAMOI RAIN	RAIN	(80E-140E, 5N-25N)
	WNP RAIN	RAIN	(110E-160E, 10N-20N)
	SEAsia RAIN	RAIN	(115E-140E, 10N-20N)
	MC RAIN	RAIN	(110E-135E, 5S-5N)
	DL RAIN	RAIN	(170E-170W, 5S-5N)
Z500	Z2030	500hPa Height	(0-360, 20N-30N)
	Z3040	500hPa Height	(0-360, 30N-40N)
	Z4050	500hPa Height	(0-360, 40N-50N)
	Z5060	500hPa Height	(0-360, 50N-60N)
Thickness	THMD	Thickness Middle	(0-360, 30N-50N, 300hPa-850hPa)
	THEX	Thickness extratropic	(0-360, 30N-90N, 300hPa-850hPa)
	THTR	Thickness tropic	(0-360, 25S-25N, 100hPa-850hPa)



1. Single Regression Model

- Open the ExerciseForGuidance.xls.
- Paste observation data on a Temperature/Precipitation worksheet.



Temperature/Precipitation worksheet

1. Single Regression Model

- Calculate the power of $\frac{1}{4}$ for normalization in case of precipitation.
- Calculate normal from 1979 to 2008.

The screenshot shows an Excel spreadsheet with the following data and formulas:

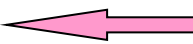
Year	Target	Observation	power of 1/4	Predictor 1	Predictor 2	Predictor 3	Forecast	Probabilistic Forecast
JJA/DJF	deg mm	deg mm	deg mm ^{0.25}	Xs	power of 4	σ n ²	N(Xs, σ n)	Prob. of above-normal
1979	JJA	256	4.0					
1980	JJA	543	4.8					
1981	JJA	403	4.5					
1982	JJA	517	4.8					
1983	JJA	508	4.7					
1984	JJA	315	4.2					
1985	JJA	642	5.0					
1986	JJA	508.5	4.7					
1987	JJA	312.5	4.2					
1988	JJA	697.5	5.1					
1989	JJA	719.5	5.2					
1990	JJA	300.5	4.2					
1991	JJA	452	4.6					
1992	JJA	357.5	4.3					
1993	JJA	988.5	5.6					
1994	JJA	330.5	4.3					
1995	JJA	387.5	4.4					
1996	JJA	321	4.2					
1997	JJA	499.5	4.7					
1998	JJA	450.5	4.6					
1999	JJA	835.5	5.4					
2000	JJA	767.5	5.3					
2001	JJA	322.5	4.2					
2002	JJA	395.5	4.5					
2003	JJA	642.5	5.0					
2004	JJA	215.5	3.8					
2005	JJA	607.5	5.0					
2006	JJA	429.5	4.6					
2007	JJA	342.5	4.3					
2008	JJA	661	5.1					

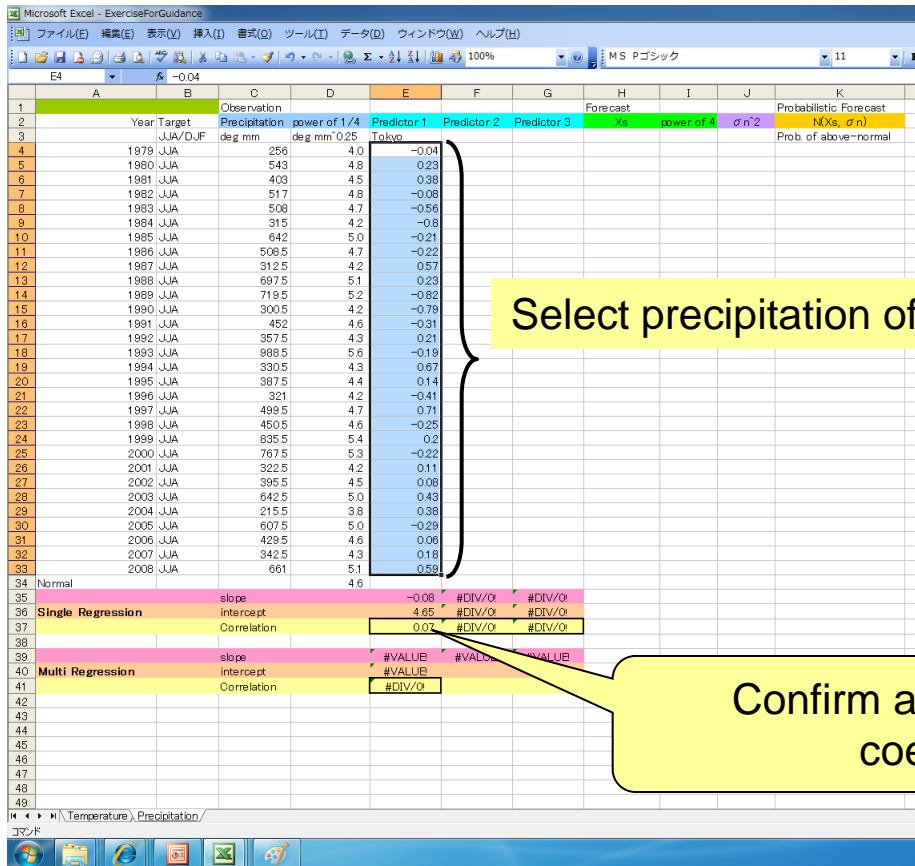
Callout 1: Calculate power of $\frac{1}{4}$ in case of precipitation. Input `"=C4^0.25"` at D4.

Callout 2: Calculate normal from 1979 to 2008. `"=AVERAGE(D4:D33)"` at D34

The line graph shows the time series of forecast and observation from 1979 to 2008. The x-axis represents the year, and the y-axis represents the precipitation amount in deg mm. The forecast is shown as a horizontal line, and the observation is shown as a fluctuating line.

1. Single Regression Model

- Open GPVdata.xls and indices.xls.  **Predictors**
- Select a predictor and paste E line. In case of temperature, paste D line.



Year	Target	Observation	power of 1/4	Predictor 1	Predictor 2	Predictor 3	Forecast	power of 4	σn^2	Probabilistic Forecast
JJA/DJF	deg mm	deg mm	deg mm ^{0.25}	Tokyo			Xs			N(Xs, σn)
1979	JJA	256	4.0	-0.04						Prob. of above-normal
1980	JJA	543	4.8	0.23						
1981	JJA	403	4.5	0.38						
1982	JJA	517	4.6	-0.08						
1983	JJA	508	4.7	-0.56						
1984	JJA	315	4.2	-0.8						
1985	JJA	642	5.0	-0.21						
1986	JJA	508.5	4.7	-0.22						
1987	JJA	312.5	4.2	0.57						
1988	JJA	697.5	5.1	0.23						
1988	JJA	719.5	5.2	-0.82						
1989	JJA	300.5	4.2	-0.78						
1991	JJA	452	4.6	-0.31						
1992	JJA	357.5	4.3	0.21						
1993	JJA	988.5	5.6	-0.19						
1994	JJA	330.5	4.3	0.67						
1995	JJA	387.5	4.4	0.14						
1996	JJA	321	4.2	-0.41						
1997	JJA	499.5	4.7	0.71						
1998	JJA	450.5	4.6	-0.25						
1998	JJA	835.5	5.4	0.2						
2000	JJA	767.5	5.3	-0.22						
2001	JJA	322.5	4.2	0.11						
2002	JJA	395.5	4.5	0.08						
2003	JJA	642.5	5.0	0.43						
2004	JJA	215.5	3.8	0.38						
2005	JJA	607.5	5.0	-0.28						
2006	JJA	429.5	4.6	0.06						
2007	JJA	342.5	4.3	0.18						
2008	JJA	661	5.1	0.58						

Normal	slope	-0.08	#DIV/0!	#DIV/0!
Single Regression	intercept	4.65	#DIV/0!	#DIV/0!
	Correlation	0.07	#DIV/0!	#DIV/0!

Multi Regression	slope	#VALUE!	#VALUE!	#VALUE!
	intercept <th>#VALUE!</th> <th>#VALUE!</th> <th>#VALUE!</th>	#VALUE!	#VALUE!	#VALUE!
	Correlation <th>#DIV/0!</th> <th>#DIV/0!</th> <th>#DIV/0!</th>	#DIV/0!	#DIV/0!	#DIV/0!

Select precipitation of GPV data over Tokyo as a predictor.

Confirm an anomaly correlation coefficient at E37.

In case of Tokyo, ACC is near zero. So I try the other predictors.

1. Single Regression Model

- Try using the other predictors.
- We can find out more effective predictor for Tokyo. It is the Indian Ocean SST.

1	A	B	C	D	E	F	G	H	I	J	K
2	Year	Target	Observation	power of 1/4	Predictor 1	Predictor 2	Predictor 3	Forecast	power of 4	$\sigma^2 n^2$	Probabilistic Forecast
3	JJA/DJF	Precipitation	deg mm	deg mm ^{0.25}	IOBSW SST			Xs			N(Xs, σn)
4	1979	JJA	256	4.0	-0.01						Prob. of above-normal
5	1980	JJA	543	4.8	0.05						
6	1981	JJA	403	4.5	-0.03						
7	1982	JJA	517	4.6	0.06						
8	1983	JJA	508	4.7	0.16						
9	1984	JJA	315	4.2	-0.16						
10	1985	JJA	642	5.0	-0.19						
11	1986	JJA	508.5	4.7	-0.19						
12	1987	JJA	312.5	4.2	0.29						
13	1988	JJA	697.5	5.1	0.06						
14	1988	JJA	719.5	5.2	-0.2						
15	1990	JJA	300.5	4.2	0.03						
16	1991	JJA	452	4.6	0.04						
17	1992	JJA	357.5	4.3	0.16						
18	1993	JJA	988.5	5.6	-0.03						
19	1994	JJA	330.5	4.3	-0.09						
20	1995	JJA	387.5	4.4	0.03						
21	1996	JJA	321	4.2	-0.01						
22	1997	JJA	499.5	4.7	0.03						
23	1998	JJA	450.5	4.6	0.36						
24	1998	JJA	835.5	5.4	-0.27						
25	2000	JJA	767.5	5.3	-0.2						
26	2001	JJA	322.5	4.2	0.04						
27	2002	JJA	395.5	4.5	0.09						
28	2003	JJA	642.5	5.0	0.03						
29	2004	JJA	215.5	3.8	-0.06						
30	2005	JJA	607.5	5.0	0.16						
31	2006	JJA	429.5	4.6	-0.06						
32	2007	JJA	342.5	4.3	0.14						
33	2008	JJA	661	5.1	-0.21						
34	Normal			4.6							
35	Single Regression	slope			-0.97	#DIV/0!	#DIV/0!				
36		intercept			4.65	#DIV/0!	#DIV/0!				
37		Correlation			0.33	#DIV/0!	#DIV/0!				
38											
39	Multi Regression	slope			#VALUE!	#VALUE!	#VALUE!				
40		intercept			#VALUE!	#VALUE!	#VALUE!				
41		Correlation			#DIV/0!	#DIV/0!	#DIV/0!				
42											
43											
44											
45											
46											
47											
48											
49											

Select the Indian Ocean SST and paste them on E line.

That make the anomaly correlation coefficient increase to 0.33.

Let's try to look for most effective predictor for your country.

1. Single Regression Model

- Calculate the forecasts using single regression equation.

Year	Target	Precipitation	power of 1/4	Predictor 1	Predictor 2	Predictor 3	Forecast	power of 4	σ n'2	Probabilistic Forecast
1979	JJA	256	4.0	-0.01			4.66			
1980	JJA	543	4.8	0.05			4.60			
1981	JJA	403	4.5	-0.03			4.68			
1982	JJA	517	4.8	0.06			4.59			
1983	JJA	508	4.7	0.16			4.49			
1984	JJA	315	4.2	-0.16			4.80			
1985	JJA	642	5.0	-0.19			4.83			
1986	JJA	508.5	4.7	-0.19			4.83			
1987	JJA	312.5	4.2	0.29			4.37			
1988	JJA	697.5	5.1	0.06			4.59			
1988	JJA	719.5	5.2	-0.2			4.84			
1990	JJA	300.5	4.2	0.03			4.62			
1991	JJA	452	4.6	0.04			4.61			
1992	JJA	357.5	4.3	0.16			4.49			
1993	JJA	988.5	5.6	-0.03			4.68			
1994	JJA	330.5	4.3	-0.09			4.74			
1995	JJA	387.5	4.4	0.03			4.62			
1996	JJA	321	4.2	-0.01			4.66			
1997	JJA	499.5	4.7	0.03			4.62			
1998	JJA	450.5	4.6	0.36			4.30			
1998	JJA	835.5	5.4	-0.27			4.91			
2000	JJA	767.5	5.3	-0.2			4.84			
2001	JJA	322.5	4.2	0.04			4.61			
2002	JJA	395.5	4.5	0.09			4.56			
2003	JJA	642.5	5.0	0.03			4.62			
2004	JJA	215.5	3.8	-0.06			4.71			
2005	JJA	607.5	5.0	0.16			4.49			
2006	JJA	429.5	4.6	-0.06			4.71			
2007	JJA	342.5	4.3	0.14			4.51			
2008	JJA	661	5.1	-0.21			4.85			

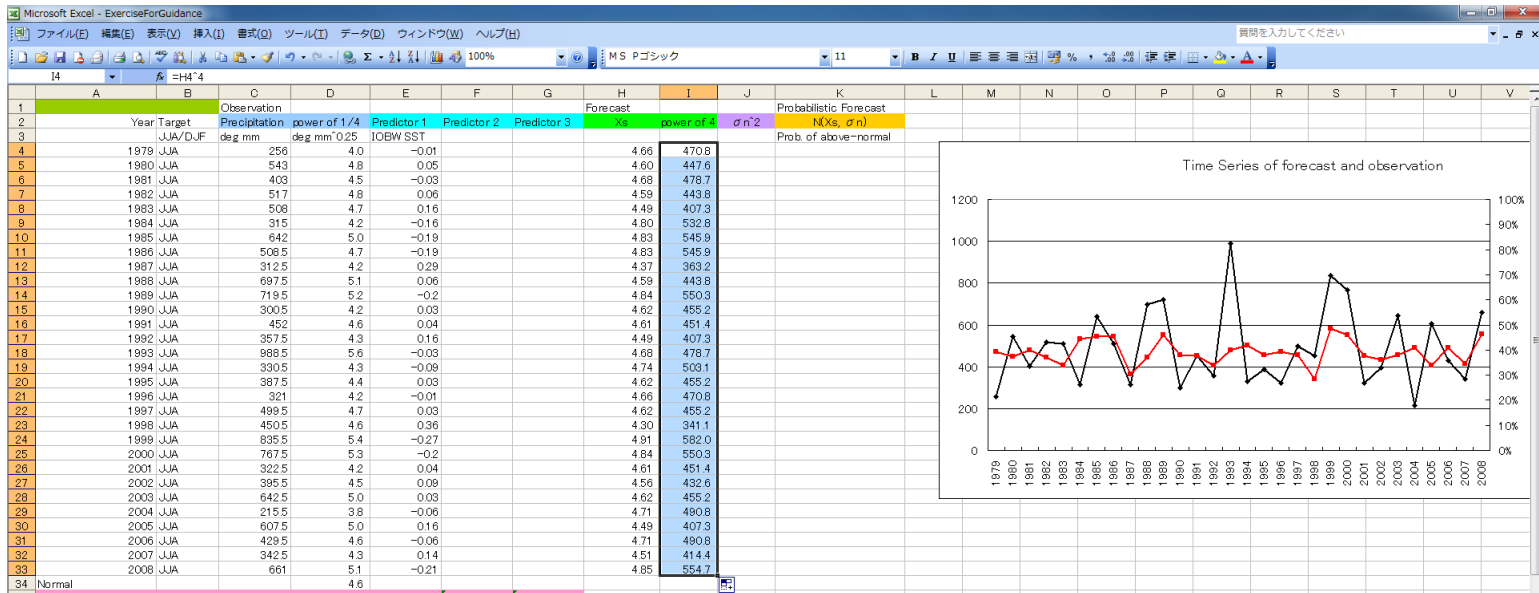
Normal										
Single Regression	slope	-0.97	#DIV/0!	#DIV/0!						
	intercept	4.65	#DIV/0!	#DIV/0!						
	Correlation	0.55	#DIV/0!	#DIV/0!						
Multi Regression	slope	#VALUE!	#VALUE!	#VALUE!						
	intercept	#VALUE!	#VALUE!	#VALUE!						
	Correlation	#DIV/0!	#DIV/0!	#DIV/0!						

Calculate the forecasts using single regression equation.

Input "=E\$35 * \$E4+\$E\$36" at H4.

1. Single Regression Model

- Calculate power of 4 on I line in case of precipitation.
- You can see a time series line chart. Red line indicates the forecasts.



Questions

1. What predictor do you select?
2. Can you get an accuracy guidance?
3. How does its guidance predict the hottest/coldest/drought/wet year in your country?

2. Multiple Regression Model

- Try to look for most effective combination of predictors.

The screenshot shows an Excel spreadsheet with the following data:

Year	Target	Precipitation	power of 1/4	Predictor 1	Predictor 2	Predictor 3	Forecast	power of 4	σn^2	Probabilistic Forecast
1979 JJA	256	4.0	-0.01	-1.61	-0.01	4.66	470.8			
1980 JJA	543	4.8	0.05	1.93	0.48	4.80	447.6			
1981 JJA	403	4.5	-0.03	-1.21	0.06	4.68	478.7			
1982 JJA	517	4.6	0.06	-1.32	0.79	4.59	443.8			
1983 JJA	508	4.7	0.16	2.59	1.95	4.49	407.3			
1984 JJA	315	4.2	-0.16	-6.63	-0.76	4.80	532.8			
1985 JJA	642	5.0	-0.19	-8.17	-0.77	4.83	545.9			
1986 JJA	508.5	4.7	-0.19	-5.93	0.06	4.83	545.9			
1987 JJA	312.5	4.2	0.29	2.64	1.42	4.37	363.2			
1988 JJA	697.5	5.1	0.06	-0.26	-1.87	4.57	455.2			
1988 JJA	719.5	5.2	-0.2	-4.87	-0.23	4.6	455.2			
1990 JJA	300.5	4.2	0.03	2.01	0.25	4.4	455.2			
1991 JJA	452	4.6	0.04	0.27	0.23	4.4	455.2			
1992 JJA	357.5	4.3	0.16	2.28	1.36	4.4	455.2			
1993 JJA	988.5	5.6	-0.03	-1.68	0.8	4.4	455.2			
1994 JJA	330.5	4.3	-0.09	-5.34	-0.37	4.4	455.2			
1995 JJA	387.5	4.4	0.03	0.31	-1.06	4.4	455.2			
1996 JJA	321	4.2	-0.01	-0.13	-0.44	4.4	455.2			
1997 JJA	499.5	4.7	0.03	4.24	1.93	4.62	455.2			
1998 JJA	450.5	4.6	0.36	9.07	-0.2	4.30	341.1			
1998 JJA	835.5	5.4	-0.27	-4.28	-1.23	4.91	562.0			
2000 JJA	767.5	5.3	-0.2	-2.64	-0.42	4.84	550.3			
2001 JJA	322.5	4.2	0.04	1.73	-0.06	4.61	451.4			
2002 JJA	395.5	4.5	0.09	4.65	-0.01	4.56	432.6			
2003 JJA	642.5	5.0	0.03	0.58	-1.25	4.62	455.2			
2004 JJA	215.5	3.8	-0.06	2.19	-0.01	4.71	490.8			
2005 JJA	607.5	5.0	0.16	3.13	-0.03	4.49	407.3			
2006 JJA	429.5	4.6	-0.06	4.25	0.16	4.71	490.8			
2007 JJA	342.5	4.3	0.14	3.15	-0.76	4.51	414.4			
2008 JJA	661	5.1	-0.21	-0.85	0.02	4.85	554.7			

Summary statistics for Single Regression (rows 35-37):

slope	-0.97	-0.03	-0.07
intercept	4.65	4.65	4.65
Correlation	0.33	0.27	0.15

Summary statistics for Multi Regression (rows 39-41):

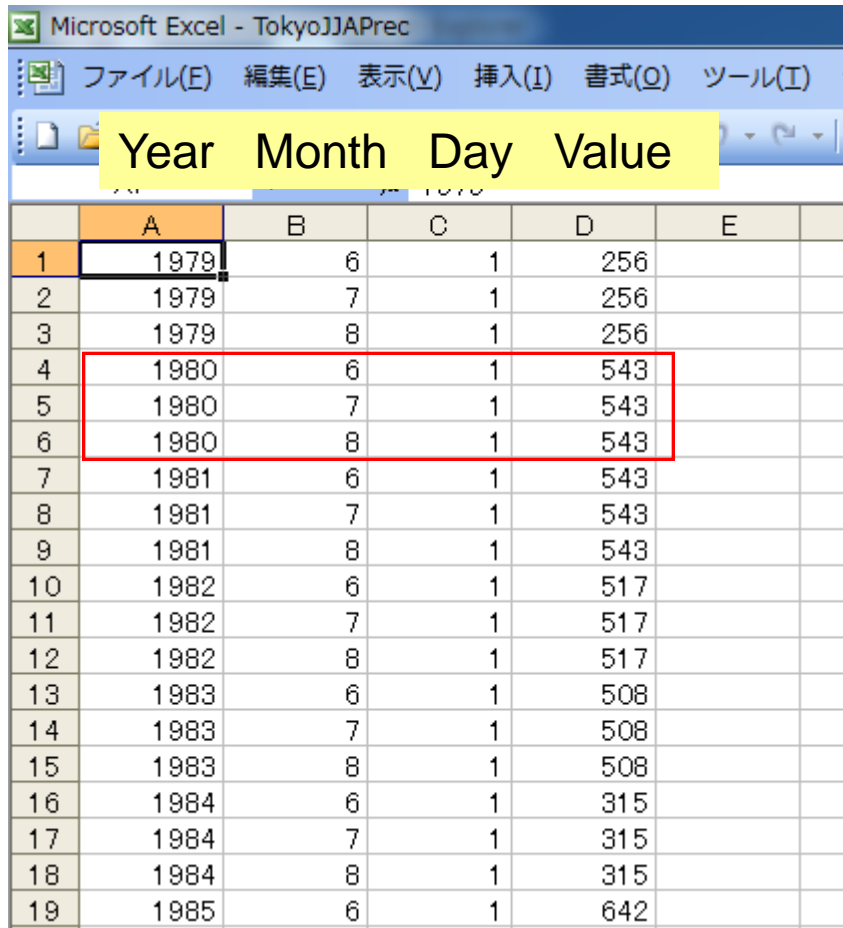
slope	-0.93	0.00	-0.01
intercept	4.65		
Correlation	0.38		

Select the most effective combination of predictors after trial and error.

The ITACS help you find them.

2. Multiple Regression Model

- Let's use ITACS for looking for effective predictors.
 - Preparing observation data for CSV format at first.



Microsoft Excel - TokyoJJAPrec

ファイル(E) 編集(E) 表示(V) 挿入(I) 書式(O) ツール(T) ...

Year Month Day Value

	A	B	C	D	E
1	1979	6	1	256	
2	1979	7	1	256	
3	1979	8	1	256	
4	1980	6	1	543	
5	1980	7	1	543	
6	1980	8	1	543	
7	1981	6	1	543	
8	1981	7	1	543	
9	1981	8	1	543	
10	1982	6	1	517	
11	1982	7	1	517	
12	1982	8	1	517	
13	1983	6	1	508	
14	1983	7	1	508	
15	1983	8	1	508	
16	1984	6	1	315	
17	1984	7	1	315	
18	1984	8	1	315	
19	1985	6	1	642	

In case of JJA, you need to input 6,7,8 on B line and JJA value on D line, respectively.

2. Multiple Regression Model

The screenshot shows the ITACS v3.0 web interface in Internet Explorer. The browser address bar shows the URL: `http://boreas.cpd.naps.kishou.go.jp/~fipx1/analyze3.0/index1.php?&dataset=SAT&element=olr&element_edit=0&dtype=ANOM&area=0,360,-90,90,DI&lat_s=-40&lat_e=40&lon_s=0&lon_e=360`. The page title is "ITACS v3.0".

The main content area is divided into two sections: "data1" and "data2".

data1

dataset	element	data type	area	level	average period	show period
SAT	OLR(W/m2)	ANOM	ALL Lat: -40 - 40 Ave <input type="checkbox"/> Lon: 0 - 360 Ave <input type="checkbox"/>	1000 hPa 1000 hPa	Year average Ave <input type="checkbox"/>	RANGE 1979 - 2008 06 - 08

analysis method : REGRESSION_COEFFICIENT

data2

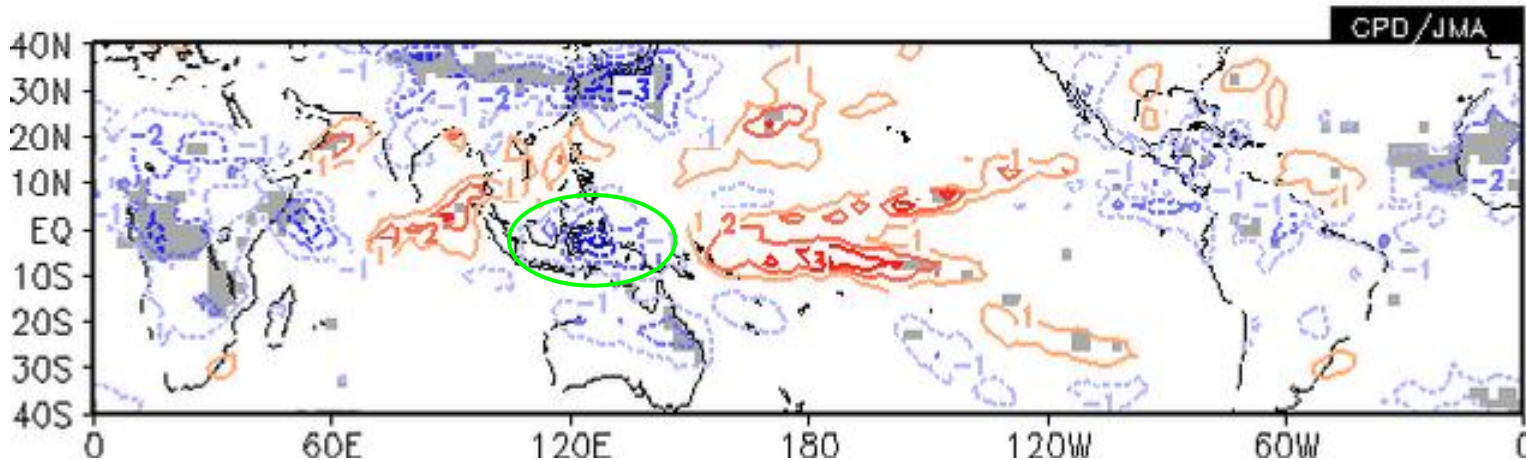
dataset	element	input txt	average period	lag
USER INPUT	UPLOAD TXT	¥¥¥¥yoho_tera¥¥SHARE¥¥person¥¥ito¥¥TCC研修2011¥¥TokyoJJAPrec.csv 1979,8,1,256 1979,7,1,256 1979,8,1,256 1980,8,1,543 1980,7,1,543 1980,8,1,543 1981,8,1,543 1981,7,1,543 1981,8,1,543 1982,8,1,517 1982,7,1,517 1982,8,1,517 1983,8,1,508 1983,7,1,508 1983,8,1,508 1984,8,1,315 1984,7,1,315 1984,8,1,315 1985,8,1,642 1985,7,1,642	Year average Ave <input type="checkbox"/>	0 YEAR

A yellow callout box points to the "analysis method" dropdown with the text: "Select REGRESSION_COEFFICIENT".

A yellow callout box points to the "input txt" text area with the text: "Upload your observation data".

2. Multiple Regression Model

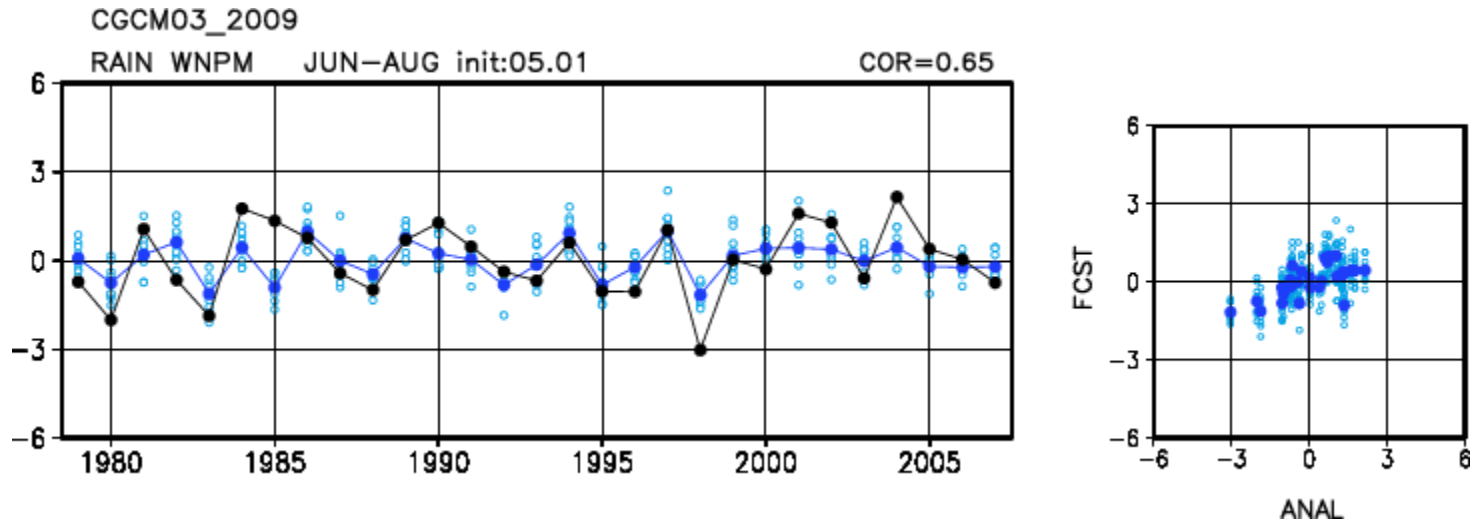
```
DATA1 SAT olr ANOM lat = -40:40 lon = 0:360 level = 1:1  
time = 1979060100:2008080100 ave = 3MONTH  
DATA2 USER_INPUT USER_INPUT1 HIST lat = -90:90 lon = 0:360 level = 1:1  
time = 1979060100:2008080100 ave = 3MONTH analysis method = REGRESSION_COEFFICIENT
```



This is a relationship between OLR and JJA precipitation in Tokyo. The OLR around Maritime Continent looks a likely predictor. Try looking for predictors in this way.

Note: ITACS can show the relationship between **analysis** and your observation data. If you use the variables as predictors, you need to confirm their forecast skill.

2. Multiple Regression Model



You can see the verification results of hindcast.

http://ds.data.jma.go.jp/tcc/tcc/gpv/model/hindcast_map/

2. Multiple Regression Model

- Calculate the forecasts using multiple regression equation.

Year	Target	Precipitation	power of 1/4	Predictor 1	Predictor 2	Predictor 3	Xs	power of 4	σn^2
1979	JJA	256	4.0	-0.01	-1.61	-0.03	4.67	473.7	
1980	JJA	543	4.8	0.05	1.93	-0.1	4.66	473.3	
1981	JJA	403	4.5	-0.03	-1.21	0.04	4.65	467.9	
1982	JJA	517	4.8	0.06	-1.32	-0.17	4.67	474.5	
1983	JJA	508	4.7	0.16	2.59	-0.14	4.56	433.3	
1984	JJA	315	4.2	-0.16	-6.63	0.05	4.76	511.3	
1985	JJA	642	5.0	-0.19	-8.17	0	4.81	534.4	
1986	JJA	508.5	4.7	-0.19	-5.93	-0.02	4.84	548.0	
1987	JJA	312.5	4.2	0.29	2.64	-0.13	4.40	375.5	
1988	JJA	697.5	5.1	0.06	-0.26	0.07	4.53	422.5	
1989	JJA	719.5	5.2	-0.2	-4.87	0.01	4.84	549.2	
1990	JJA	300.5	4.2	0.03	2.01	0.01	4.62	457.0	
1991	JJA	452	4.6	0.04	0.27	-0.17	4.70	489.6	
1992	JJA	357.5	4.3	0.16	2.28	-0.23	4.61	452.9	
1993	JJA	988.5	5.6	-0.03	-1.68	-0.47	4.95	600.0	
1994	JJA	330.5	4.3	-0.09	-5.34	-0.05	4.74	505.5	
1995	JJA	387.5	4.4	0.03	0.31	0	4.62	453.8	
1996	JJA	321	4.2	-0.01	-0.13	0.17	4.56	431.9	
1997	JJA	499.5	4.7	0.03	4.24	-0.31	4.83	544.9	
1998	JJA	450.5	4.6	0.36	9.07	0.12	4.22	318.1	
1999	JJA	835.5	5.4	-0.27	-4.28	0.03	4.92	584.7	
2000	JJA	797.5	5.3	-0.2	-2.64	0.15	4.78	520.5	
2001	JJA	322.5	4.2	0.04	1.73	0.27	4.46	394.0	
2002	JJA	395.5	4.5	0.09	4.65	0.01	4.57	437.6	
2003	JJA	642.5	5.0	0.03	0.58	0.26	4.46	397.0	
2004	JJA	215.5	3.8	-0.06	2.19	0.06	4.69	484.2	
2005	JJA	607.5	5.0	0.16	3.13	0.19	4.37	365.1	
2006	JJA	429.5	4.6	-0.06	4.25	0.15	4.67	474.1	
2007	JJA	342.5	4.3	0.14	3.15	0.19	4.40	373.2	
2008	JJA	661	5.1	-0.21	-0.95	0.03	4.87	564.0	

Normal	slope	-0.97	-0.03	-0.50
Single Regression	intercept	4.65	4.65	4.65
	Correlation	0.33	0.27	0.19
Multi Regression	slope	-1.19	0.01	-0.59
	intercept	4.65		
	Correlation	0.48		

Calculate the forecasts using multiple regression equation.

Input `"= E39 * $E4 + F39 * $F4 + G39 * $G4 + E40"` at H4.

Multiple Anomaly Correlation Coefficient

Questions

1. What predictors do you select?
2. Can you get more accuracy guidance than single regression model?

3. Probabilistic Forecast

- Calculate square of regression error.
- Calculate root mean square error.

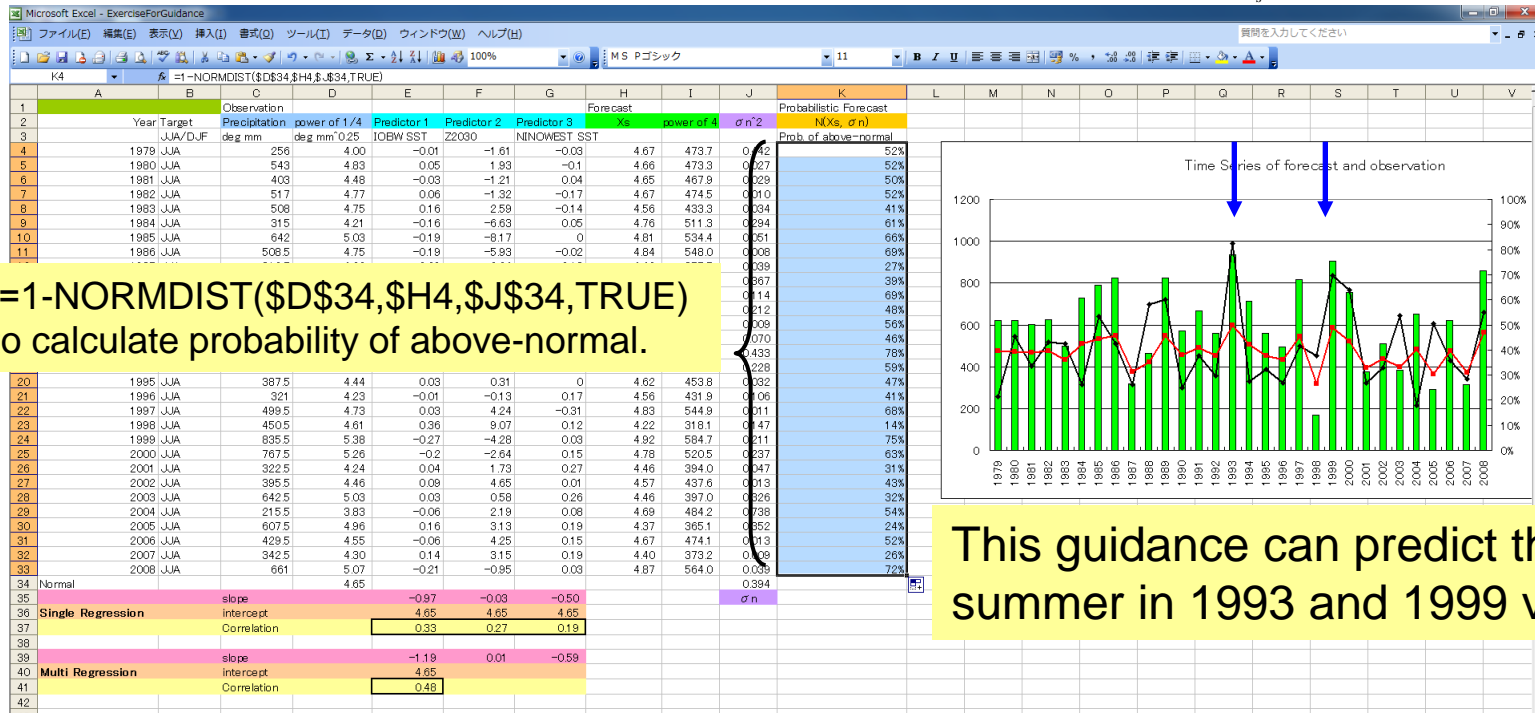
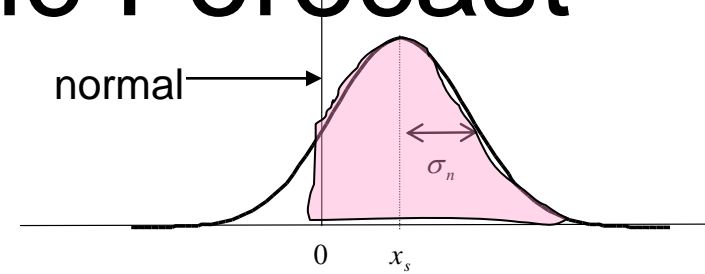
Year	Target	Observation	power of 1/4	Predictor 1	Predictor 2	Predictor 3	Forecast	power of 4	σn^2	Probabilistic Forecast
1978	JJA	256	4.00	-0.01	-1.61	-0.03	4.67	473.7	0.442	0.442
1980	JJA	543	4.83	0.05	1.93	-0.1	4.66	473.3	0.027	0.027
1981	JJA	403	4.48	-0.03	-1.21	0.04	4.65	467.9	0.029	0.029
1982	JJA	517	4.77	0.06	-1.32	-0.17	4.67	474.5	0.010	0.010
1983	JJA	508	4.75	0.16	2.59	-0.14	4.56	433.3	0.034	0.034
1984	JJA	315	4.21	-0.16	-6.63	0.05	4.76	511.3	0.294	0.294
1985	JJA	642	5.03	-0.19	-8.17	0	4.81	534.4	0.051	0.051
1986	JJA	508.5	4.75	-0.19	-5.93	-0.02	4.84	548.0	0.008	0.008
1987	JJA	312.5	4.20	0.29	2.64	-0.13	4.40	375.5	0.039	0.039
1988	JJA	697.5	5.14	0.06	-0.26	0.07	4.53	422.5	0.367	0.367
1988	JJA	719.5	5.18	-0.2	-4.87	0.01	4.84	549.2	0.114	0.114
1990	JJA	300.5	4.16	0.03	2.01	0.01	4.62	457.0	0.212	0.212
1991	JJA	452	4.61	0.04	0.27	-0.17	4.70	489.6	0.008	0.008
1992	JJA	357.5	4.35	0.16	2.28	-0.23	4.61	452.9	0.070	0.070
1993	JJA	988.5	5.61	-0.03	-1.68	-0.47	4.95	600.0	0.433	0.433
1994	JJA	330.5	4.26	-0.09	-5.34	-0.05	4.74	505.5	0.228	0.228
1995	JJA	387.5	4.44	0.03	0.31	0	4.62	453.8	0.032	0.032
1996	JJA	321	4.23	-0.01	-0.13	0.17	4.56	431.9	0.106	0.106
1997	JJA	499.5	4.73	0.03	4.24	-0.31	4.83	544.9	0.011	0.011
1998	JJA	450.5	4.61	0.36	9.07	0.12	4.22	318.1	0.147	0.147
1998	JJA	835.5	5.38	-0.27	-4.28	0.03	4.92	584.7	0.211	0.211
2000	JJA	767.5	5.26	-0.2	-2.64	0.15	4.78	520.5	0.237	0.237
2001	JJA	322.5	4.24	0.04	1.73	0.27	4.46	394.0	0.047	0.047
2002	JJA	395.5	4.46	0.09	4.65	0.01	4.57	437.6	0.013	0.013
2003	JJA	642.5	5.03	0.03	0.58	0.26	4.46	397.0	0.326	0.326
2004	JJA	215.5	3.83	-0.06	2.19	0.08	4.69	484.2	0.738	0.738
2005	JJA	607.5	4.96	0.16	3.13	0.19	4.37	365.1	0.352	0.352
2006	JJA	429.5	4.55	-0.06	4.25	0.15	4.67	474.1	0.013	0.013
2007	JJA	342.5	4.30	0.14	3.15	0.19	4.40	373.2	0.009	0.009
2008	JJA	661	5.07	-0.21	-0.85	0.03	4.87	564.0	0.039	0.039
Normal			4.65						0.394	
Single Regression	slope			-0.97	-0.03	-0.50			σn	
Single Regression	intercept			4.65	4.65	4.65				
Single Regression	Correlation			0.33	0.27	0.19				
Multi Regression	slope			-1.19	0.01	-0.59				
Multi Regression	intercept			4.65						
Multi Regression	Correlation			0.48						

Input “ $=($H4-$D4)^2$ ” at J4 to calculate square error.

Input “ $=SQRT(AVERAGE(J4:J33))$ ” at J34 to calculate root mean square error.

3. Probabilistic Forecast

- Calculate probability of above-normal.



Input `"=1-NORMDIST(D34,$H4,$J$34,TRUE)` at K4 to calculate probability of above-normal.

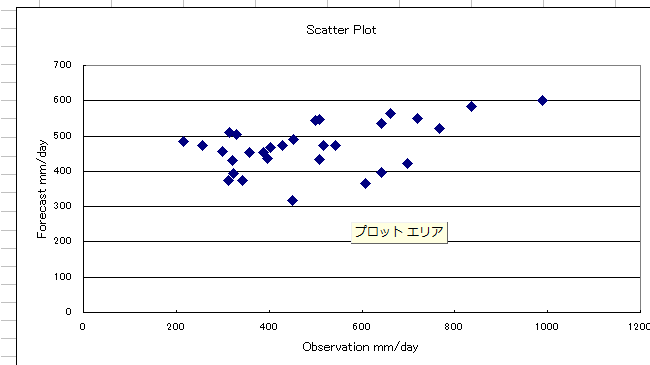
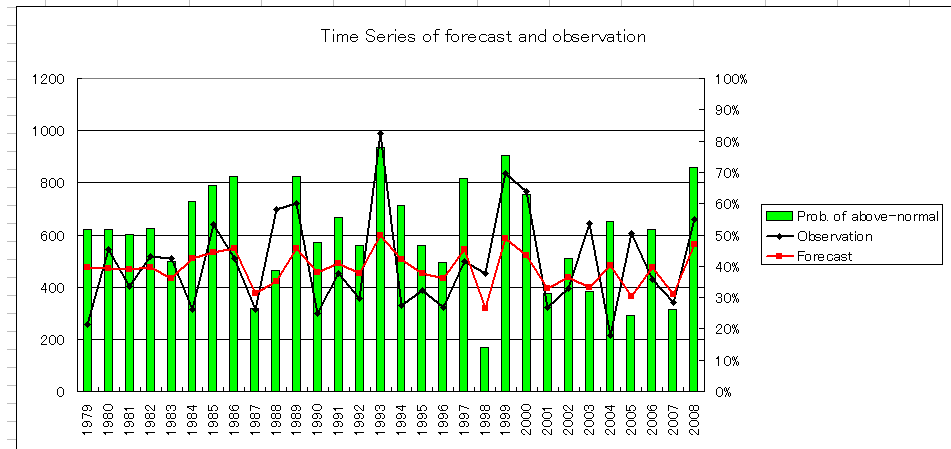
This guidance can predict the wet summer in 1993 and 1999 very well.

Question

- What is the difference of probability between temperature and precipitation?
- Have you made some guidance for temp/prec to do probabilistic verification?

4-1. Deterministic Verification

- Anomaly Correlation
- Root Mean Square Error
- Time series chart
- Scatter plot



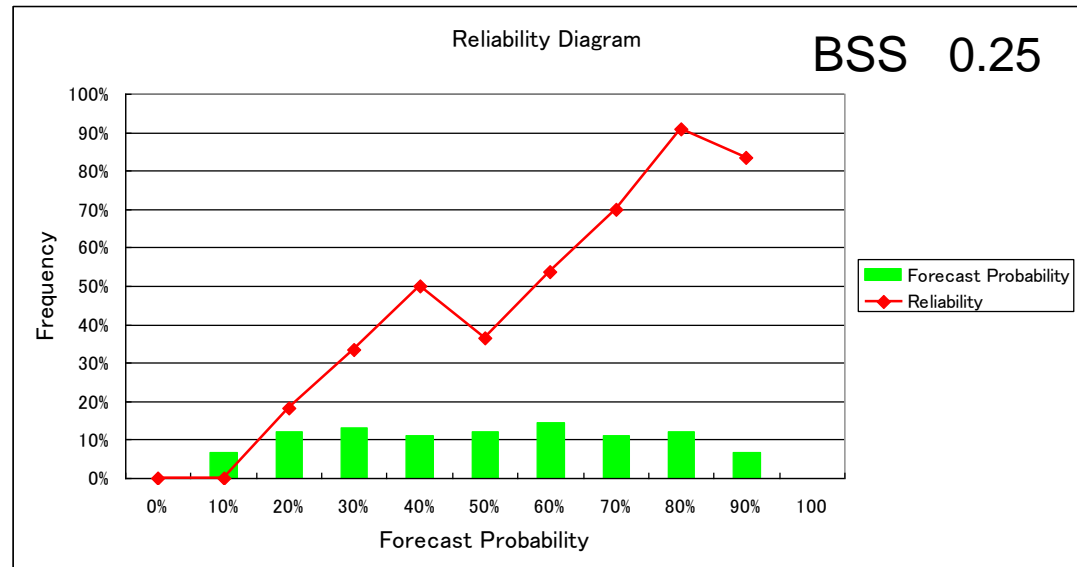
4-2. Probabilistic Verification

- Brier Skill Score (BSS)
- Reliability Diagram

$$b = \frac{1}{N} \sum_{i=1}^N (p_i - v_i), \quad 0 \leq p_i \leq 1, \quad v_i \in \{0,1\}$$

$$BSS = \frac{bc - b}{bc}$$

See the textbook in detail



Skill of JMA Routine Guidance from 2005 to 2009.

Target Period	BSS
1 st weekly temp.	0.45
2 nd weekly temp.	0.25
1 st monthly temp.	0.30
3-month temp.	0.35

Reliability Diagram of JJA temperature, totaling Tokyo, Sapporo and Naha station in Japan.



5. Create a Presentation

An example

- What predictors do you use?
- Why do you select their predictors?
 - By ITACS regression tools, traditional, ...
- Verification results

Create a presentation with your originality.

Summary of Japanese guidance

Station	Tokyo
Season	JJA
Predictand	Temperature
Predictors	WNP RAIN, Extratropical Thickness, Indian Ocean SST
Correlation	0.52
Brier Skill Score	0.28 (including Sapporo and Naha stations)

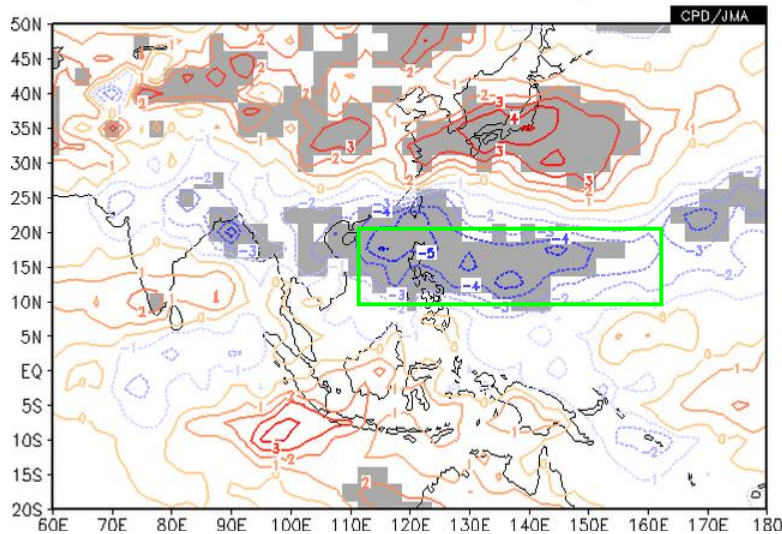


	slope	0.63	1.75	0.33
Multiple Regression	intercept	24.99		
	Correlation	0.52		

The reason I select the predictors

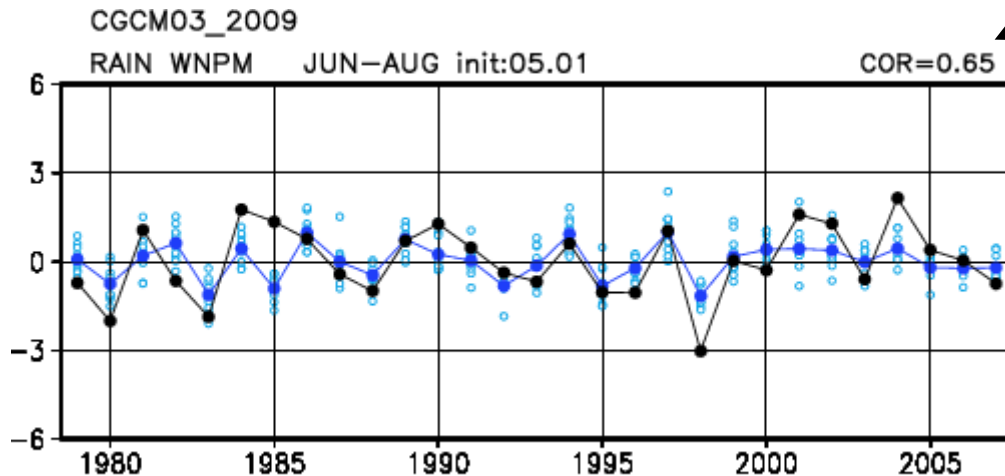
DATA1 SAT_clr ANOM lat = -20:50 lon = 60:180 level = 1:1
time = 1979060100:2008080100 ave = 3MONTH

DATA2 USER_INPUT USER_INPUT1 HIST lat = -90:90 lon = 0:360 level = 1:1
time = 1979060100:2008080100 ave = 3MONTH analysis method = REGRESSION_COEFFICIENT

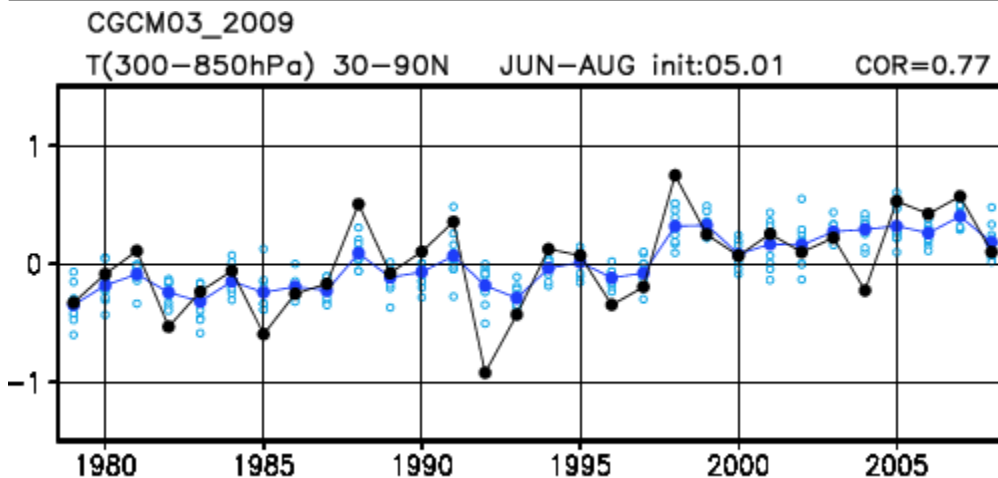
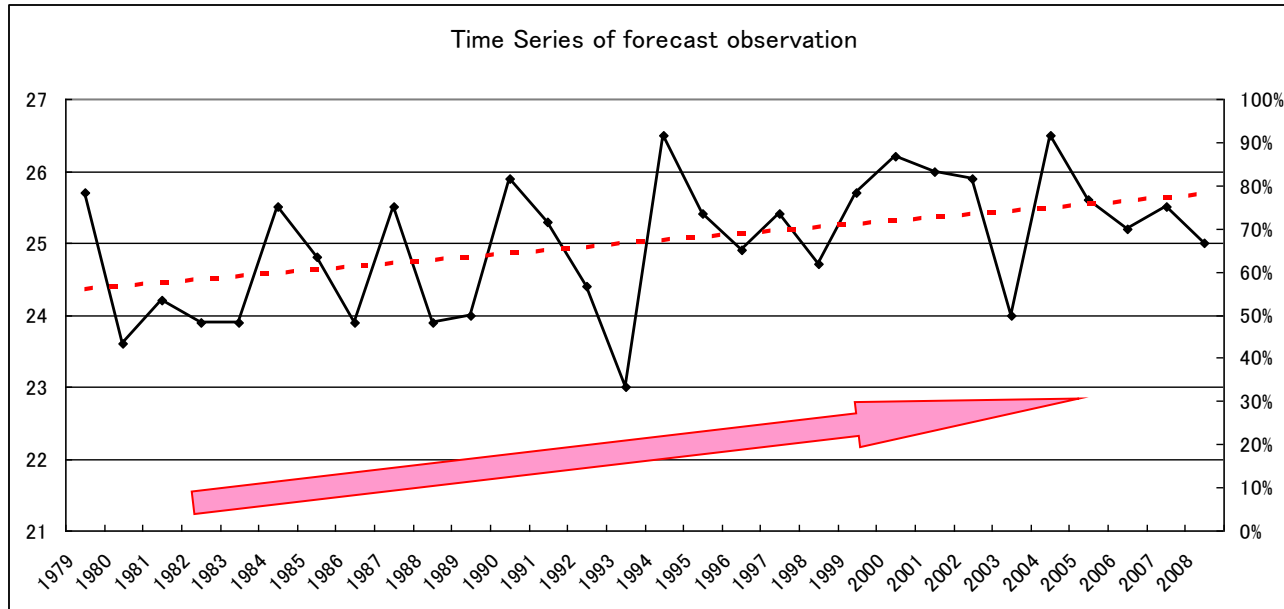


Western North Pacific OLR is deeply associated with JJA temperature in Tokyo.

And the skill of WNP RAIN is relatively high, so the predictor may be useful.

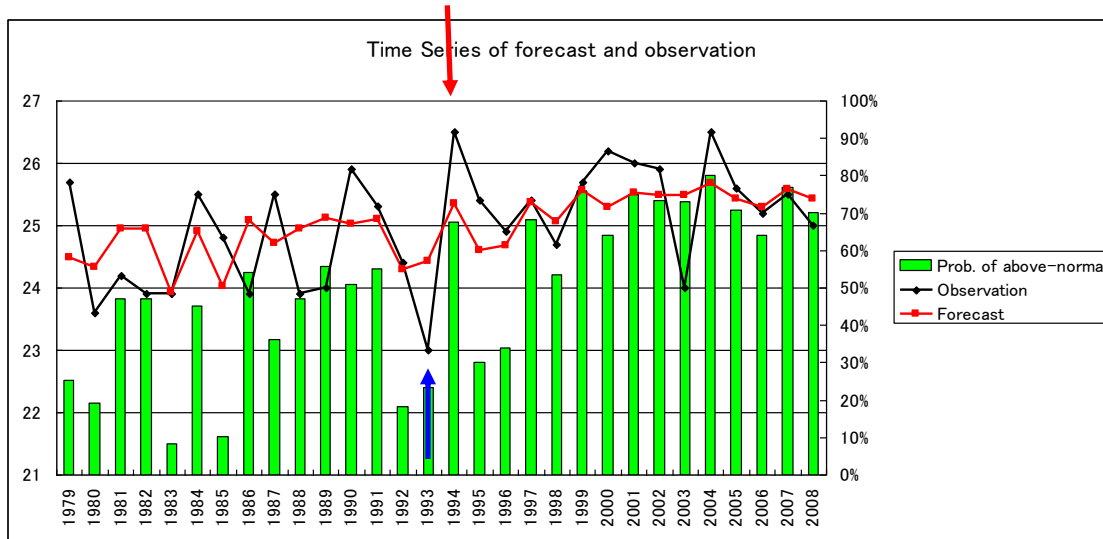


The reason I select the predictors



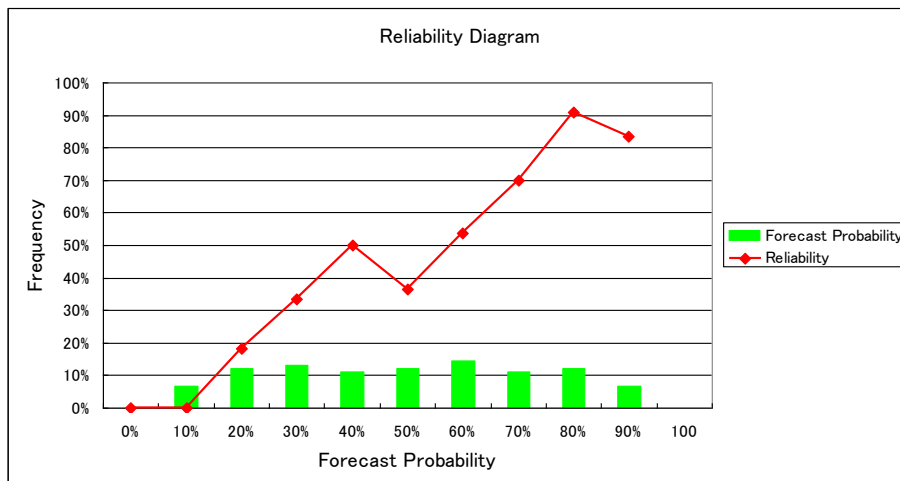
The JJA temperature have upward trend, so I use extratropical thickness as global warming predictor.

Verifications



A clear upward trend is well predicted. Moreover, the guidance can predict extreme cool/hot summer in 1993/1994.

It is important to be able to predict cool summer. Because, extreme cool summer can cause severe damage of crop production in Japan.



The figure shows reliability diagram of JJA temperature, totaling Tokyo, Sapporo and Naha station in Japan. The forecast probability is reliable to some degree.



Thank you