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CONDITIONAL CLIMATOLOGY GIVEN THE LOCAL OBSERVATION

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GEM

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Editor's Note: Some of the principles that I have followed in selecting articles to be published in the Digest are that they be practical and written so that they appeal to the majority of the readership of the Digest. I have long been familiar with Dr. Miller's work in the Techniques Development Laboratory, NWS, and have often asked Bob to write an article for the Digest because, I felt, his work would greatly add to the advancement of practical meteorology. Dr. Miller's GEMTRIX is a novel, powerful idea, yet easy to understand and can be adapted to a forecasting scheme, particularly when applied to AFOS. The Digest thanks Dr. Miller for his article and invites your comments on it.

Randy Racer
Managing Editor

Brief Biographical Sketch

The author is Chief of the Objective Forecast Branch in the National Weather Service's Techniques Development Laboratory. He has spent almost thirty years developing and applying statistical methods to weather forecasting. His career began as a World War II weather observer serving in India and Burma. Prior to receiving his M.S. in meteorology from New York University he was an observer on ocean station vessels. His research work began at Massachusetts Institute of Technology and continued at the Travelers Research Center, Bell Telephone Laboratories, and the Life Insurance Marketing and Research Association. Prior to joining TDL he was Chief Scientist of the Air Weather Service. He holds a Ph.D. from Harvard in statistics and is the author of numerous technical papers including an AMS Meteorological Monograph entitled "Statistical Prediction by Discriminant Analysis".

INTRODUCTION

Classical climatology gives an average picture of the weather. It can be made more and more specific by conditioning on the basis of time and

location. The ultimate conditional climatology would be the one that specifies the probability distribution of all elements everywhere, conditioned on the full state of the atmosphere both past and present. A more practical conditional climatology would be one that takes into account the set of all locally observed meteorological variables to arrive at future conditional probability distributions. This paper will demonstrate the application of such a conditional climatology called GEM, an acronym for Generalized Equivalent Markov model.

METHODOLOGY

The approach will be to estimate the conditional probability distribution of every weather element one hour into the future, given the present condition of those same weather elements. Estimation of the probabilities is performed by multiple linear regression equations. To accomplish this, historical observations are used in the regression analysis. These are schematically represented in Table 1.

The X's and Y's shown in Table 1 represent, for example, month of the year, hour of the day, temperature, ceiling height, visibility, weather, cloud cover, and so forth.

Next, the observations of the X's and Y's are transformed into intervals (or categories). Each is assigned a 1.0 or 0.0 depending, respectively, upon whether that observation is in or out of the interval (or category). This transformation process serves two purposes. First, the regression equations will then actually estimate the desired probabilities (see Miller 1964). Second, the method to extend the conditional probability-distribution estimates beyond one hour is greatly facilitated. Consequently, each column in Table 1 expands into two or more columns which are mutually exclusive and exhaustive. A complete list of the transformations in this paper is given in Table 2.

Table 1. Schematic representation of an historical data sample of N observations, for the initial weather-variable conditions $X_{n,p}$ and the subsequent weather-variable conditions $Y_{n,p}$ where the subscript n denotes the general observation and the subscript p denotes the general variable.

Observation	Initial Weather Variables			Subsequent Weather Variables				
	X_1	X_2	X_p	Y_1	Y_2	Y_p		
1	$x_{1,1}$	$x_{1,2}$	\dots	$x_{1,p}$	$y_{1,2}$	$y_{1,2}$	\dots	$y_{1,p}$
2	$x_{2,1}$	$x_{2,2}$	\dots	$x_{2,p}$	$y_{2,1}$	$y_{2,2}$	\dots	$y_{2,p}$
.								
.								
.								
n	$x_{n,1}$	$x_{n,2}$	\dots	$x_{n,p}$	$y_{n,1}$	$y_{n,2}$	\dots	$y_{n,p}$
.								
.								
.								
N	$x_{N,1}$	$x_{N,2}$	\dots	$x_{N,p}$	$y_{N,1}$	$y_{N,2}$	\dots	$y_{N,p}$

A 10-year historical sample of data, consisting of 86,782 hourly observations for Washington National Airport (DCA), was organized in the fashion described above. Five years (the odd numbered years) were held aside for subsequent independent tests. The 5 even numbered years became the sample from which the regression equations were developed. The resulting 175 regression equations constitute a matrix of 175 columns and rows. For demonstration purposes, the set of all cloud amount equations is given in Table 3, where the 175 predictors are those identified in Table 2.

To appreciate the effect any variable has on, say, the overcast event, look at the coefficient of that variable. Since the variables are all either 1.0 or 0.0, the coefficient amount is either added (algebraically) or not added. Thus, if the observation is for January (variable number 2), then .02676120 is added to the conditional probability that overcast will be the observed condition one hour hence, all other things held constant. Since one of the other months must be a 1.0 if January is not "on", the situation where it is really August (variable number 9) causes -.01619480 to be added algebraically. This is a swing of .04295600 in the conditional probability of overcast.

Given an observation of all the weather elements at DCA at a particular time, the set of 175 regression equations is then used to estimate the 175 conditional probabilities one hour hence. To arrive at estimates of the 175 conditional probabilities two hours hence the original observation is replaced by the conditional probabilities estimated for one hour. This process of replacing the observation with the previous estimate may be continued ad infinitum. Obviously, the estimates of the conditional probabilities will reflect the fact that only seasonal and diurnal effects are of importance out beyond a certain period of time. The initial observation will eventually lose all of its influence. The process of iteration is referred to as the equivalent Markov process (see Miller 1968 and Whiton 1977).

Another important requirement of a useful conditional climatology is that it be convertible into "best" categorical estimates. This suggests applying a thresholding method that yields certain desired effects, such as producing categorical frequencies that are close to observed frequencies. A method developed by Miller and Best (1979) has been employed here, where the necessary parameters are the readily available multiple correlation coefficients and sample climatologies.

Table 2. Each X and Y in Table 1 has been transformed into the intervals and categories shown in this Table. Each element has been numbered for future reference. The number one element is always unity.

2	MONTH (LOCAL)	JAN	51	13 to 14
3		FEB	52	15 to 17
4		MAR	53	18 to 22
5		APR	54	23 to 27
6		MAY	55	28 to 32
7		JUN	56	33 to 45
8		JUL	57	SLP (MB) 980.1 to 990.0
9		AUG	58	990.1 to 1000.0
10		SEP	59	1000.1 to 1005.0
11		OCT	60	1005.1 to 1010.0
12		NOV	61	1010.1 to 1020.0
13		DEC	62	1020.1 to 1030.0
14	HOUR (LOCAL)	00	63	1030.1 to 1090.0
15		01	64	DB TEMP (DEG F) -99 to 14
16		02	65	15 to 23
17		03	66	24 to 27
18		04	67	28 to 29
19		05	68	30 to 31
20		06	69	32
21		07	70	33 to 34
22		08	71	35 to 36
23		09	72	37 to 38
24		10	73	39 to 40
25		11	74	41 to 42
26		12	75	43 to 50
27		13	76	51 to 59
28		14	77	60 to 68
29		15	78	69 to 77
30		16	79	78 to 86
31		17	80	87 to 140
32		18	81	DPT DEPRESSION 0
33		19	82	(DEG F) 1
34		20	83	2 to 4
35		21	84	5 to 6
36		22	85	7 to 11
37		23	86	12 to 15
38	WIND DIRECTION	CALM	87	16 to 99
39		NNE to NE	88	SKY COVER CLR
40		ENE to E	89	SCD
41		ESE to SE	90	BKN
42		SSE to S	91	OVC
43		SSW to SW	92	VISIBILITY .00 to .49
44		WSW to W	93	(MI) .50 to .74
45		WNW to NW	94	.75 to .99
46		NNW to N	95	1.0 to 1.49
47	WIND SPEED	0 to 3	96	1.5 to 1.99
48	(KT)	4 to 7	97	2.0 to 2.49
49		8 to 9	98	2.5 to 2.99
50		10 to 12	99	3.0 to 3.99

100	VISIBILITY (CONT)	4.0 to 4.99	154	-19 to -10
101		5.0 to 5.99	155	-9 to 0
102		6.0 to 6.99	156	1 to 9
103		7.0 to 99.00	157	10 to 19
104	WEATHER	NO WX	158	20 to 29
105		WX	159	30 to 39
106		NO F,IF	160	40 to 900
107		F,IF	161	CIG (100's FT) 0 to 1
108		NO GF	162	2 to 4
109		GF	163	5 to 6
110		NO K,H,D,KH,KD,HD,KHD	164	7 to 9
111		K,H,D,KH,KD,HD,KHD	165	10 to 14
112		NO BS,BD,BN,BY	166	15 to 19
113		BS,BD,BN,BY	167	20 to 24
114		NO L	168	25 to 29
115		L-	169	30 to 39
116		NO R	170	40 to 49
117		R-	171	50 to 59
118		R	172	60 to 74
119		R+	173	75 to 99
120		NO RW	174	100 to 150
121		RW-	175	151 to UNL
122		RW		
123		RW+		
124		NO S,IC		
125		S-,IC-		
126		S,IC		
127		S+,IC+		
128		NO SW,IP,SP		
129		SW-,IP-,SP-		
130		SW,IP,SP		
131		NO ZL		
132		ZL-		
133		NO ZR		
134		ZR-		
135		NO T,A-		
136		T,A-		
137	CLOUD COVER	CLR < 200 FT		
138		OVC < 200 FT		
139		CLR < 500 FT		
140		SCD < 500 FT		
141		BKN < 500 FT		
142		OVC < 500 FT		
143		CLR < 1000 FT		
144		SCD < 1000 FT		
145		BKN < 1000 FT		
146		OVC < 1000 FT		
147		CLR < 3000 FT		
148		SCD < 3000 FT		
149		BKN < 3000 FT		
150		OVC < 3000 FT		
151	PRES CHG (TENTHS OF MB)	-900 to -40		
152		-39 to -30		
153		-29 to -20		

Table 3. Regression equations for the four cloud amount categories. The number to the left of each regression coefficient corresponds to the predictor number provided in Table 2.

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VERIFICATION RESULTS

Applying the methodology described above to the 5 years of independent data produced the verification statistics shown in Table 4 for 1-h conditional-probability estimates.* Additional verification tests have yielded favorable results for GEM at six other locations around the country, for both 1- and 6-h projections (see also Miller, Whiton, and Kelly 1977).

Table 4. Comparative verification for six ceiling and visibility categories between classical climatology, conditional persistence, and the GEM conditional climatology model. Shown are the Brier scores for 43,047 1-h independent projections (lower values are better).

	Brier Scores	
	<u>Ceiling</u>	<u>Visibility</u>
Classical Climatology	.43709	.36571
Conditional Persistence	.20579	.19006
GEM conditional climatology	.19249	.17240

where the six categories are defined as:

<u>Ceiling</u>	<u>Visibility</u>
1. < 200'	1. < 1/2 mile
2. 200' ≤ to < 500'	2. 1/2 mile ≤ to < 1 mile
3. 500' ≤ to < 1000'	3. 1 mile ≤ to < 3 miles
4. 1000' ≤ to < 3000'	4. 3 miles ≤ to < 5 miles
5. 3000' ≤ to < 7500'	5. 5 miles ≤ to < 7 miles
6. ≥ 7500'	6. ≥ 7 miles

AN EXAMPLE

The derivation of a particular conditional climatology given the local observation will normally be performed automatically in a minicomputer such as an AFOS Eclipse. For demonstration purposes, an application of the methodology described above will now be made for a particular DCA observation. The steps performed are given below and in the flowchart shown in Figure 1.

Step 1: Collect input observation and transform into 1.0's or 0.0's depending upon observed interval or category.

Step 2: Estimate the 1-h conditional probabilities for each element using regression equations. Iterate out hour by hour to twenty-four hours using equivalent Markov principle, thereby creating a GEMTRIX (A Generalized Equivalent Markov Probability Matrix).

Step 3: Convert the probabilities in GEMTRIX into "best" hourly categorical estimates

*Classical climatology is defined as the dependent sample relative frequencies, and conditional persistence is represented by the dependent sample relative frequencies conditioned on the currently observed ceiling or visibility category.

using threshold probabilities, thereby creating a CATRIX (A GEMTRIX converted to a Categorical matrix).

Step 4: Assume hourly categorical estimates to be what will actually be observed and create an FT either subjectively or objectively.

Step 5: Forecaster than accepts or modifies CATRIX through feedback after studying FT and additional current weather information. A final FT is ultimately issued.

PERFORMING STEP 1:

The DCA observation chosen for this example is for 1600 local standard time (2100 GMT) on June 17, 1979 and was as follows:

Month : June

Hour : 1600 LST

Wind Direction : 330°

Wind Speed : 10 Kt

Sea Level Pressure : 1011.2 mb

Temperature : 74°F

Dew Point Depression ($T-T_d$) : 1°F

Sky Cover : Overcast

Visibility : 7 mi

Weather : None

Cloud Cover : Overcast under 3000'

Three Hour Pressure Change : -.7 mb

Ceiling : 1100 ft

The corresponding transformed observed values are shown in the first column of the GEMTRIX of Table 5.

PERFORMING STEP 2:

The estimated conditional probability distributions for all weather elements, projected from one to 24 hours, are given in Table 5 in the form of a GEMTRIX. The probabilities have been truncated and renormalized when necessary to have them lie between 0.0 and 1.0 and their sums equal to 1.0. In Table 5 the actual observed interval or category is parenthesized, when not obvious, for purposes of visual verification.

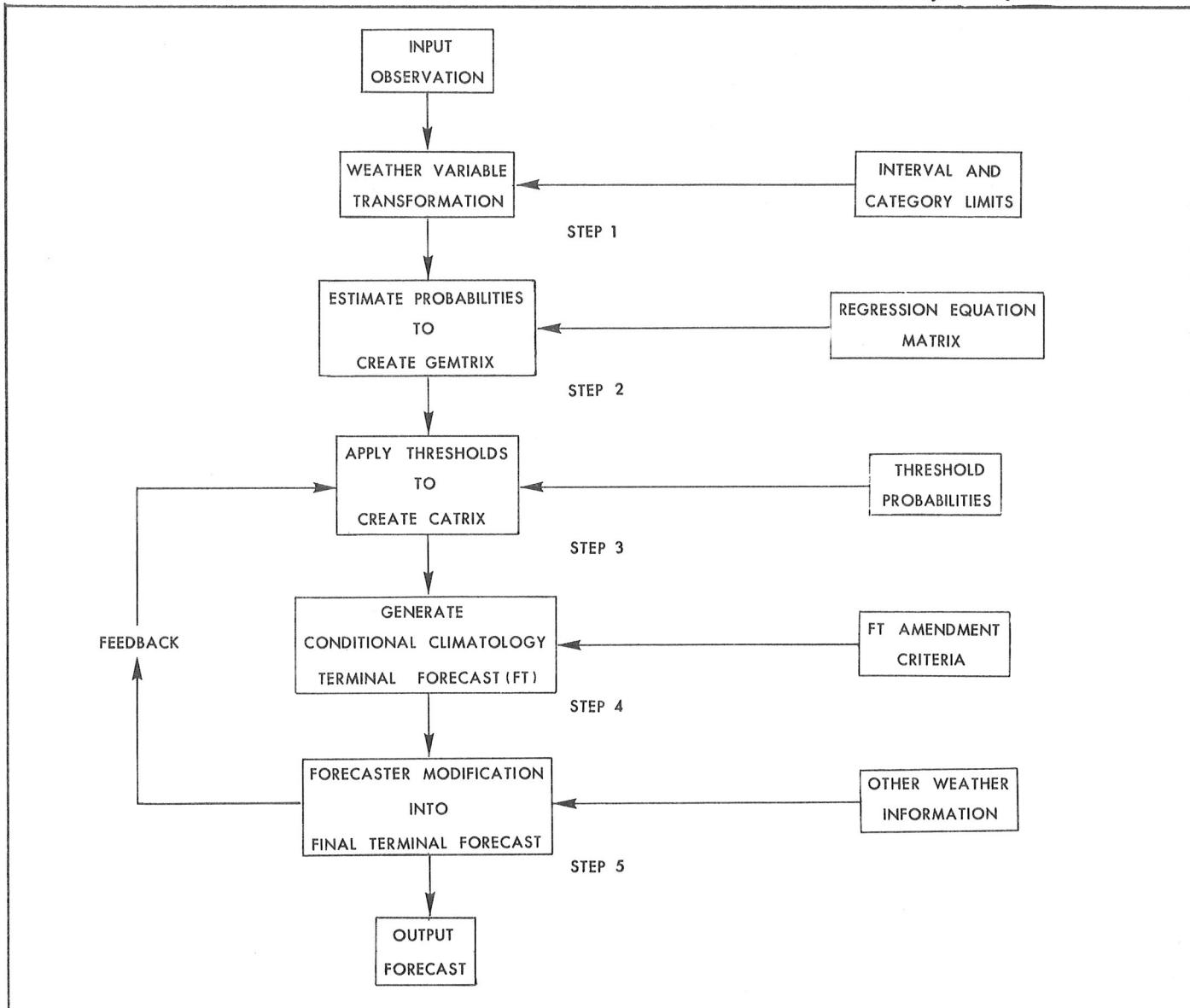


Figure 1. Flowchart of steps to follow in applying GEM to making an aviation terminal forecast.

PERFORMING STEP 3:

Table 6 gives the CATRIX for each weather element, again from one to 24 hours. The numbers in the CATRIX represent the "best" category. These are identified by using Table 2, starting with one if the first category, two if the second category, etc. for each weather element. Thus, a 4 for SKY COVER at hour 23 LST means the fourth category of SKY COVER or OVC.

PERFORMING STEP 4:

Assume the categories shown in the CATRIX will actually occur. A minicomputer algorithm could be written to produce an FT using amendment criteria. However, for this example two experienced forecasters generated the climatological FT shown below from only the information contained in the CATRIX of Table 6.

FORECASTER'S CLIMATOLOGICAL FT

DCA FT 8 SCT C12 OVC 3410.

05Z 10 SCT C15 OVC 6 H.

11Z C15 OVC 3FH.

17Z C15 BKN 5H BKN V SCT 1010.

21Z MVFR CIG BCMG VFR.

NNNN

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GENERALIZED EQUIVALENT MARKOV (GEMTRIX)
TECHNIQUES DEVELOPMENT LABORATORY

FOR STATION "DCA"

VALID FROM 790617Z TO 790618Z
BASED ON 79061716 SINGLE STATION SURFACE OBSERVATION

CATEGORY	OBS	24 HOURLY FORECASTS AFTER OBSERVATION (LOCAL STANDARD TIME)																							
		16	17	18	19	20	21	22	23	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
HOUR (LOCAL)																									
MONTH (LOCAL)	HEB																								
MONTH (LOCAL)	MAR																								
MONTH (LOCAL)	JUN																								
MONTH (LOCAL)	JUL																								
MONTH (LOCAL)	SEP																								
MONTH (LOCAL)	OCT																								
MONTH (LOCAL)	DEC																								
HOUR (LOCAL)	10																								
HOUR (LOCAL)	11																								
HOUR (LOCAL)	12																								
HOUR (LOCAL)	13																								
HOUR (LOCAL)	14																								
HOUR (LOCAL)	15																								
HOUR (LOCAL)	16																								
HOUR (LOCAL)	17																								
HOUR (LOCAL)	18																								
HOUR (LOCAL)	19																								
HOUR (LOCAL)	20																								
HOUR (LOCAL)	21																								
HOUR (LOCAL)	22																								
HOUR (LOCAL)	23																								
WIND DTR	NNW	NE																							
WIND DTR	NNW	NE																							
WIND DTR	SSW	TOD																							
WIND DTR	SSW	TOD																							
WIND DTR	NNN	TOD																							
WIND DTR	NNN	TOD																							
WIND SPD	(KTS)	5	7	10	12	15	17	19	21	23	25	27	29	31	33	35	37	39	41	43	45	47	49	51	53
WIND SPD	(KTS)	5	7	10	12	15	17	19	21	23	25	27	29	31	33	35	37	39	41	43	45	47	49	51	53
WIND SPD	(KTS)	5	7	10	12	15	17	19	21	23	25	27	29	31	33	35	37	39	41	43	45	47	49	51	53
WIND SPD	(KTS)	5	7	10	12	15	17	19	21	23	25	27	29	31	33	35	37	39	41	43	45	47	49	51	53
WIND SPD	(KTS)	5	7	10	12	15	17	19	21	23	25	27	29	31	33	35	37	39	41	43	45	47	49	51	53
WIND SPD	(KTS)	5	7	10	12	15	17	19	21	23	25	27	29	31	33	35	37	39	41	43	45	47	49	51	53
SLP	MB	992.1	992.2	992.3	992.4	992.5	992.6	992.7	992.8	992.9	993.0	993.1	993.2	993.3	993.4	993.5	993.6	993.7	993.8	993.9	994.0	994.1	994.2	994.3	
SLP	MB	992.1	992.2	992.3	992.4	992.5	992.6	992.7	992.8	992.9	993.0	993.1	993.2	993.3	993.4	993.5	993.6	993.7	993.8	993.9	994.0	994.1	994.2	994.3	
SLP	MB	992.1	992.2	992.3	992.4	992.5	992.6	992.7	992.8	992.9	993.0	993.1	993.2	993.3	993.4	993.5	993.6	993.7	993.8	993.9	994.0	994.1	994.2	994.3	
SLP	MB	992.1	992.2	992.3	992.4	992.5	992.6	992.7	992.8	992.9	993.0	993.1	993.2	993.3	993.4	993.5	993.6	993.7	993.8	993.9	994.0	994.1	994.2	994.3	
SLP	MB	992.1	992.2	992.3	992.4	992.5	992.6	992.7	992.8	992.9	993.0	993.1	993.2	993.3	993.4	993.5	993.6	993.7	993.8	993.9	994.0	994.1	994.2	994.3	
DPT TEMP	(DEG F)	59	60	61	62	63	64	65	66	67	68	69	70	71	72	73	74	75	76	77	78	79	80	81	82
DPT TEMP	(DEG F)	59	60	61	62	63	64	65	66	67	68	69	70	71	72	73	74	75	76	77	78	79	80	81	82
DPT TEMP	(DEG F)	59	60	61	62	63	64	65	66	67	68	69	70	71	72	73	74	75	76	77	78	79	80	81	82
DPT TEMP	(DEG F)	59	60	61	62	63	64	65	66	67	68	69	70	71	72	73	74	75	76	77	78	79	80	81	82
DPT TEMP	(DEG F)	59	60	61	62	63	64	65	66	67	68	69	70	71	72	73	74	75	76	77	78	79	80	81	82
DPT TEMP	(DEG F)	59	60	61	62	63	64	65	66	67	68	69	70	71	72	73	74	75	76	77	78	79	80	81	82
DPT DPM	INCHES	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	
DPT DPM	INCHES	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	
DPT DPM	INCHES	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	
DPT DPM	INCHES	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	
Sky EVN	SCD																								
Sky EVN	CVR																								
VSBY ST	M1	0.00	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32
VSBY ST	M1	0.00	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32
VSBY ST	M1	0.00	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32
VSBY ST	M1	0.00	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32
VSBY ST	M1	0.00	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32
VSBY ST	M1	0.00	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32
WS W	MPH	0.00	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34
WS W	MPH	0.00	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34
WS W	MPH	0.00	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34
WS W	MPH	0.00	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34
WS W	MPH	0.00	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34
WS W	MPH	0.00	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34
WS W	MPH	0.00	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31	32	33	34
WS W	MPH	0.00	12	13	14	15	16	17	18	1															

XXXX X XXXX X XXXX X XXXX X XXXX X
 X X XXXX X X XXXX X X XXXX X X XXXX X
 X X XXXX X X XXXX X X XXXX X X XXXX X
 CATEGORICAL GEMTRIX
 TECHNIQUES DEVELOPMENT LABORATORY

FOR STATION "DCA"
 VALID FROM 79061717 TO 79061816
 BASED ON 79061716 SINGLE STATION SURFACE OBSERVATION

CATEGORY	OBS	24 HOURLY FORECASTS AFTER OBSERVATION (LOCAL STANDARD TIME)																							
		1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16								
MONTH (LOCAL)	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	6	
MONTH DIR	17	18	19	20	21	22	23	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	
WND SPD (KTS)	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	
SLP (MB)	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	5	
DBT TEMP (DEG F)	15	15	15	15	15	15	15	15	15	15	15	15	15	15	15	15	15	15	15	15	15	15	15	15	
DPT DPR (DEG F)	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	4	
SKY CUR (%)	12	12	12	12	12	12	12	12	12	12	12	12	12	12	12	12	12	12	12	12	12	12	12	12	
VSATY (ST MI)	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	
WLLATHER																									
FOG/GND FOG																									
KH HD KD HD KH																									
BS BD RN BY																									
DRIZZLE																									
RAIN SHOWERS																									
SNOW SHOWERS IC																									
FREEZING DRIZZLE																									
FREEZING RAIN																									
THUNDERSTORM A																									
CLDD CUR < 200 FT																									
CLDD CUR < 500 FT																									
CLDD CUR < 1000 FT																									
PRES CHG (.1 MB)																									
CIG (100'S FT)																									

Table 6. Categorical matrix (CATRIX) generated from GEMTRIX in Table 5 from threshold probabilities designed to create distributions that agree with the observed distributions. The numbers in each cell are identified by referring to the intervals or categories in Table 2.

Table 7. Proposed weather element transformation intervals and categories to succeed those in Table 2.

1	MONTH (LOCAL)	JAN	51	10-19KTS
2		FEB	52	NNW-N 1 - 9KTS
3		MAR	53	10-19KTS
4		APR	54	NE \geq 20KTS
5		MAY	55	SE \geq 20KTS
6		JUN	56	SW \geq 20KTS
7		JUL	57	NW \geq 20KTS
8		AUG	58	SLP (MB) SLP \leq 985.0
9		SEP	59	985.1 to 990.0
10		OCT	60	990.1 to 995.0
11		NOV	61	995.1 to 1000.0
12		DEC	62	1000.1 to 1005.0
13	HOUR (LOCAL)	00	63	1005.1 to 1010.0
14		01	64	1010.1 to 1015.0
15		02	65	1015.1 to 1020.0
16		03	66	1020.1 to 1025.0
17		04	67	1025.1 to 1030.0
18		05	68	1030.1 to 1035.0
19		06	69	1035.1 to 1040.0
20		07	70	1040.1 to 1090.0
21		08	71	DB TEMP -130 to -31
22		09	72	(DEG F) -30 to -26
23		10	73	-25 to -21
24		11	74	-20 to -16
25		12	75	-15 to -11
26		13	76	-10 to -6
27		14	77	-5 to -1
28		15	78	0 to 4
29		16	79	5 to 9
30		17	80	10 to 14
31		18	81	15 to 19
32		19	82	20 to 24
33		20	83	25 to 29
34		21	84	30 to 34
35		22	85	35 to 39
36		23	86	40 to 44
37	WIND ROSE	CALM	87	45 to 49
38	NNE-NE	1 - 9KTS	88	50 to 54
39		10-19KTS	89	55 to 59
40	ENE-E	1 - 9KTS	90	60 to 64
41		10-19KTS	91	65 to 69
42	ESE-SE	1 - 9KTS	92	70 to 74
43		10-19KTS	93	75 to 79
44	SSE-S	1 - 9KTS	94	80 to 84
45		10-19KTS	95	85 to 89
46	SSW-SW	1 - 9KTS	96	90 to 94
47		10-19KTS	97	95 to 99
48	WSW-W	1 - 9KTS	98	100 to 104
49		10-19KTS	99	105 to 109
50	WNW-NW	1 - 9 KTS	100	110 to 114

101	DB TEMP	115 to 150°F	157	HGT (100's FT)	0 to 1
102	DPT DEPRESSION	0	158		2 to 4
103	(DEG F)	1	159		5 to 6
104		2 to 4	160		7 to 9
105		5 to 7	161		10 to 14
106		8 to 11	162		15 to 19
107		12 to 15	163		20 to 24
108		16 to 19	164		25 to 29
109		20 to 25	165		30 to 39
110		26 to 35	166		40 to 49
111		36 to 50	167		50 to 59
112		51 to 115	168		60 to 75
113	SKY1	PARTIAL OBSCURATION	169		76 to 99
114		TOTAL OBSCURATION	170		100 to 150
115	CLR	0/10	171		≥ 151
116	SCD	1/10 to 5/10	172	VISIBILITY	0 to .49
117	BKN	6/10 to 9/10	173	(MI)	.50 to .74
118	OVC	10/10	174		.75 to .99
119	HGT1	(100's FT) 0 to 1	175		1.00 to 1.49
120		2 to 4	176		1.50 to 1.99
121		5 to 6	177		2.00 to 2.49
122		7 to 9	178		2.50 to 2.99
123		10 to 14	179		3.00 to 3.99
124		15 to 19	180		4.00 to 4.99
125		20 to 24	181		5.00 to 5.99
126		25 to 29	182		6.00 to 6.99
127		30 to 39	183		7.00 to 100.00
128		40 to 49	184	WEATHER	NO WX
129		50 to 59	185		F, IF
130		60 to 75	186		GF
131		76 to 99	187	K, H, D, KH, KD, HD, KHD	
132		100 to 150	188	BS, BD, BN, BY	
133		≥ 151	189		L-
134	SKY2	CLR 0/10	190		L
135	SCD	1/10 to 5/10	191		L+
136	BKN	6/10 to 9/10	192		R-
137	OVC	10/10	193		R
138	HGT2	(100's FT) 0 to 1	194		R+
139		2 to 4	195		RW-
140		5 to 6	196		RW
141		7 to 9	197		RW+
142		10 to 14	198		S-, IC-
143		15 to 19	199		S, IC
144		20 to 24	200		S+, IC+
145		25 to 29	201		SW-, IP-, SP-
146		30 to 39	202		SW, IP, SP
147		40 to 49	203		SW+, IP+, SP+
148		50 to 59	204		ZL-
149		60 to 75	205		ZL
150		76 to 99	206		ZL+
151		100 to 150	207		ZR-
152		≥ 151	208		ZR
153	SKY3	CLR 0/10	209		ZR+
154	SCD	1/10 to 5/10	210		T LIGHT
155	BKN	6/10 to 9/10	211		T HEAVY
156	OVC	10/10	212	PRES CHNG	-90.0 to -4.0 (MB)

Table 7.
continued

213	PRES CHNG (CONT)	-3.9 to -3.0
214		-2.9 to -2.0
215		-1.9 to -1.0
216		-.9 to 0.0
217		.1 to .9
218		1.0 to 1.9
219		2.0 to 2.9
220		3.0 to 3.9
221		4.0 to 90.0

PERFORMING STEP 5:

The same forecasters were then provided the most recent synoptic charts. As a result they made one modification to the climatological FT they originally created. Namely, they added "Chance of Thundershowers" for 05Z. Thus the final FT became:

FORECASTER'S FINAL FT

DCA 8 SCT C12 OVC 3410.
 05Z 10 SCT C16 OVC CHC TRW
 11Z C15 OVC 3FH.
 17Z C15 BKN 5H BKN V SCT 1010.
 21Z MVFR CIG BCMG VFR.

NNNN

COMMENTS ON EXAMPLE

Favorable:

- o Weather can actually be initiated and dissipated when employing the CATRIX. This, in fact, occurred in the example at numerous times, e.g., fog, haze, and their dissipation.
- o Note that it was possible to enter "probabilities" for the month to split the effect of the observation not being at midmonth. The same thing could have been done if the time was a fractional part of the hour.
- o In general there is a high degree of consistency among the weather elements in CATRIX throughout the 24-h period, e.g., between visibility, dewpoint depression, and the types of weather anticipated.

Unfavorable:

- o Snow probabilities greater than 0.0 in June with temperatures in the 60's is disconcerting. This is due to there being no guarantee that adding terms in the regression equation, for certain input conditions, will yield acceptable probability estimates (technically referred to as the nonadditivity problem). The saving grace in this situation is that it did not, and would not likely, appear in CATRIX since CATRIX acts as a filter. This problem will be completely eliminated with the new statistical procedure called Discrete Likelihood Functions (DLF) because the probability under these initial conditions would come out to be 0.0 (see Miller 1979).
- o Lack of consistency between the overall no weather variable and its component parts. However, this is correctable. In this example, the proper hierarchy would have been to ignore the NO WX variable and only allow for weather in the CATRIX when a particular type of weather, such as fog or haze, is categorically estimated to occur.
- o Ceiling conditions proved to be estimated on the pessimistic side relative to the actual outcome although a certain amount of fluctuation was observed.
- o The intervals and categories chosen here do not appear to be optimal. Our future plans call for revamping them.

SUMMARY AND FUTURE WORK

It would seem that the procedure described and demonstrated here would be useful as an aid to the forecaster especially at short range, say, one

to six hours. In particular, it has the following beneficial characteristics:

1. More skillful than climatology and persistence where independent verifications have shown GEM to be as good as or better than TAP (Terminal Alert Procedure; see Vercelli and Hefferman 1978) for ceiling and visibility. Thus, GEM would seem to be a logical successor to TAP for AFOS since it includes projections on all of the remaining weather elements.
2. Capable of giving interesting and useful "best" category estimates as demonstrated in the "creation" and "dissipation" of fog and haze.
3. Since only the current observation is required to use GEM, any significant changes encountered in the present weather can initiate an instantaneous update of future conditions. There is no need to wait for the record observation. It is anticipated that this will be extremely well suited to the AFOS era both as a monitoring device and as a highly refined climatological aid to the forecaster.
4. Needs only a minicomputer such as exists in AFOS to perform the necessary calculations.
5. Is fully objective with interpretable regression coefficients as can be seen by studying the cloud cover equations in Table 3.
6. Provides a "what if" capability for the forecaster by merely modifying the input.

Future plans call for the following refinements:

1. Defining better intervals and categories in the weather variable transformations. For example, see Table 7.
2. Developing a generalized operator system so that the same regression equations are applicable anywhere provided an observation and that location's climatology are available. This is made possible by extracting location climatology at the outset, before developing the generalized regression equations, and then having it reinstated to make the equation "location specific".
3. Mapping forecasts when these are available for the entire area. This would give a capability for analysis and "time-lapse" projection.
4. Accounting for interactions among the observed variables by means of a technique such as DLF will be tested.
5. Performing a spectral decomposition of the regression-equation matrix employing eigen-

value analysis. This should yield fundamental climatological concepts and interpretations.

6. Obtaining best single estimates in CATRIX for elements such as temperature in degrees F or C. These can and will be gotten by using expectation or some similar procedure.

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