

AD-749 260

USADAC TECHNICAL LIBRARY

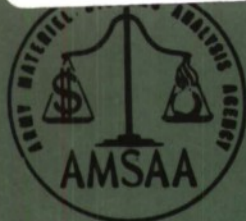


5 0712 01015629 6

TECHNICAL
LIBRARY

749260

AD



AMSAA

TECHNICAL MEMORANDUM NO. 140

METHODS FOR APPLYING A TRACKING SIGNAL TO
MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS

MICHAEL R. CHERNICK

AUGUST 1972

Approved for public release;
distribution unlimited.

U. S. ARMY MATERIEL SYSTEMS ANALYSIS AGENCY
Aberdeen Proving Ground, Maryland

1301-07 APL-AMSAA
The Memo-140

DISPOSITION

Destroy this report when no longer needed. Do not return it to the originator.

DISCLAIMER

The findings in this report are not to be construed as an official Department of the Army position.

WARNING

Information and data contained in this document are based on the input available at the time of preparation. The results may be subject to change and should not be construed as representing the AMC position unless so specified.

TECHNICAL MEMORANDUM NO. 140

METHODS FOR APPLYING A TRACKING SIGNAL TO
MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS

Michael R. Chernick

August 1972

Approved for public release;
distribution unlimited.

RDT&E Project No. 1P76580IMM1102

U.S. ARMY MATERIEL SYSTEMS ANALYSIS AGENCY
ABERDEEN PROVING GROUND, MARYLAND

U.S. ARMY MATERIEL SYSTEMS ANALYSIS AGENCY

TECHNICAL MEMORANDUM NO. 140

MRChernick/sjj
Aberdeen Proving Ground, Md.
August 1972

METHODS FOR APPLYING A TRACKING SIGNAL TO
MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS

ABSTRACT

The tracking signal developed by Trigg which has gained wide acceptance is useful for monitoring single exponential smoothing forecasts. If forecasts become biased, the tracking signal will increase in absolute value indicating the possible need to change the smoothing constant which is the one adjustable parameter in single exponential smoothing.

Two methods of applying the tracking signal to monitor exponential smoothing forecasts are discussed in this document. The method which was applied to depot workload forecasts is described in detail and the tables of confidence limits which are necessary to apply this method are given.

FOREWORD

Single exponential smoothing is a technique used for determining a forecast based on historical data. Given the historical data, the forecast is uniquely determined on the basis of one parameter known as the smoothing constant. In order to maintain as accurate a forecast as possible on the basis of this technique it is important to find the best value for the smoothing constant. Since the historical pattern can change with time or because there is not sufficient data to adequately develop the optimal smoothing constant, this constant should be revised periodically. The tracking signal is a statistic developed by D. W. Trigg (Reference 1) which indicates when a change in the smoothing constant might be necessary.

Confidence limits for the tracking signal were determined by M. Batty (Reference 2). These limits are useful when 50 or more periods of historical data are available for the computation of the tracking signal. However, in many practical situations as few as 10 periods of historical data are available. In such situations the distribution of the tracking signal is significantly different.

The objective of this report is to illustrate how the tracking signal can be used to monitor exponential smoothing forecasts and to determine confidence limits when only 10 to 20 periods of data are available. One who applies the exponential smoothing procedure to his forecast can use these tables to decide whether to revise the smoothing constant. When the absolute value of the tracking signal exceeds the confidence limit the smoothing constant should be revised.

¹Trigg, D. W.; Monitoring a Forecasting System, Opl Res Q15, 271-274, (1964).

²Batty, M.; Monitoring an Exponential Smoothing Forecasting System, Opl Res Q20, No. 3, 319-325, (1969).

CONTENTS

	Page
ABSTRACT	3
FOREWORD	4
1. THE NEED FOR A TRACKING SIGNAL IN EXPONENTIAL SMOOTHING FORECASTS	7
2. DEFINITION AND ELEMENTARY PROPERTIES OF THE TRACKING SIGNAL. .	7
3. TECHNIQUES FOR IMPROVING SMOOTHING FORECASTS BY USE OF THE TRACKING SIGNAL	8
3.1 Adaptive Smoothing	8
3.2 Monitoring Exponential Smoothing Forecasts Using Confidence Limits for the Tracking Signal	9
LITERATURE CITED	13
APPENDIX - 90 PERCENT CONFIDENCE LIMITS	15
DISTRIBUTION LIST	17

METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS

1. THE NEED FOR A TRACKING SIGNAL IN EXPONENTIAL SMOOTHING FORECASTS

Single exponential smoothing is a forecasting technique that is useful when the historical data have no trend. In this case, the data are assumed to be relatively constant with random errors causing fluctuations. Based on the historical data an estimate of the assumed constant is determined and used for the next forecast period, t . When additional data become available the estimate is revised for the forecast period $t + 1$.

The smoothing constant is the one parameter in single exponential smoothing which can be adjusted. A method for optimizing the smoothing constant and its application to a depot supply workload problem is given in Reference 3.

Because historical patterns change, the smoothing constant that works best during a given period may not continue to be the best in the future and thus, it may be necessary to revise it. Also, in some practical problems the limited amount of historical data available is not sufficient to adequately determine the optimal smoothing constant. In such situations, it would also be desirable to monitor the forecasts to determine when to revise the smoothing constant.

The tracking signal provides a method for monitoring the forecast. Two applications for adjusting the forecast by use of the tracking signal are presented in Section 3.

2. DEFINITION AND ELEMENTARY PROPERTIES OF THE TRACKING SIGNAL

Let A = the smoothing constant,

S_t = data for period, t ,

$\hat{S}_{t-1}$ = the forecast made for period, t , and

$$\epsilon_t = \hat{S}_{t-1} - S_t .$$

ϵ_t represents the error between the forecast and actual for period, t . The smoothing forecast for period, $t + 1$ is obtained from the following formula:

³Chernick, M. R.; Use of Exponential Smoothing to Forecast Depot Supply Workloads, SM&RSO Interim Note No. 1, (1971), U.S. Army Materiel Systems Analysis Agency, Aberdeen Proving Ground, Maryland.

$$\hat{S}_t = AS_t + (1-A) \hat{S}_{t-1} \text{ where } 0 < A < 1 .$$

We define the smoothed error recursively as follows

$$\hat{\epsilon}_t = \alpha \epsilon_t + (1-\alpha) \hat{\epsilon}_{t-1} \text{ where } 0 < \alpha < 1 \text{ and } \alpha \text{ is fixed.}$$

Let $|\epsilon_t|$ denote the absolute value of the error in period, t . We then define the smoothed absolute error recursively by the following:

$$\delta_t = \alpha |\epsilon_t| + (1-\alpha) \delta_{t-1} .$$

Let T_t denote the tracking signal for the period, t . We then define the tracking signal as

$$T_t = \frac{\hat{\epsilon}_t}{\delta_t} .$$

The tracking signal has the following properties:

- a. $-1 \leq T_t \leq 1$ for every t .
- b. $T_t = 1$ if and only if the forecast exceeds the actual in each of the previous periods.
- c. $T_t = -1$ if and only if the actual exceeds the forecast in each of the previous periods.

Properties b and c suggest that as the tracking signal approaches unity, the forecasts will have tended to consistently exceed the actual and that as the tracking signal approaches -1 , the forecasts will have tended to be consistently below the actual. This is equivalent to saying that as the tracking signal approaches 1 in absolute value the forecast becomes biased.

3. TECHNIQUES FOR IMPROVING SMOOTHING FORECASTS BY USE OF THE TRACKING SIGNAL

3.1 Adaptive Smoothing.

Adaptive smoothing is a technique devised by Trigg and Leach (Reference 4) to enable the forecast to automatically react when the tracking signal indicates the occurrence of biased forecasts.

⁴Trigg, D. W., and Leach, A. G.; Exponential Smoothing With an Adaptive Response Rate, Opl Res Q18, No. 1, 53-59.

When the smoothing constant A is small and a sudden change (either a rapid increase or decrease) occurs in the data, one obvious way to react would be to increase the smoothing constant. This would enable the forecast to home in on the new situation more quickly. Once the forecast has adjusted to the new situation, it may be desirable to lower the smoothing constant again. It may be that the pattern becomes stable in the new situation and if this is the case the high smoothing constant may then be causing too much reaction to the random noise.

To achieve these types of reaction Trigg and Leach thought of a very simple method; namely, to set the smoothing constant equal to the absolute value of the tracking signal. This method is called adaptive smoothing. This means that the smoothing constant will be continually changing from forecast to forecast according to the way the tracking signal is changing. As the absolute value of the tracking signal approaches unity, the smoothing constant increases toward one and thus the forecast reacts more quickly to the change. After the reaction takes place the errors should return to being unbiased and hence the tracking signal should decrease in absolute value. This would then produce the desired decrease in the smoothing constant.

A modification to this procedure would be to let the smoothing constant be equal to k times the absolute value of the tracking signal where $0 < k < 1$. This would produce the same effect and be less restrictive. Trigg's method would correspond to $k = 1$.

The adaptive smoothing technique allows for an automated response to changes and thus would be desirable and easy to apply when forecasts are handled by computer for a large number of items. On the other hand in some situations it may be better for the individuals involved in preparing the forecast to be made more aware of the changes that are occurring. The tracking signal can also be used to alert the forecaster of changes and allow him to decide on the response. One such method is described in Section 3.2.

3.2 Monitoring Exponential Smoothing Forecasts Using Confidence Limits for the Tracking Signal.

If the single exponential smoothing forecast is behaving well, the error between the forecast and actual will have zero mean. If we assume that the errors are normally distributed with zero mean, we can attempt to determine the distribution of the tracking signal. This will yield confidence limits for the tracking signal which can then be used to detect when a change in the smoothing constant is necessary.

⁴Ibid.

The variance of the smoothed error can be determined analytically in terms of the variance of the error in the case where α is set equal to A and errors are not serially dependent. Batty shows that in this case, the smoothed error is then normally distributed with mean zero and variance

$$2\sigma_e^2(A/(2-A)^3), \text{ (Reference 2).}$$

If we then had the distribution of the smoothed absolute error, we would still have the problem that the tracking signal is the ratio of these two variables which are serially correlated. It is therefore an extremely difficult problem to analytically derive the distribution of the tracking signal in general. However, for small values of A, if observations of the error are serially independent the tracking signal is approximately normally distributed with zero mean and variance:

$$\frac{A\pi}{2(2-A)^2}, \text{ (Reference 2).}$$

Batty uses this distribution to determine confidence limits for A = 0.1 at the 0.3, 0.2, 0.15, 0.10, 0.05, 0.04, 0.02 and 0.01 significance levels. When A is not small, the assumptions fail.

For larger values of A one can resort to Monte Carlo to generate samples for the data with the appropriate error distribution. From these data, sample values for the tracking signal can be generated and confidence limits can be obtained from the histograms. Batty did this for various values of σ_e and found that the tracking signal variance was independent of the error variance. He also found for A = 0.1, close agreement between the confidence limits obtained from Monte Carlo and those gotten by assuming a constant smoothed absolute error. He found that the distribution of the tracking signal becomes bimodal as A increases and this is very noticeable for high values (i.e., 0.4 or 0.5) of the smoothing constant.

The confidence limits obtained by Batty include A = 0.1, 0.2, 0.3, 0.4, and 0.5. He did not however, determine the limits for higher values of the smoothing constant and for a small number of periods of data available. Since higher values of the smoothing constant were commonly preferred and few periods of data were available in an application of exponential smoothing to depot supply workload, it was necessary to determine those confidence limits.

Consequently, a sample of 1,000 values of the tracking signal was generated for various values of N (number of periods) and A. These values were gotten by generating values for the error in each of the

²Batty, M.; Op. Cit., pp 322-323.

periods, computing the smoothed error and the smoothed absolute error, and getting the tracking signal as the quotient of these two quantities. The sample values for the error were gotten from a pseudo-random number generator for a normal distribution with mean zero and variance 1. The distribution of the tracking signal is believed to be independent of the error variance σ_ϵ (Reference 2). To test this, a value of $\sigma_\epsilon = 101$ was used and it gave the same tracking signal histogram. This agrees with Batty's observation for large N.

From the 1,000 values of the tracking signal, the point below which 90 percent of the values of the absolute value of the tracking fell, was used as a 90 percent confidence limit for the absolute value of the tracking signal.

It was found that when the tracking signal was based on only a few periods of data its distribution appeared to be much different from when very many periods of data were used. It was discovered that for 50 to 100 periods the confidence limits were in close agreement with Batty's for $A = 0.1, 0.2, 0.3, 0.4,$ and 0.5 . However, when the number of periods was less than 20, confidence limits were significantly higher.

The Appendix gives the tables for the confidence limits for various values of N. A comparison for large N with Batty's confidence limits are also given in the Appendix.

The user can apply these tables to indicate when a bias exists in an exponential smoothing forecast. He must compute a tracking signal based on the errors he has experienced in the past. Given the value of the smoothing constant, for example, suppose a value of $A = 0.4$ is used, and given the number of periods of past data used to compute the forecast, say 13, he will find from Table 2, that the probability that the tracking signal will exceed 0.72 in absolute value is only 0.1. So, if the computed value is greater than 0.72, he may wish to infer that the forecast is biased. In that event, he may decide to revise the smoothing constant. One way to do this would be to recompute an optimal smoothing constant by using a technique such as the one mentioned in Reference 3.

²Loc. Cit.

³Loc. Cit.

LITERATURE CITED

1. Trigg, D. W.; Monitoring a Forecasting System, Opl Res Q15, 271-274, (1964).
2. Batty, M.; Monitoring an Exponential Smoothing Forecasting System, Opl Res Q20, No. 3, 319-325, (1969).
3. Chernick, M. R.; Use of Exponential Smoothing to Forecast Depot Supply Workloads, SM&RSO Interim Note No. 1, (1971), U.S. Army Materiel Systems Analysis Agency, Aberdeen Proving Ground, Maryland.
4. Trigg, D. W., and Leach, A. G.; Exponential Smoothing With an Adaptive Response Rate, Opl Res Q18, No. 1, 53-59.

APPENDIX

90 PERCENT CONFIDENCE LIMITS

A comparison of 90 percent confidence limits computed by the author for large N versus Batty's confidence limits.*

TABLE I A COMPARISON OF 90 PERCENT CONFIDENCE LIMITS

A	<u>Author's Test</u>	<u>Batty's Test²</u>
	Confidence Limit on the Absolute Value of the Tracking Signal	Confidence Limit on the Absolute Value of the Tracking Signal
0.1	0.36	0.35
0.2	0.50	0.50
0.3	0.61	0.63
0.4	0.73	0.72
0.5	0.80	0.82

* Differences are well within limits of variation due to Monte Carlo Sampling.

²Batty, M.; Op. Cit., p 324.

TABLE 2 90 PERCENT CONFIDENCE LIMITS FOR ABSOLUTE VALUE OF
THE TRACKING SIGNAL FOR SMALL VALUES OF N

N \ A	10	11	12	13	14	15	16	17	18	19	20	25	30	35	40	50
0.1	0.96	0.93	0.91	0.87	0.85	0.83	0.80	0.76	0.71	0.73	0.65	0.55	0.47	0.39	0.36	0.34
0.2	0.81	0.77	0.71	0.65	0.63	0.59	0.56	0.54	0.53	0.52	0.53	0.49	0.50	0.51	0.49	0.49
0.3	0.68	0.70	0.65	0.64	0.62	0.64	0.60	0.61	0.62	0.62	0.60	0.62	0.62	0.62	0.65	0.62
0.4	0.76	0.74	0.72	0.72	0.74	0.72	0.73	0.75	0.74	0.75	0.73	0.74	0.71	0.71	0.71	0.71
0.5	0.84	0.82	0.82	0.82	0.81	0.81	0.82	0.82	0.82	0.82	0.81	0.80	0.80	0.81	0.80	0.82
0.6	0.87	0.88	0.89	0.88	0.88	0.87	0.89	0.90	0.88	0.89	0.89	0.88	0.87	0.90	0.87	0.88
0.7	0.94	0.93	0.93	0.92	0.92	0.92	0.92	0.93	0.93	0.93	0.94	0.93	0.93	0.94	0.93	0.93
0.8	0.97	0.98	0.97	0.97	0.97	0.98	0.97	0.96	0.97	0.97	0.97	0.97	0.97	0.97	0.97	0.97
0.9	0.99	0.99	0.99	0.99	0.99	0.99	0.99	0.99	0.99	0.99	0.99	1.00	0.99	0.99	0.99	0.99

DISTRIBUTION LIST

<u>No. of Copies</u>	<u>Organization</u>	<u>No. of Copies</u>	<u>Organization</u>
12	Commander Defense Documentation Center ATTN: TIPCR Cameron Station Alexandria, Virginia 22314	1	Commanding General U.S. Army Materiel Command ATTN: AMCRD-E Washington, D.C. 20315
1	Commanding General U.S. Army Materiel Command ATTN: AMCCP Washington, D.C. 20315	1	Commanding General U.S. Army Materiel Command ATTN: AMCRD-G Washington, D.C. 20315
1	Commanding General U.S. Army Materiel Command ATTN: AMCDL Washington, D.C. 20315	1	Commanding General U.S. Army Materiel Command ATTN: AMCRD-M Washington, D.C. 20315
1	Commanding General U.S. Army Materiel Command ATTN: AMCDT Washington, D.C. 20315	1	Commanding General U.S. Army Materiel Command ATTN: AMCRD-P Washington, D.C. 20315
1	Commanding General U.S. Army Materiel Command ATTN: AMCDT-P Washington, D.C. 20315	1	Commanding General U.S. Army Materiel Command ATTN: AMCRD-R Washington, D.C. 20315
1	Commanding General U.S. Army Materiel Command ATTN: AMCMA Washington, D.C. 20315	1	Commanding General U.S. Army Materiel Command ATTN: AMCRD-T Washington, D.C. 20315
1	Commanding General U.S. Army Materiel Command ATTN: AMCMS Washington, D.C. 20315	1	Commanding General U.S. Army Materiel Command ATTN: AMCRD-W Washington, D.C. 20315
1	Commanding General U.S. Army Materiel Command ATTN: AMCPA-SA Washington, D.C. 20315	1	Commanding General U.S. Army Materiel Command ATTN: AMCPA-PM-MBT Washington, D.C. 20315
1	Commanding General U.S. Army Materiel Command ATTN: AMCQA Washington, D.C. 20315	1	Commanding General U.S. Army Materiel Command ATTN: AMCPM-SA Picatinny Arsenal Dover, New Jersey 07801

DISTRIBUTION LIST

<u>No. of Copies</u>	<u>Organization</u>	<u>No. of Copies</u>	<u>Organization</u>
1	Director MUCOM Operations Research Group ATTN: Mr. Crumb Edgewood Arsenal, Maryland 21010	1	Commanding General U.S. Army Mobility Equipment Command ATTN: AMSME-G 4300 Goodfellow Blvd St. Louis, Missouri 63120
1	Commanding Officer U.S. Army Harry Diamond Laboratories ATTN: Sys Anal Ofc Washington, D.C. 20438	1	Commanding General U.S. Army Munitions Command ATTN: AMSMU-RE-R Dover, New Jersey 07801
1	Commanding Officer Frankford Arsenal ATTN: Mr. George Schechter Philadelphia, Pennsylvania 19137	1	Commanding General U.S. Army Tank-Automotive Command ATTN: AMSTA-CV Warren, Michigan 48090
1	Chief, Analytical Sciences Office USA Biological Defense Research Laboratory ATTN: AMXBL-AS Dugway, Utah 84022	1	Commanding General U.S. Army Weapons Command ATTN: AMSWE-Y Rock Island, Illinois 61202
1	Commanding General U.S. Army Aviation Systems Command ATTN: AMSAV-R-X P.O. Box 209, Main Office St. Louis, Missouri 63166	1	Commanding Officer U.S. Army Logistics Management Center ATTN: AMXMC-LR-LOS (Mr. Griswold) Fort Lee, Virginia 23801
1	Commanding General U.S. Army Electronics Command ATTN: AMSEL-PL Fort Monmouth, New Jersey 07703	1	Commanding Officer USA Major Item Data Agency ATTN: AMXMI-O (Mr. W. Nye) Letterkenny Army Depot Chambersburg, Pennsylvania 17201
			<u>Aberdeen Proving Ground</u>
1	Commanding General U.S. Army Missile Command ATTN: AMSMI-DA Redstone Arsenal, Alabama 35809		CG, USATECOM ATTN: AMSTE-TS

AD	METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP765801MM1102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive Smoothing 5. Test for bias errors	Chernick, Michael R. U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP765801MM1102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors
AD	METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP765801MM1102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive Smoothing 5. Test for bias errors	AD

AD	METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP76580IMMI102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive Smoothing 5. Test for bias errors	Chernick, Michael R. U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP76580IMMI102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors
AD	Tracking signal U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP76580IMMI102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors	Monitoring forecasts U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP76580IMMI102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors

AD Exponential smoothing U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP765801MM1102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors	AD Adaptive smoothing U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP765801MM1102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors
AD Test for bias errors U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP765801MM1102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors	AD U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP765801MM1102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors

AD U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP76580IMM1102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors	AD U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP76580IMM1102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking Signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors
AD U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP76580IMM1102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors	AD U.S. Army Materiel Systems Analysis Agency (AMSAA), Aberdeen Proving Ground, Maryland 21005, AMSAA Technical Memorandum No. 140, Report UNCLASSIFIED METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS, Michael R. Chernick, AMSAA; Project No. RDT&E IP76580IMM1102, August 1972 Trace Index Terms Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors 1. Tracking signal 2. Monitoring forecasts 3. Exponential smoothing 4. Adaptive smoothing 5. Test for bias errors

DOCUMENT CONTROL DATA - R & D

(Security classification of title, body of abstract and indexing annotation must be entered when the overall report is classified)

1. ORIGINATING ACTIVITY (Corporate author) U.S. Army Materiel Systems Analysis Agency Aberdeen Proving Ground, Maryland		2a. REPORT SECURITY CLASSIFICATION Unclassified	
		2b. GROUP	
3. REPORT TITLE METHODS FOR APPLYING A TRACKING SIGNAL TO MONITOR SINGLE EXPONENTIAL SMOOTHING FORECASTS			
4. DESCRIPTIVE NOTES (Type of report and inclusive dates)			
5. AUTHOR(S) (First name, middle initial, last name) Michael R. Chernick			
6. REPORT DATE August 1972		7a. TOTAL NO. OF PAGES 18	7b. NO. OF REFS 4
8a. CONTRACT OR GRANT NO.		9a. ORIGINATOR'S REPORT NUMBER(S)	
b. PROJECT NO. RDT&E IP76580IMMII02		Technical Memorandum No. 140	
c.		9b. OTHER REPORT NO(S) (Any other numbers that may be assigned this report)	
d.			
10. DISTRIBUTION STATEMENT Approved for public release; distribution unlimited.			
11. SUPPLEMENTARY NOTES		12. SPONSORING MILITARY ACTIVITY U.S. Army Materiel Command Washington, D.C. 20315	
13. ABSTRACT The tracking signal developed by Trigg which has gained wide acceptance is useful for monitoring single exponential smoothing forecasts. If forecasts become biased, the tracking signal will increase in absolute value indicating the possible need to change the smoothing constant which is the one adjustable parameter in single exponential smoothing. Two methods of applying the tracking signal to monitor exponential smoothing forecasts are discussed in this document. The method which was applied to depot workload forecasts is described in detail and the tables of confidence limits which are necessary to apply this method are given.			

14	KEY WORDS	LINK A		LINK B		LINK C	
		ROLE	WT	ROLE	WT	ROLE	WT
	Tracking signal Monitoring forecasts Exponential smoothing Adaptive smoothing Test for bias errors						