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Forecasting the Diffusion of Innovation: A Stochastic Bass Model with Log-Normal and Mean-Reverting Error Process

Juho Kanninen*, Saku Mäkinen, Robert Piché, Alok Chakrabarti

Abstract

Forecasting the diffusion of innovations plays a major role in managing technology development and in engineering management overall. In this paper, we extend the conventional Bass model stochastically by specifying the error process of sales as log-normal and mean-reverting. Our model satisfies the following reasonable properties that are generally ignored in the existing literature: sales cannot be negative, the error process can have a memory, and sales fluctuate more when they are high and less when they are low. The conventional and widely used model that assumes normally distributed error term does not have these properties. We address how to forecast properly under the log-normal and mean-reverting error process and show analytically and numerically that in our extended model sales forecasts can substantially alter conventional Bass forecasts. We also analyze the model empirically, showing that our extension can improve the accuracy of future sales forecasts.

Keywords: Innovation forecasting; diffusion models; stochastic processes

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I. INTRODUCTION

Changing expectations of future sales has become a common theme in the news lately during the current global financial crisis. This reflects the fact that uncertainty plays a key role in the adoption of innovations, a fact that holds in all industry sectors. Future sales are stochastic because of unexpected changes in several factors such as economic and financial conditions, technological improvements, and competition over market share.

Absolute accuracy in forecasting the future demand of current innovations even in the short term is, of course, unattainable. This, however, does not lessen the importance and relevance of forecasting, because the point is to make the best possible forecasts. Since the work of Bass [1], many models have been introduced in the literature, and widely applied, to explain and predict product diffusions (see, for example Meade and Islam [2] and Linton [3], and references therein). However, most of these studies have focused on model selection without much attention to the properties of the model's error process. For example, most studies assume that the error process of sales volume is normally distributed, leading to possible negative sales expectations. Moreover, it would be reasonable to assume that unexpected changes in the sales volume are on average greater the greater the sales volume, or, in other words, that the sales volume is a heteroscedastic variable. Yet the sales volume is typically assumed to be homoscedastic.

From the engineering management point of view, innovation forecasting presents multiple possibilities and challenges as well. For example traditional technology road mapping [4] and product characteristics' influence on portfolio management [5] heavily rely on views of expected future diffusion of innovations and market evolution. The roadmaps further define the actual technology development and engineering effort of a company. Similarly, operations and production engineering are dependent on the company's ability to forecast future demand of its offering. However, innovation forecasting as a task is subject to uncertainties both endogenous and exogenous to the company. For example, the recent economic turbulence has created shocks on the demand of innovations and also directly to organizations' operations. Therefore, an innovation forecast model should explicitly consider the error processes resulting from volatility in order to improve forecasting accuracy.

This paper shows how to forecast future sales volumes under a stochastic version of the Bass model assuming a log-normal and mean-reverting error process. To start with, we assume that the stochastic evolution of log-sales, rather than sales, is normally distributed in time and possibly mean-reverting. In particular, we assume that the sales volume is driven by the Bass model such that the logarithmic sales volume is the sum of the logarithm value of the Bass model and a normally distributed error process, which can be mean-reverting. In contrast to the assumption of normally distributed sales, we consider the log-normal assumption plausible, because sales cannot be negative and variance in the sales volume can depend on the level of the volume. Our model can thus capture heteroscedastic errors. Moreover, our assumption of mean reversion implies that the error process can be auto-correlated. The paper aims to provide better forecasts in a situation when the underlying data or theory suggest the sales are log-normal and mean-reverting. To examine the validity of the assumptions of log-normality and mean reversion with respect to empirical data, one can estimate the diffusion model in both the conventional way and as suggested in this study. In the empirical part of this paper, we find that our log-normal and mean-reverting model gives a better out-of-sample forecasting accuracy with telecommunication data. Though we focus here on the Bass model, which has dominated the forecasting literature (see [3]), the lessons learned are likely to apply to non-Bass diffusion models as well, such as any deterministic extension of the Bass model, see, for example, [6], or other non-Bass deterministic diffusion models, such as the Gompertz model (see, e.g., [2], [7], [8] for the survey of various models).

Some stochastic diffusion models have been developed to capture the "econometrics" of diffusion dynamics. Skiadas and Giovanis [9] introduced a stochastic version of the Bass model but with a very different starting point and without mean reversion. Recently, Boskwijk and Franses [10] have presented a model that satisfies non-negativeness, heteroscedasticity, and mean reversion. However, their model is characterized differently from ours: they assume that the sales volume follows a square root process, such

as that of Cox, Ingresoll, and Ross [11] for interest rates; however, ours is a log-normal characterization. Our approach has the advantage of an analytical solution to the error process that can be subsequently used in future studies. Moreover, in comparison to [10], our paper focuses on forecasting with an extensive out-of-sample analysis. Furthermore, also the Gompertz model embraces some stochastic generalizations (see, for example, [12], [13]). The log-normal error process of the sales volume has previously been assumed in the context of some models (see, for example, [14], [15], and [16]). However, there is little research focused on forecasting under the log-normal error process.

This paper shows that our settings impact forecasting in two important ways: first, Bass's expectations must be adjusted because of log-normality and mean reversion; second, as time goes by, expectations of future sales must be further adjusted when observed sales do not equal the value of the Bass model. The adjustments depend on the speed of the mean reversion and the volatility of the sales, meaning that it is not enough to estimate the Bass parameters m , p , and q , but that to forecast future sales, also the speed of the mean reversion and volatility must be estimated. As we show, these two parameters may, indeed, play a key role in sales forecasting. Our stochastic specification yields analytical expressions for conditional expectations and variances of future sales, thus constituting an important property in forecasting and risk-management. The contribution of this paper is twofold. The primary contribution is methodological; we show how the conventional expectations must be adjusted under log-normal and mean-reverting error process. The secondary contribution is empirical. We examine six time series representing analog mobile telephone adoption in different countries and carry out 540 estimations to have an extensive out-of-sample analysis. Third, we discuss the role and meaning of mean reversion in innovation forecasting.

The paper is organized as follows. The following section introduces our model in continuous time. Section 3 discusses and illustrates the theory of forecasting in terms of our settings and concretizes the roles of the speed of mean reversion and volatility in forecasting. Section 4, starting with discretization of the continuous-time process, gives an empirical demonstration. We compare our model with two alternative error models, using data for analog mobile phones in six countries to test our improvements in underlying forecasting methodology. Finally, we present our conclusions and discuss further research.

II. MODEL

Our model can be seen as a stochastic extension of the Bass [1] model. Basically, we add mean-reverting noise to the logarithmic forecast of the Bass model. Specifically, we assume that future sales, S_t , can be characterized by a continuous time process represented logarithmically as the sum of two components – the logarithm of deterministic function of time, g_t , and a stochastic diffusion, X_t ; that is,

$$\ln S_t = \ln g_t + X_t. \quad (1)$$

Here g_t is the sales volume forecast of the Bass model. It is important to note that g_t could be any deterministic extension of the Bass model, such as [6], or other non-Bass deterministic diffusion models. In this characterization, we require that for $T > t$

$$\mathbb{E}_t[\ln S_T | X_t = 0] = \ln g_t, \quad (2)$$

which means that if the stochastic term X_t is zero, then the best forecast at time t of logarithmic sales at time T is the logarithmic forecast of a deterministic model (in this case, the Bass model). Later, we consider forecasts of non-logarithmic sales with non-zero stochastic terms in our model, but (2) serves as our basic assumption for further analysis.

First, let us look at the formulation of the Bass [1] model. It assumes that, $F(t)$, the fraction of individuals adopting the product by time t , evolves as

$$\frac{dF}{dt} = (p + qF)(1 - F),$$

where p and q are constant real numbers, called the coefficients of innovation and imitation, respectively. Bass [1] gives the closed form solution of F as

$$F(s; p, q) = \frac{1 - e^{-(p+q)s}}{(q/p)e^{-(p+q)s} + 1},$$

where time $s > 0$. By denoting $f = dF/dt$, the sales at time $s > 0$, $g_s = g(s; m, p, q)$ is given by

$$\begin{aligned} g(s; m, p, q) &= mf(s; p, q) \\ &= \frac{m(p+q)^2 e^{-(p+q)s} p}{(p + qe^{-(p+q)s})^2}, \end{aligned} \quad (3)$$

where m is the total number of adopters.¹ As mentioned before, in our model, $\ln g_s$ provides the best logarithmic sales forecast for time s when the stochastic term is zero.

Second, let us characterize the stochastic component of our model. We assume that $\{X_t; t \geq 0\}$ follows the Ornstein-Uhlenbeck process:

$$dX_t = -\kappa X_t dt + \sigma dW_t, \quad X_0 = x, \quad (4)$$

where $\kappa \geq 0$ is the speed of the mean reversion, and dW_t represents an increment to a Wiener process. By applying Ito's lemma with (1) and (4), we can write the sales process as

$$dS_t = \kappa(Y_t - \ln S_t)S_t dt + \sigma S_t dW_t, \quad S_0 > 0, \quad (5)$$

where

$$Y_t = \frac{1}{\kappa} \left(\frac{\sigma^2}{2} + \frac{d \ln g}{dt}(t; m, p, q) \right) + \ln g_t.$$

Notice that, under our model, $S_t > 0$ for all $t > 0$, satisfying then the condition of positive sales. Such a mean-reversion concept was earlier used to model interest rates and commodity prices (see, for example, [17], [18]). Our sales model can also be seen as an extension of Schwartz's model of commodity prices [18]. In particular, we assume that the equilibrium level of log-sales, Y_t , is time-varying, whereas in [18] it is constant. Moreover, our model is quite analogous to that proposed by Lucia and Schwartz [19] for modeling electricity prices.²

To understand the model given in Eq. (5), notice that when $\ln S_t > Y_t$, the drift term of the sales process is negative, resulting in a pull down toward the instantaneous equilibrium level or "instantaneous normal level," and that if $\ln S_t < Y_t$, the drift term is positive, pulling up to the higher equilibrium value. This feature causes the sales process to oscillate around the time-varying instantaneous equilibrium levels driven by the Bass model. The Wiener process continually pushes the sales volume away from the instantaneous equilibrium level, and the mean-reverting term pushes the sales volume back to the equilibrium level. The greater the speed of the mean reversion, κ , the greater the pull back to the instantaneous equilibrium level.

To calibrate our model, in addition to estimating the Bass parameters m, p , and q for forecasts of future sales, we must also estimate the speed of the reversion, κ , and volatility, σ . To see how these parameters can be estimated, let us write our continuous time model (model A) in discrete form. Equation 4 takes the form

$$X_t = \psi X_{t-1} + u_t^A,$$

where $\psi := (1 - \kappa \Delta t)$, and u^A is i.i.d. normal random variable with mean zero and variance $\sigma^2 \Delta t$. Thus

$$\begin{aligned} \ln S_t &= \ln g(t; m, p, q) + X_t, \\ X_t &= \psi X_{t-1} + u_t^A, \end{aligned}$$

¹We assume, following most of existing literature, that each adopter buys only one unit.

²Lucia and Schwartz [19] incorporate periodic behavior in the deterministic function via trigonometric functions to reflect the general seasonal pattern of electricity price.

where $t = 1, 2, \dots, n$, which can also be expressed as

$$S_t^* = \psi S_{t-1}^* + g_t^* - \psi g_{t-1}^* + u_t^A, \quad (6)$$

where $S_t^* := \ln S_t$ and $g_t^* := \ln g(t; m, p, q)$. Parameters ψ , m , p , and q can be estimated simultaneously using a non-linear least squares (NLS) approach. We take $\tilde{\kappa} = (1 - \tilde{\psi})/\Delta t$ as the estimate of the speed of reversion. In addition, the above expression implies that the standard error of regression can be taken as the estimate of $\sigma\sqrt{\Delta t}$.

In addition to our log-normal and mean-reverting error model we consider two alternative error models. We call our error model (5) model A and the others B and C. The alternative error models are based on the error model proposed by Srinivasan and Mason [20] that with $\Delta t = 1$ has a normally distributed error process without mean reversion:

$$S_t = m(F(t) - F(t-1)) + u_t, \quad (7)$$

where u_t is normally distributed with a positive variance. As the authors note, this can lead to negative sales. Moreover, this characterization ignores heteroscedasticity and under this model, the forecasts of future sales will not converge to the Bass curve. In order to carry out comparisons, we use *continuous time* versions of their model. Specifically, instead of (7), we write model B as

$$S_t = g_t + u_t^B, \quad (8)$$

where u_t^B is a zero-mean error term with variance σ_B^2 . Clearly, the expected sales under model B is

$$\mathbb{E}_t S_T = g_T.$$

Thus, according to model B, the expected observable sales level equals the Bass curve g_T regardless of the current observed sales, which means that the observed sales are expected *immediately and discontinuously* to jump to the Bass curve. Assuming a fixed set of parameters, the prediction under error term B is the same whether the current observed sales is, for example, 100 or 200. That is because the sales observations affect the prediction only via the estimation procedure. Another drawback is that the variance of future sales,

$$\text{Var} S_T = \sigma_B^2,$$

is independent of the length of the forecasting period, $T - t$. This means that the future sales level for the next minute is as uncertain as for the next decade.

Model C has the form

$$S_t = g_t + Z_t, \quad (9)$$

where

$$dZ_t = \sigma_C dW_t.$$

In discrete time, model C can be written as

$$S_t - S_{t-1} = g_t - g_{t-1} + u_t^C, \quad (10)$$

where u_t^C is a zero-mean error term with variance $\sigma_C^2 \Delta t$. The expected sales volume under model C is

$$\mathbb{E}_t S_T = g_T + S_t - g_t.$$

Error model C adjusts the expected sales by the difference of the current observed sales and Bass's prediction, $S_t - g_t$, and hence the current sales observation has a direct effect on the prediction. In addition, the variance of future sales,

$$\text{Var}_t S_T = (T - t)\sigma_C^2,$$

is directly proportional to the length of the forecasting period constituting the ‘‘cone of uncertainty’’, as is true also for model A.

Clearly, both conventional error models, B and C, can give negative simulated values. Specifically, model B assumes that the sales level is normally distributed with the mean of g_t and model C assumes that the absolute change of the sales is normally distributed with the mean of g_t . In the case of model C, the expected sales can become negative if the observed sales volume is less than originally predicted by the Bass model, in which case $g_t - S_t$ can be greater than g_T .

In our model the mean reversion term makes the error process autoregressive, that is, roughly speaking, the next realization is dependent on the current one, or in other words, that the sequential changes are correlated. Why could the error structure of the sales process be autoregressive? First, we can say that empirically the error process can be autoregressive (see [21] and [10] and the empirical part of this study). The current literature of the diffusion of innovations has little or no discussion of the mean reversion in the error process, so we now proceed to a more detailed elucidation of it.

Most importantly, as we will demonstrate, if we make conventional assumptions about normally distributed changes in sales without mean reversion (model C), then the forecasts of the future sales will not converge to the Bass curve. This means that in the long term, future sales are expected to be negative if the current realized sales volume is less than forecasted by the Bass model for the current time. Similarly, under these assumptions about normally distributed changes in sales, the future sales volume is expected to converge to a strictly positive constant if the current realized sales volume is greater than forecasted by the Bass model. Obviously, it is unreasonable to expect that, in the long term, the sales volume will not converge to zero, or at least we can expect the sales to be non-negative. On the other hand, as we will see, if we employ the assumption of normally distributed sales (model B), rather than normally distributed changes in sales, we (implicitly) assume that the observed sales volume is expected immediately and discontinuously to converge to the Bass curve. This too is somewhat unrealistic because under these assumptions the expected sales volumes are not directly dependent on the current observed sales level. If the error process is mean-reverting, future sales are expected to converge to the Bass curve in the long term, and the above problems are avoided. The mean reversion plays an important role because if we ignore it by assuming that the changes in sales are i.i.d. normally distributed (model C), the expected sales level never converges to the Bass curve. Alternatively, if we suppose that the sales are normally distributed with i.i.d. error term (model B) without allowing an arbitrary mean reversion, the expected sales jumps immediately to the Bass curve, an unrealistic situation as well.

It must be noted here that the assumption of log-normal error process ensures that sales volumes remain non-negative, and if we assume both mean reversion and log-normality, forecasts have plausible properties regarding the convergence of sales to the Bass curve. One consequence of the mean reverting log-normal error process is that the greater κ , the easier it is to forecast the future sales. This is illustrated in Figure 1, which shows ten possible sales trajectories generated with our model using Monte Carlo simulation methods. The figure also plots the Bass curve, and we can see that the possible sales trajectories oscillate around it. In plot (a), the speed of the mean reversion $\kappa = 0$, which means that the stochastic variable x does not tend toward zero; in other words, the drift term of the sales process does not act against stochastic shocks, and thus the sales volume will not return to the Bass curve when the sales volume is pushed randomly away from the equilibrium. In plot (b), the stochastic variable x is strongly mean-reverting; consequently, the sales volume oscillates nearer the Bass curve.

Notice from Figure 1 how our model A captures the heteroscedasticity, too: the greater the sales volume, the greater its variance. Why should the instantaneous variance of the sales volume (expressed in units) be dependent on the sales volume? Consider a situation in which the “long-term mean” of the sales volume, i.e. the value of g_t , is very low, say close to zero (remember that according to the Bass model, sales eventually approach zero). Then the variance of the sales (in units) must be infinitesimal because sales are non-negative. That is, if the sales level is close to zero, it cannot fluctuate much around its mean. Otherwise the sales would have to be able to be negative. The greater the mean of the sales is, the more the sales can vary. This phenomenon can well be observed from our data. Figure 2 plots the monthly telecommunication data from the USA that we use in the empirical part of this study (along with data from other countries). Note that this property can also be captured with multiplicative error structures.

Fig. 1. Sample paths of sales volume with parameters $x_0 = 0$; $p = 0.01$; $q = 0.8$; $m = 100000$; $\sigma = 0.35$; $\Delta t = 1/52$. In plot (a) the stochastic term is non-mean-reverting with $\kappa = 0$, whereas in plot (b) mean reversion is strong with $\kappa = 1$.

Fig. 2. The heteroscedasticity of the sales. Time series represents monthly analog mobile telephone adoption in the USA.

III. FORECASTING

What may be surprising is that according to our error process characterization, the Bass model does not provide forecast of future sales, because, on one hand, the sales process is log-normally distributed, and, on the other hand, the error process is mean-reverting. Figure 3 illustrates this fact by showing probability distributions and expected values of future sales volumes together with the Bass curve for different forecasting periods with $x_0 = 0$. Note how the expected sales volumes do not lie on the Bass curve, especially if mean reversion is low ($\kappa = 0$). With the figure, we can postulate that Bass forecasts must be adjusted more when the mean reversion is smaller and the volatility is greater. This section examines mainly how to adjust the Bass model to obtain non-skewed forecasts.

To put this formally, let us address the expected level and variance of future sales according to our model. The logarithmic sales volume, $\ln S_T$, $T > t$, is normally distributed with conditional mean and variance

$$\begin{aligned}\mathbb{E}_t \ln S_T &= \ln g_T + X_t e^{-\kappa(T-t)} \\ &= \ln g_T + (\ln S_t - \ln g_t) e^{-\kappa(T-t)} \\ \text{Var}_t \ln S_T &= \frac{\sigma^2}{2\kappa} (1 - e^{-2\kappa(T-t)}).\end{aligned}$$

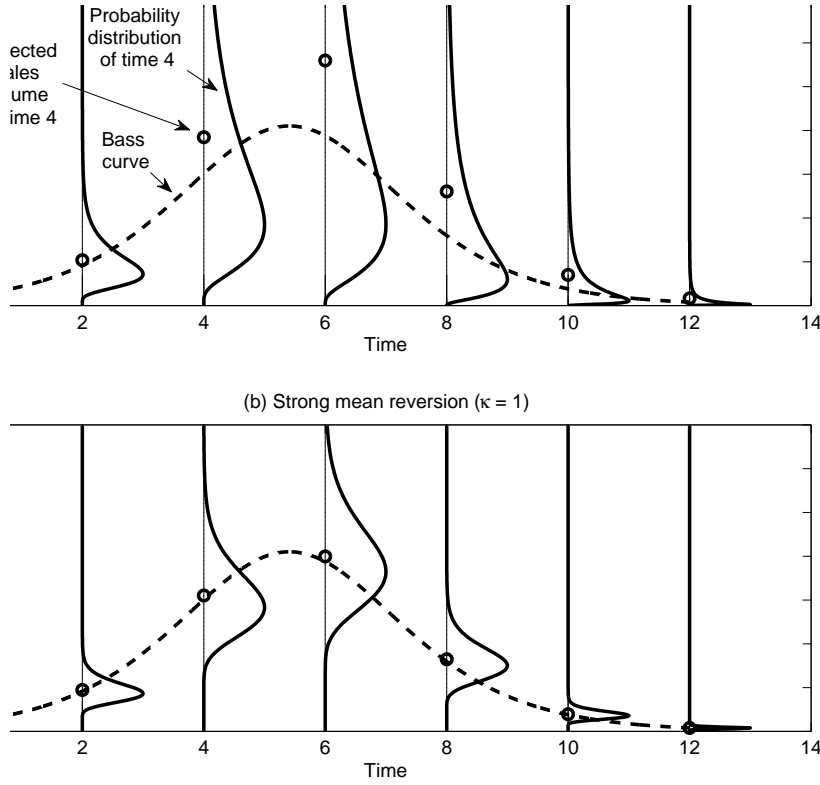


Fig. 3. Probability distributions of future sales volumes, expected values of sales volumes, and the Bass curve. Parameter values are $x_0 = 0$; $p = 0.01$; $q = 0.8$; $m = 100000$; $\sigma = 0.35$. The Bass curve is a dashed line and expected values of future sales volume by the symbol 'o'. Note that for graphical purposes the distributions are scaled to have equal amplitudes.

Because log sales are normally distributed, sales are log-normally distributed with a conditional mean

$$\begin{aligned}
 \mathbb{E}_t S_T &= \exp \left(\mathbb{E}_t \ln S_T + \frac{1}{2} \text{Var}_t \ln S_T \right) \\
 &= \exp \left(\ln g_T + (\ln S_t - \ln g_t) e^{-\kappa(T-t)} + \frac{\sigma^2}{4\kappa} (1 - e^{-2\kappa(T-t)}) \right) \\
 &= g_T b_1(t, T) b_2(t, T),
 \end{aligned} \tag{11}$$

where

$$b_1(t, T) = \exp \left((\ln S_t - \ln g_t) e^{-\kappa(T-t)} \right)$$

and

$$b_2(t, T) = \exp \left(\frac{\sigma^2}{4\kappa} (1 - e^{-2\kappa(T-t)}) \right).$$

The above implies that if the actual sales volume differs from the Bass value at the current time, then also expected future sales volumes differ from the forecasts of the Bass model, which is intuitively reasonable. What is more interesting is that even if $S_t = g_t$, that is, even if the error process were zero at the current time t , the expected value of future sales volume S_T would still be affected by the parameters of the stochastic diffusions, σ and κ . Thus there are two reasons why the pure Bass model, or any other deterministic model, cannot be used to predict future sales if the sales process is assumed log-normal with a possible mean reversion component. First, as time goes by, and as the Bass model does not fully

explain current observed sales, the Bass forecast at time t of the sales volume at a future time T , $T > t$, must be adjusted by the factor

$$b_1 = \exp\left((\ln S_t - \ln g_t)e^{-\kappa(T-t)}\right).$$

If the current sales volume, S_t , is greater (resp. less) than the Bass value, g_t , then b_1 is greater (resp. less) than one and decreases (resp. increases) with respect to T . Note that if $\kappa > 0$, then $b_1 \rightarrow 1$ as $T \rightarrow \infty$.

Second, the log-normal property of the sales volume together with mean reversion imply that the Bass model, or any other deterministic model, does not fully predict future sales, even when the current stochastic term is zero, that is, $x_t = 0$. In particular, Bass forecasts of S_T must be further adjusted by the factor

$$b_2 = \exp\left(\frac{\sigma^2}{4\kappa}\left(1 - e^{-2\kappa(T-t)}\right)\right).$$

This adjustment takes place even if the current sales volume equals the value of the Bass model, which is the case if, for example, the product has not yet been launched. Note that for $\kappa > 0$, $b_2 > 1$; consequently, the adjustment makes the expectations greater than the values implied by the Bass model. However, for a strongly mean-reverting error process, $b_2 \approx 1$. To see this, note that $b_2 \rightarrow 1$ as $\kappa \rightarrow \infty$. This is also intuitively clear, because if the stochastic variable is strongly mean-reverting, the stochastic term is negligible. Note also that the greater the time interval, $T - t$, the greater the adjustment factor b_2 . As $T \rightarrow \infty$, $b_2 \rightarrow \exp(\sigma^2/(4\kappa))$. Moreover, the diffusion coefficient, σ , increases b_2 and thus also the expected future sales. Therefore, the greater the uncertainty in future sales, the greater the expected level of future sales at any future time instant.

For $\kappa > 0$, $\mathbb{E}_t \ln S_T = \ln g_T = -\infty$ and $\mathbb{E}_t S_T = g_T = 0$ as $T \rightarrow \infty$. This means that if the error-term is *both* log-normal and mean-reverting, the expected future sales volume converges to the Bass curve in the long term.

However, the case of $\kappa = 0$ (no mean reversion) is more complex. It is true that also in this case $\mathbb{E}_t \ln S_T = \ln g_T = -\infty$; however, this does not necessarily imply that for $\kappa = 0$, $\mathbb{E}_t S_T = 0$ as $T \rightarrow \infty$. In particular, for $\kappa = 0$ and $T \rightarrow \infty$, $\mathbb{E}_t S_T = \infty$ if $p + q < \frac{1}{2}\sigma^2$ and $\mathbb{E}_t S_T = 0$ if $p + q \geq \frac{1}{2}\sigma^2$. To see this, note that if $\kappa = 0$, then $b_1 = S_t/g_t$ and $b_2 = \exp(\frac{1}{2}\sigma^2(T-t))$, and the expectation takes the form

$$\mathbb{E}_t S_T = \frac{m(p+q)^2 \exp\left(-\left(p+q - \frac{1}{2}\sigma^2\right)T - \frac{1}{2}\sigma^2 t\right) p S_t}{(p + qe^{-(p+q)T})^2} \frac{p S_t}{g_t}.$$

Violating the condition of $\lim_{T \rightarrow \infty} \mathbb{E}_t S_T = 0$, however, suggests a somewhat unrealistic situation for two reasons. First, no matter how small the speed of the mean reversion, κ , if it is strictly positive, the expected value of future sales eventually converges to the Bass curve. Second, many empirical studies have shown that in annual terms, $p + q$ is typically over 0.5. This would mean an extremely high annual volatility.

Conditional variance can be obtained from

$$\begin{aligned} \text{Var}_t S_T &= \{\exp(\text{Var}_t \ln S_T) - 1\} \exp(2\mathbb{E}_t \ln S_T + \text{Var}_t \ln S_T) \\ &= \{\exp(\text{Var}_t \ln S_T) - 1\} (\mathbb{E}_t S_T)^2 \\ &= (b_2^2 - 1) b_1^2 b_2^2 g_T^2. \end{aligned}$$

Note that for $\kappa > 0$, $\text{Var}_t S_T = 0$ as $T \rightarrow \infty$, because $b_1 = S_t/g_t$ and $b_2 = \exp(\sigma^2/(4\kappa))$ as $T \rightarrow \infty$. Moreover, it is quite straightforward to show that for $\kappa = 0$, $\lim_{T \rightarrow \infty} \text{Var}_t S_T = 0$ if $p + q \geq \sigma^2$. Thus even though $\text{Var}_t \ln S_T$ is increasing for all T , $\text{Var}_t S_T$ can be decreasing, especially for large T .

The above arguments can be seen to hold true in Figure 1 (a) and Figure 3 (a), where both the mean and variance visibly decrease after seven years. They both eventually decline to zero, even if $\kappa = 0$. If, however, the volatility σ were high enough (that is, very large), the mean and the variance would increase without bound.

Forecasting with our model is illustrated in Figure 4. The stochastic path of the sales volume is simulated with Eq. (5) up to the fourth year. Then by using the true parameter values (as used in the simulation),

future sales volumes are forecast by taking the conditional means using Eq. (11). In plot (a), the actual sales volume at time 4 is above the Bass curve; consequently, the expected future sales volumes must be adjusted upward from the Bass curve. The forecast of future sales under model A converges to the Bass curve quite fast (note that $\kappa = 1$), whereas the error model C implies that the forecast of the sales never meet the Bass curve, and hence, model C implies that in this case, the future sales are expected to remain substantially positive forever. Note that according to the other conventional error model B, the expected sales lie on the Bass curve and the sales are expected to *jump* to the Bass curve immediately from the current observed level. That is, as the observed sales level at time 4 is about 20320, the expected sales level after one day (at time 4.0027) is the Bass value at that time, 15070. This would mean -25% expected change in *one day*.

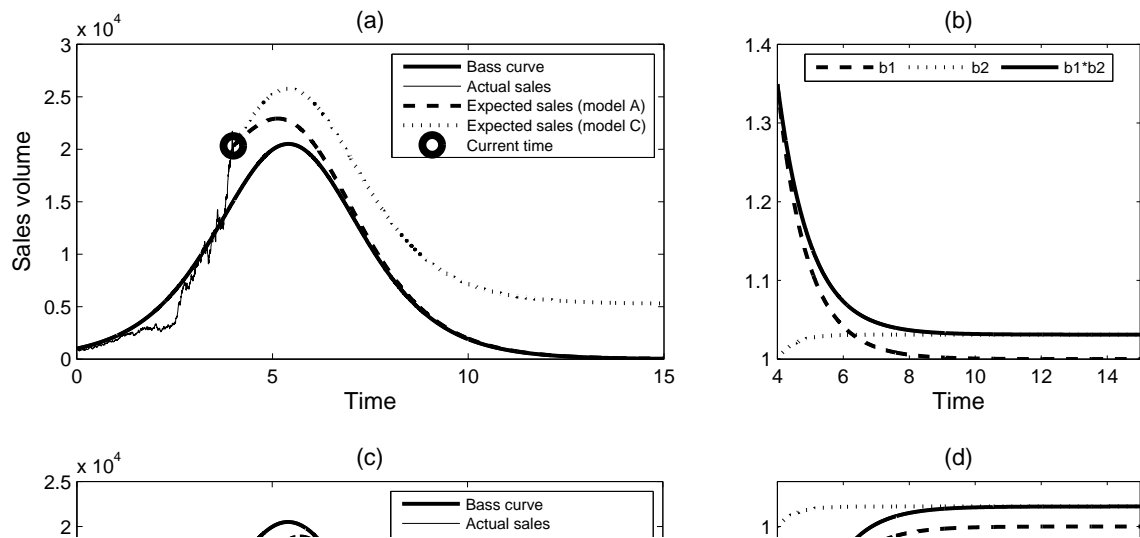


Fig. 4. Illustration of the adjustment of expectation of future sales volume. The product was launched four years ago, and the sales volume has evolved stochastically over the past four years. The figure shows how the adjusted expected values differ from the Bass curve. Plot (b) shows the adjustment factors b_1 and b_2 of plot (a), and plot (c) shows the factors of plot (c). Parameter values are $x_0 = 0$; $p = 0.01$; $q = 0.8$; $m = 100000$; $\kappa = 1$; $\sigma = 0.35$; $\Delta t = 1/365$.

Plot (b) shows the adjustment factors b_1 and b_2 related to plot (a). Both are now greater than one, which is intuitive. Moreover, b_1 is decreasing and b_2 increasing, which is in line with our above analysis. Their product, $b_1 b_2$ is decreasing. Plot (c) shows a situation in which the observed sales volume at time 4 is less than originally predicted with the Bass model. In this case, expected sales volumes lie below the Bass curve.³ A great problem with the conventional error-term (model C) here is that the forecasts of future sales become negative at time 8.4 and finally the forecast is substantially negative. This happens due to

³To be precise, expected sales volumes lie below the Bass curve for a low T , but, as plot (d) shows, $b_1 b_2$ is greater than one for large T . This, however, has no great absolute effect on expected sales for large T , because the Bass model eventually drives the sales down.

TABLE I

THE MAIN PROPERTIES OF THE DATA SET. THE LENGTH OF THE TIME INTERVAL BETWEEN THE OBSERVATIONS IS THREE MONTHS.

	Japan	Finland	UK	US	Australia	Korea
First time point	Dec 1981	Mar 1982	Mar 1985	Sep 1986	Mar 1987	Mar 1984
Last time point	Jun 1996	Dec 1995	Dec 1995	Mar 1998	Dec 1996	Sep 1996
Num of obs.	59	56	44	47	40	51

the conventional assumption of normally distributed sales. Note that in the case of (c), both b_1 and b_2 are increasing (see plot d). If κ were greater (resp. lesser), expected sales would more quickly (resp. more slowly) converge to the Bass curve.

IV. EMPIRICAL ANALYSIS

This section aims to demonstrate empirical estimation procedures and forecasting and it examines the usefulness of our error-term characterization. We estimate models A, B, and C using time series data for analogue mobile phones in six countries using NLS approach.⁴ To present an extensive out-of-sample test, we estimate the three models 30 times for country. In particular, the models are estimated at the 30 last time series points using the data up to these times, i.e. the estimates were obtained using the data from the first observation to the observation of $n - 29, n - 28, \dots, n$, where n is the number of the observation. Thus, we estimate the three models 180 times, and altogether calculate 540 model estimations. In fact, this empirical analysis is among the most extensive out-of-sample analysis we have encountered in the literature.

Our data consisted of six countries selected to represent different cultural, economic, social and legal environments. The sources for our data were International Telecommunication Union (ITU), UN (United Nations), and OECD (Organisation for Economic Co-operation and Development), equipment vendors, telecommunication operators and trade journals and magazines like Telecommunications, Communications International, Cellular Business etc. Multiple data sources were used to validate our original data set of quarterly data and to additionally ensure its reliability. We further calculated the sales figures from the original adoption time series in order to have at least 40 consecutive positive sales time series points. We have attempted to validate our approach with empirical analysis that considers the merits of our approach as thoroughly as possible in one technology area, rather than only select one or two time series from different technological areas in order to increase generalizability of our results. Further, for practical reasons we had to limit our empirical testing to six countries due to the sheer number of estimations. Therefore, we settled on two countries from EU, namely UK and Finland, USA, and three countries from APAC-region, namely, Australia, Japan, and Korea. Table I summarizes our data set.

We estimate model A according to Eq. (6), and we estimate also the initial error, X_1 , which can be positive or negative. Thus, the parameter set of model A is $\{p, q, m, \psi, \sigma, X_1\}$. Models B and C can be estimated using equations (8) and (9). We estimate the initial error, X_1 , also for model C.

Tables V – XIV (Appendix) summarize NLS estimates for models A, B, and C for all the six countries. The first row presents the estimates at the first time point and the second row presents the standard error of the coefficients, and so on. The parameters are presented in terms of three-month periods, but the parameters can be easily converted to annual levels.⁵ We observe that the data is strongly mean-reverting: On average (and in terms of three-month periods), ψ is between 0.147 (Korea) and 0.352 (Australia), depending on the country. This means that the annual value of κ is, on average, between 2.59 and 3.43,

⁴Schmittlein and Mahajan [22] proposed a maximum likelihood estimation (MLE) approach that would yield better forecasting accuracy and more stable parameter estimates than the ordinary least squares. Srinivasan and Mason [20] proposed the NLS approach, because the MLE approach underestimates standard errors of the estimated parameters.

⁵The conversion formulas are: $p_a = p_q/\Delta t$; $q_a = q_q/\Delta t$; $m_a = m_q\Delta t$; $\sigma_a = \sigma_q/\sqrt{\Delta t}$; $\kappa_a = \kappa_q/\Delta t = (1 - \psi_q)/\Delta t$, where $\Delta t = 1/4$, and where x_a represents parameter annual value and x_q parameter value in the terms of three months (quarter). Note that for model B, $\sigma_a^B = \sigma_q^B$.

which represents a very high level of mean reversion. It is important to note here that these estimation results support the assumption made about the mean-reverting error process. If we ignored the property of mean reversion in the error process (that is, if we assumed that $\kappa = 0$), our forecasts would be worse. Naturally, κ and σ are not shown for model B and C because these parameters do not appear in these models.

We perform our analysis as follows (let us take UK data as an example): First, we estimate models A, B, and C using the in-sample-data $(\hat{S}_1, \hat{S}_2, \dots, \hat{S}_{n-i})$, at time points $n-i$, $i = 0, 1, \dots, n-1$, where \hat{S}_i denotes the observed adoption at i th time point. That is, for UK (see Table VII) we estimate the three models 30 times using the sets of observations of $\{S_1, S_2, \dots, S_{14}\}$, $\{S_1, S_2, \dots, S_{15}\}$, \dots , $\{S_1, S_2, \dots, S_{44}\}$. Then we calculate the forecasting errors as squared errors for each model:

$$\begin{aligned} \mathbf{SE}_{n-i,j}^A &= \left(\hat{S}_{n-i+j} - \mathbb{E}_{n-i}^A(S_{n-i+j}) \right)^2 = \left(\hat{S}_{n-i+j} - g_{n-i+j}^A b_1(n-i, n-i+j) b_2(n-i, n-i+j) \right)^2, \\ \mathbf{SE}_{n-i,j}^B &= \left(\hat{S}_{n-i+j} - \mathbb{E}_{n-i}^B(S_{n-i+j}) \right)^2 = \left(\hat{S}_{n-i+j} - g_{n-i+j}^B \right)^2, \\ \mathbf{SE}_{n-i,j}^C &= \left(\hat{S}_{n-i+j} - \mathbb{E}_{n-i}^C(S_{n-i+j}) \right)^2 = \left(\hat{S}_{n-i+j} - \left(g_{n-i+j}^C + \hat{S}_{n-i} - g_{n-i}^C \right) \right)^2, \end{aligned}$$

where for $n = 30$, $i = 1, 2, \dots, n-1$ and $j = 1, 2, \dots, i$, \hat{S}_{n-i+j} is the observation at time $n-i+j$, and $\mathbb{E}_{n-i}(S_{n-i+j})$ is the forecast of sales level at time $n-i+j$ based on the information up to time $n-i$.⁶ We emphasize here that only past data is used such that the end of fitting equals the start of forecasting. Moreover, g^A is the Bass model with parameters estimated with error model A, and similarly g^B and g^C are the Bass models whose parameters are estimated with error models B and C, respectively. The forecasting error matrices are too large to be presented in the paper, but they are available upon request.⁷ After that, we take the mean squared errors for i step predictions, $i = 1, 2, \dots, n-1$:

$$\mathbf{MSE}_i = \sum_{j=1}^i \mathbf{SE}_{n-i,j}$$

Thus, \mathbf{MSE}_i is the average forecasting error when predicting i periods ahead. These average forecasting error vectors (1×29) are calculated for each model and each country, making altogether 18 vectors. The results are summarized in Table II, which shows the ratios of $\mathbf{MSE}_i^B / \mathbf{MSE}_i^A$ and $\mathbf{MSE}_i^C / \mathbf{MSE}_i^A$ for $i = 1, 2, \dots, n-1$ period forecasts. The summary of the results comparing MSEs are shown in Table II.

We observe that model A outperforms models B and C in the most of the cases. Specifically, for Japanese data, in 22 of the 29 cases model A yields more accurate forecasts than model B and in 29 of the 29 cases (i.e. in every case) model A outperforms model C. The corresponding numbers for the other countries are (first A/B and then A/C): Finnish 26/29, 19/29; UK 26/29, 20/29; US 26/29, 18/29; Australian 25/29, 15/29; and Korean 16/29, 29/29. Thus, with respect to the number of ratios that are more than one, for all the countries model A outperforms the alternative models. Japanese data seems to favor model A the most. The worst for model A results were obtained with Australian data in comparison to model C and Korean data in comparison to model B. When we take averages over the ratios of MSEs according to the countries, i.e., we calculate

$$\sum_{i=1}^{n-1} \frac{\mathbf{MSE}_i^B}{\mathbf{MSE}_i^A} \quad \text{and} \quad \sum_{i=1}^{n-1} \frac{\mathbf{MSE}_i^C}{\mathbf{MSE}_i^A}$$

for each country, we see that the average ratios are clearly greater than one for each country (see the second last row of Table II). On the other hand, if we take the average over countries for given values of i , $i = 1, 2, \dots, n-1$, we again see that for almost each forecasting period length, i , model A yields

⁶To be more specific, the expectation $\mathbb{E}_{n-i}(\cdot)$ is based on parameters values that are estimated using the observations $1, 2, \dots, n-i$.

⁷We have 18 30-by-30 matrices, that is three for each six country.

more accurate forecasts than the other models. Specifically, the average ratio of models C to model A (with respect to countries) is always greater than one, and the ratio of the models B to A is less than one only for $i = 2$. Overall, our results confirm that the assumptions of log-normality and mean reversion are plausible and they can make the forecasting more accurate.

V. CONCLUSIONS

This paper dealt with a new stochastic extension to the Bass model for the diffusion of innovations in order to provide more realistic forecasts of future sales. Conventional assumptions about a normally distributed error process without mean reversion can yield unreasonable forecasts, such as negative sales. Our model assumes a log-normal and mean-reverting error process and under our characterization sales volumes are strictly positive and heteroscedastic with autoregressive error structure. The heteroscedasticity here means that sales fluctuate more when they are high and less when they are low. Mean reversion drives the stochastic sales process towards the Bass curve, and this can make especially the long-term forecasts more accurate. We showed that the forecasts of future sales must substantially be adjusted upward from the Bass curve due to the log-normality, especially when the error process is not mean-reverting and volatility is high. Therefore, even if we rely on the Bass model as a good deterministic part of our model, we need to adjust our forecast. We analyzed the model empirically by using a large set of telecommunication data and found that the error process displayed substantially mean revision, implying a fast pull-back to the Bass curve. We compared our error model with conventional ones and found that heteroscedastic sales and mean-reverting error process can make forecasts more accurate. The greater forecasting accuracy is reasonable because the error-term was estimated to be quite strongly mean-reverting, hence driving the stochastic sales back to the Bass curve.

Our results present several promising avenues in relation to engineering management and its utilization of forecasting results. For example, as volatility increases in the diffusion of innovations, future development projects become increasingly risky and this further highlights the need to pay special attention to their management. This insight might direct the engineering effort and further guide road mapping of a company. Naturally, improved forecasting accuracy is important as such, but the explicit consideration of error terms in innovation diffusion modelling may guide the organization's engineering efforts.

The existing literature on the diffusion of innovations has traditionally dealt with the problem of selecting a model mostly by comparing deterministic models or deterministic parts of models to produce favorable fit statistics or, at best, good forecasting capability. However, this represents only one side of the forecasting problem. As this study has shown, characterizing the error process can greatly impact the performance of forecasting. Hence, when selecting a model, one should consider not only the model's deterministic term but also its stochastic term. We anticipate future work to address this point and we view our characterization as one possibility among others. A natural extension of our study is to consider also other diffusion models with log-normal and mean-reverting error process, such as the Gompertz model. Moreover, one may also want to consider other possible error structures than presented in this paper. For example, Poisson processes could be used to model extreme changes in the sales. Furthermore, higher-order autoregressive error processes should be studied in discrete time for larger data sets. Finally, this paper also serves as a preliminary starting point for real option valuation with stochastic product life cycles in continuous time.

TABLE II

OUT-OF-THE-SAMPLE FORECASTING ERRORS FOR 1-29 PERIOD FORECASTS (THE LENGTH OF ONE PERIOD IS THREE MONTH) FOR SIX COUNTRIES. HERE B/A MEANS MSE_t^B / MSE_t^A AND C/A MEANS MSE_t^C / MSE_t^A . THE STANDARD DEVIATIONS (OVER THE PERIODS AND OVER THE COUNTRIES) ARE NOT CALCULATED BECAUSE THE RATIOS CANNOT BE ASSUMED TO BE NORMALLY DISTRIBUTED.

Periods ahead	Japanese Data		Finnish Data		UK Data		US Data		Australian Data		Korean Data		Average B/A over countries		Average C/A over countries	
	B/A	C/A	B/A	C/A	B/A	C/A	B/A	C/A	B/A	C/A	B/A	C/A	B/A	C/A	B/A	C/A
1	1.455	1.019	1.107	1.444	1.296	1.197	0.870	1.539	1.091	1.216	1.101	1.484	1.153	1.316		
2	0.972	1.118	1.012	1.723	1.164	2.781	0.932	2.448	1.046	1.224	0.927	1.500	1.009	1.799		
3	0.916	1.523	0.998	1.231	1.116	3.636	0.907	2.461	0.941	1.162	0.661	1.272	0.923	1.881		
4	1.024	1.275	1.075	2.283	1.049	4.165	1.079	4.477	1.026	0.912	1.075	1.897	1.055	2.501		
5	1.111	1.311	1.267	4.423	1.182	3.451	1.426	5.107	0.626	0.646	1.159	2.819	1.129	2.959		
6	1.212	1.339	1.361	7.016	1.174	2.439	1.442	4.662	0.834	0.706	1.007	3.062	1.172	3.204		
7	1.150	1.315	1.441	10.269	1.181	1.782	1.584	2.167	1.089	0.731	1.177	2.181	1.270	3.074		
8	1.181	1.443	1.681	17.931	1.099	1.492	2.003	1.861	1.272	0.538	1.497	3.434	1.456	4.450		
9	1.241	1.501	1.965	25.847	1.036	1.149	1.940	2.137	1.512	0.598	1.843	3.732	1.589	5.827		
10	1.359	1.843	2.253	32.661	1.014	0.832	1.674	2.066	1.376	0.715	2.120	4.001	1.633	7.020		
11	1.646	2.259	2.631	35.195	0.975	0.756	1.625	1.357	1.304	0.775	2.186	4.190	1.728	7.422		
12	1.989	3.280	3.073	33.280	0.964	0.593	1.647	1.068	1.215	0.751	1.948	3.912	1.806	7.147		
13	3.932	6.196	3.598	25.430	0.961	0.599	1.345	1.287	1.157	0.783	2.304	4.247	2.216	6.424		
14	3.684	5.896	4.358	16.655	1.029	0.606	1.475	1.233	1.177	0.897	1.957	3.236	2.280	4.754		
15	2.676	3.813	5.182	9.987	1.095	0.579	1.310	1.312	1.120	0.925	1.490	2.726	2.145	3.224		
16	1.774	2.375	5.295	6.164	1.145	0.628	1.226	1.057	1.118	1.010	2.097	4.258	2.109	2.582		
17	1.716	2.069	5.937	4.797	1.292	0.709	1.189	1.048	1.081	0.988	1.724	4.340	2.156	2.325		
18	1.764	2.263	5.865	1.428	1.548	0.868	1.126	0.938	1.147	1.072	1.011	4.122	2.077	1.782		
19	1.794	2.139	6.097	0.963	2.107	1.317	1.076	0.998	1.124	1.115	0.898	4.302	2.182	1.806		
20	1.498	2.012	6.296	0.750	2.383	1.874	1.146	0.875	1.156	1.045	0.993	3.294	2.245	1.642		
21	1.324	1.734	7.476	0.683	4.443	3.509	1.100	0.665	1.210	1.156	0.834	3.041	2.731	1.798		
22	1.129	1.645	6.382	0.698	6.497	5.980	1.028	0.879	1.228	1.171	0.754	3.429	2.836	2.300		
23	1.082	1.400	4.995	0.730	6.852	5.961	1.036	0.833	1.322	1.314	0.801	3.526	2.681	2.294		
24	0.964	1.252	4.288	0.412	5.986	6.124	1.045	0.879	1.211	1.246	0.785	3.133	2.380	2.174		
25	0.930	1.034	4.316	0.449	7.528	7.438	1.056	0.792	1.192	0.710	0.744	2.499	2.628	2.153		
26	0.886	1.241	3.467	0.606	4.981	8.527	1.060	1.038	1.552	1.053	0.760	2.479	2.118	2.490		
27	0.914	1.588	1.756	0.283	5.471	6.012	1.068	0.916	5.373	5.956	0.653	2.590	2.539	2.891		
28	0.799	4.320	0.318	0.586	4.205	3.723	1.072	0.917	0.762	2.120	0.665	1.974	1.304	2.273		
29	1.495	1.781	0.183	1.257	2.732	1.880	1.085	0.897	1.661	1.907	0.713	1.807	1.312	1.588		
Average over periods	1.504	2.137	3.299	8.454	2.535	2.780	1.261	1.652	1.308	1.188	1.237	3.051	1.312	1.588		

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APPENDIX

The following tables summarize NLS estimates for models A, B, and C for all the six countries. The first row presents the estimates at the first time point and the second row presents the standard error of the coefficients, and so on.

TABLE III
NLS ESTIMATES FOR JAPANESE DATA, MODEL A.

Data used up to (yyymmdd)	Num of obs.	Model A					
		p	q	m	ψ	σ	X_1
19881201	29	0.000158 (0.000708)	0.113826 (0.011956)	5367081 (24450123)	0.186452 (0.063975)	0.137898 (0.004994)	13.385648 (4.343982)
19890301	30	0.000156 (0.000585)	0.114969 (0.011084)	5367081 (20581702)	0.193031 (0.063155)	0.136926 (0.004841)	12.998911 (4.020694)
19890601	31	0.000148 (0.000492)	0.118011 (0.011074)	5459749 (18702669)	0.209859 (0.066391)	0.145233 (0.005358)	12.132390 (3.612113)
19890901	32	0.000138 (0.000415)	0.120095 (0.010811)	5646678 (17432945)	0.226993 (0.067180)	0.147288 (0.005423)	11.337574 (3.137939)
19891201	33	0.000135 (0.000366)	0.125857 (0.012590)	5320947 (14990711)	0.267847 (0.078494)	0.173365 (0.007399)	9.892849 (2.663520)
19900301	34	0.000134 (0.000374)	0.128632 (0.015548)	5297960 (15565332)	0.328113 (0.091973)	0.203449 (0.010039)	7.244156 (1.820171)
19900601	35	0.000113 (0.000343)	0.135586 (0.020946)	5297940 (17275862)	0.496025 (0.094754)	0.210337 (0.010576)	5.618356 (0.875487)
19900901	36	0.000114 (0.000293)	0.135586 (0.019598)	5297940 (14841300)	0.496025 (0.093006)	0.208058 (0.010203)	5.618356 (0.867218)
19901201	37	0.000115 (0.000247)	0.135586 (0.018208)	5297960 (12490546)	0.496024 (0.091291)	0.209394 (0.010194)	5.618357 (0.872875)
19910301	38	0.000114 (0.000208)	0.135586 (0.016891)	5297960 (10707674)	0.496024 (0.089282)	0.209044 (0.010025)	5.618358 (0.867286)
19910601	39	0.000115 (0.000164)	0.135586 (0.014929)	5297960 (8502291)	0.496024 (0.083191)	0.207151 (0.009718)	5.618358 (0.846121)
19910901	40	0.000115 (0.000134)	0.135586 (0.013638)	5297940 (6991853)	0.496024 (0.080065)	0.206211 (0.009508)	5.618357 (0.837999)
19911201	41	0.000061 (0.000083)	0.137261 (0.009304)	9791364 (14491901)	0.371134 (0.071860)	0.189913 (0.007966)	7.631591 (1.374865)
19920301	42	0.000056 (0.000077)	0.135015 (0.008574)	10991454 (16112480)	0.355999 (0.071680)	0.190815 (0.007945)	7.854021 (1.492206)
19920601	43	0.000063 (0.000065)	0.134635 (0.008051)	9871671 (11041615)	0.370088 (0.068984)	0.189683 (0.007760)	7.447495 (1.333255)
19920901	44	0.000222 (0.000058)	0.152143 (0.016235)	2192764 (732276)	0.437775 (0.115052)	0.316571 (0.021366)	6.884820 (1.635294)
19921201	45	0.000199 (0.000053)	0.144182 (0.011494)	2801125 (867923)	0.364674 (0.091827)	0.317906 (0.021306)	7.896236 (2.014037)
19930301	46	0.000211 (0.000049)	0.146413 (0.011145)	2563483 (614262)	0.374956 (0.090124)	0.316142 (0.020840)	7.779784 (1.902082)
19930601	47	0.000212 (0.000046)	0.146733 (0.010530)	2535276 (514994)	0.376414 (0.087908)	0.312797 (0.020183)	7.761305 (1.859692)
19930901	48	0.000216 (0.000045)	0.147385 (0.010055)	2469177 (423985)	0.377760 (0.086496)	0.309623 (0.019569)	7.704636 (1.818120)
19931201	49	0.000191 (0.000042)	0.141150 (0.009507)	3059799 (627930)	0.359391 (0.089446)	0.322308 (0.020987)	7.912872 (2.063625)
19940301	50	0.000198 (0.000040)	0.142406 (0.009190)	2902857 (489913)	0.353020 (0.088248)	0.320247 (0.020512)	8.099929 (2.102232)
19940601	51	0.000176 (0.000038)	0.136885 (0.008894)	3531759 (710504)	0.342581 (0.091622)	0.332874 (0.021943)	8.096854 (2.281117)
19940901	52	0.000155 (0.000037)	0.133338 (0.009004)	4205476 (993267)	0.364344 (0.092082)	0.337918 (0.022394)	7.450183 (2.022227)
19941201	53	0.000126 (0.000036)	0.129464 (0.009197)	5395881 (1647232)	0.388919 (0.092726)	0.344343 (0.023034)	6.893205 (1.803713)
19950301	54	0.000109 (0.000034)	0.135574 (0.011433)	5297960 (1542429)	0.496016 (0.094177)	0.356164 (0.024413)	5.618369 (1.210251)
19950601	55	0.000108 (0.000032)	0.135572 (0.011388)	5297960 (1412117)	0.496022 (0.092980)	0.359195 (0.024603)	5.618361 (1.193958)
19950901	56	0.000109 (0.000032)	0.133998 (0.011617)	5379844 (1441682)	0.496166 (0.095733)	0.373926 (0.026424)	5.618331 (1.231455)
19951201	57	0.000074 (0.000028)	0.123099 (0.009162)	9885429 (4376830)	0.422607 (0.091920)	0.361592 (0.024491)	6.246105 (1.539155)
19960301	58	0.000109 (0.000030)	0.133749 (0.010961)	5461761 (1222513)	0.490775 (0.093002)	0.371624 (0.025645)	5.661810 (1.224327)
19960601	59	0.000111 (0.000032)	0.135586 (0.011297)	5297940 (1057019)	0.496024 (0.095770)	0.390608 (0.028091)	5.618414 (1.258663)

TABLE IV
NLS ESTIMATES FOR JAPANESE DATA, MODELS B AND C.

Data used up to (yyymmdd)	Num of obs.	Model B			Model C			
		p	q	m	p	q	m	X_1
19881201	29	0.000143909 (0.001085717)	0.115993 (0.032427)	5827921 (46391342)	4.187E-05 (0.000239)	0.314293 (0.259322)	188153 (253864)	11548.29 (2331.93)
19890301	30	0.000129484 (0.00074014)	0.120297 (0.028641)	5888590 (36011163)	4.624E-06 (0.000324)	0.226477 (0.492250)	6211129 (504695768)	11522.94 (2460.43)
19890601	31	9.70051E-05 (0.000375041)	0.138224 (0.030127)	5156829 (22441037)	2.196E-05 (0.000921)	0.171280 (0.342919)	6725480 (334131910)	11383.70 (2811.84)
19890901	32	7.74888E-05 (0.000214695)	0.147020 (0.027391)	5156842 (16763808)	9.888E-08 (6.11E-06)	0.334634 (0.417638)	9741835 (707794463)	11381.78 (2470.62)
19891201	33	2.1498E-05 (4.78371E-05)	0.195777 (0.042771)	4802222 (14960868)	1.175E-07 (7.36E-07)	0.451762 (0.232568)	321422 (297121)	11558.02 (2547.01)
19900301	34	4.86085E-05 (8.99912E-05)	0.171589 (0.044832)	5113805 (14060296)	7.714E-10 (6.16E-09)	0.500951 (0.356223)	5143063 (90372890)	10262.63 (3118.36)
19900601	35	3.85294E-05 (2.61782E-05)	0.187691 (0.037509)	4138035 (5821981)	0.0002811 (0.000422)	0.139096 (0.113272)	5134110 (11655602)	0.00 (7743.18)
19900901	36	2.73987E-05 (1.12206E-05)	0.198812 (0.032784)	3941623 (3812359)	8.03E-05 (0.000152)	0.167942 (0.137286)	5133939 (14994014)	0.47 (5996.02)
19901201	37	9.91382E-06 (6.32647E-06)	0.257557 (0.028196)	1898392 (452479)	5.238E-10 (1.2E-09)	0.594449 (0.070263)	595734 (129630)	11600.66 (4001.10)
19910301	38	1.30416E-06 (1.04273E-06)	0.341618 (0.026879)	1146458 (90822)	1.861E-06 (5.54E-06)	0.319502 (0.096223)	1346296 (662399)	11555.56 (5736.17)
19910601	39	3.2529E-06 (2.19667E-06)	0.308121 (0.022628)	1303141 (94798)	2.826E-07 (6.7E-07)	0.391507 (0.072299)	1095690 (393718)	11536.09 (6668.11)
19910901	40	1.80186E-05 (1.06971E-05)	0.242223 (0.021376)	1791627 (192710)	9.129E-06 (2.56E-05)	0.274251 (0.087224)	1515072 (865462)	11541.49 (8172.27)
19911201	41	6.545E-05 (2.54577E-05)	0.169228 (0.022810)	4104230 (1568218)	0.0001676 (0.000484)	0.144110 (0.124816)	4478809 (7878110)	7026.41 (9553.17)
19920301	42	0.000151242 (4.32706E-05)	0.138133 (0.020603)	5133951 (2374496)	0.0001364 (0.000461)	0.140036 (0.153995)	5134120 (12965820)	0.01 (11059.67)
19920601	43	4.63932E-05 (2.22716E-05)	0.194952 (0.017558)	2769879 (335893)	2.375E-05 (8.29E-05)	0.221749 (0.103433)	2394077 (1910929)	11925.34 (12264.51)
19920901	44	8.303E-06 (7.22242E-06)	0.261681 (0.026285)	1853944 (154422)	7.114E-05 (0.00024)	0.181285 (0.102668)	3093912 (2890831)	11295.65 (12374.98)
19921201	45	8.73061E-06 (6.93758E-06)	0.260029 (0.023536)	1864459 (134878)	1.058E-09 (5.56E-09)	0.481469 (0.135523)	1389042 (804551)	11530.56 (18647.25)
19930301	46	1.02361E-05 (7.45431E-06)	0.255018 (0.021283)	1893924 (124081)	4.832E-06 (2.83E-05)	0.272949 (0.165223)	1762416 (2203113)	11542.07 (21964.41)
19930601	47	1.32997E-05 (9.04225E-06)	0.246281 (0.019692)	1954824 (121865)	1.995E-06 (1.35E-05)	0.296761 (0.187701)	1468094 (1865902)	11554.44 (21734.56)
19930901	48	1.92492E-05 (1.22692E-05)	0.234312 (0.018378)	2034667 (124059)	1.008E-06 (8.11E-06)	0.316790 (0.222020)	1142357 (1532743)	11594.57 (21655.95)
19931201	49	9.56798E-05 (7.0157E-05)	0.179290 (0.022299)	2548214 (258425)	8.084E-05 (0.000359)	0.194633 (0.132330)	2554961 (3536487)	9496.38 (21596.67)
19940301	50	0.000131954 (8.87154E-05)	0.167239 (0.020529)	2716574 (272914)	3.173E-05 (0.000473)	0.132904 (0.730786)	5137121 (106221348)	9.91 (26054.81)
19940601	51	0.000367752 (0.000219492)	0.117952 (0.022773)	4167677 (947763)	6.444E-06 (6.6E-05)	0.273654 (0.291438)	1055719 (2064946)	11567.64 (27446.82)
19940901	52	0.000331373 (0.000138753)	0.094548 (0.025149)	7917779 (5145751)	3.038E-05 (0.000554)	0.132916 (0.718917)	5136332 (86907080)	9.95 (31604.94)
19941201	53	9.33087E-05 (0.000154703)	0.092480 (0.029094)	23484398 (57816253)	2.608E-05 (0.0005)	0.132957 (0.736657)	5134110 (90296507)	0.03 (31710.49)
19950301	54	1.75214E-05 (3.14969E-05)	0.168859 (0.050840)	5134138 (2237851)	2.032E-05 (0.000441)	0.132933 (0.863310)	5134110 (116613006)	0.02 (33023.99)
19950601	55	4.85185E-05 (8.61249E-05)	0.148877 (0.051061)	5134133 (2225721)	1.57E-05 (0.000401)	0.132899 (1.066352)	5134110 (160379129)	0.02 (35945.42)
19950901	56	8.06642E-05 (0.000203146)	0.133326 (0.076282)	5134110 (3968104)	1.814E-05 (0.000427)	0.132913 (0.803126)	5134110 (95912619)	0.02 (35489.21)
19951201	57	8.06642E-05 (0.000193855)	0.133326 (0.069601)	5134110 (3329657)	1.244E-05 (0.000418)	0.132788 (1.284850)	5137799 (189065605)	9.87 (44384.33)
19960301	58	8.06642E-05 (0.000184087)	0.133326 (0.063509)	5134110 (2811456)	4.757E-05 (0.00083)	0.139943 (0.397765)	5134100 (24219446)	0.00 (71924.62)
19960601	59	8.06642E-05 (0.000176726)	0.133326 (0.058835)	5134110 (2426420)	7.449E-05 (0.001117)	0.134360 (0.338619)	5134100 (22729659)	0.00 (72694.22)

TABLE V
NLS ESTIMATES FOR FINNISH DATA, MODEL A.

Data used up to (yyyymmdd)	Num of obs.	Model A					
		p	q	m	ψ	σ	X_1
19880601	26	0.002878 (0.000619)	0.134259 (0.015610)	202021 (57990)	0.000157 (0.188434)	0.201455 (0.011256)	-3874.412 (4635448)
19880901	27	0.002796 (0.000510)	0.132926 (0.013786)	209443 (50515)	0.000258 (0.174656)	0.199343 (0.010815)	-2384.863 (1611882)
19881201	28	0.002509 (0.000493)	0.128302 (0.013066)	239317 (61456)	0.000625 (0.172482)	0.200517 (0.010746)	-1018.654 (281021.6)
19890301	29	0.001972 (0.000596)	0.119672 (0.013547)	320658 (120643)	0.030609 (0.173651)	0.208131 (0.011376)	-22.24901 (125.2218)
19890601	30	0.001165 (0.000881)	0.108249 (0.015213)	586146 (500975)	0.125943 (0.168276)	0.213926 (0.011816)	-5.930361 (7.801703)
19890901	31	0.001112 (0.000736)	0.107564 (0.013628)	617077 (462423)	0.131379 (0.156662)	0.210482 (0.011253)	-5.714994 (6.793448)
19891201	32	0.001006 (0.000655)	0.106243 (0.012654)	687762 (505698)	0.135449 (0.153537)	0.207385 (0.010752)	-5.599067 (6.338110)
19900301	33	0.000862 (0.000604)	0.104399 (0.011881)	812605 (636289)	0.141257 (0.150839)	0.204712 (0.010317)	-5.445232 (5.812777)
19900601	34	0.000663 (0.000576)	0.103128 (0.011516)	1056977 (1006056)	0.156545 (0.151862)	0.207554 (0.010448)	-4.666754 (4.577451)
19900901	35	0.000636 (0.000497)	0.101994 (0.010427)	1116997 (957503)	0.143538 (0.145083)	0.204704 (0.010017)	-5.439798 (5.548084)
19901201	36	0.000814 (0.000383)	0.104363 (0.009685)	858864 (462631)	0.144262 (0.145141)	0.204868 (0.009893)	-5.319091 (5.424780)
19910301	37	0.000961 (0.000299)	0.106735 (0.009149)	715446 (265554)	0.156910 (0.143487)	0.203823 (0.009659)	-4.798733 (4.483176)
19910601	38	0.000821 (0.000288)	0.104273 (0.008735)	853081 (353328)	0.144853 (0.142539)	0.203263 (0.009479)	-5.306700 (5.266427)
19910901	39	0.001011 (0.000204)	0.108406 (0.007811)	668432 (165388)	0.113923 (0.143045)	0.206375 (0.009645)	-6.465999 (8.206378)
19911201	40	0.001191 (0.000172)	0.112769 (0.007928)	548654 (94704)	0.156520 (0.146126)	0.213499 (0.010192)	-4.520241 (4.387373)
19920301	41	0.001220 (0.000153)	0.113569 (0.007635)	532559 (77297)	0.170791 (0.139742)	0.211264 (0.009858)	-4.112717 (3.530323)
19920601	42	0.001122 (0.000146)	0.110287 (0.007493)	595702 (94509)	0.155819 (0.141890)	0.214906 (0.010078)	-4.671145 (4.361734)
19920901	43	0.001182 (0.000127)	0.113015 (0.006669)	549454 (63179)	0.117619 (0.138702)	0.215874 (0.010050)	-5.968117 (7.190385)
19921201	44	0.001169 (0.000118)	0.112481 (0.006335)	558219 (58060)	0.113161 (0.136517)	0.213641 (0.009731)	-6.240801 (7.652795)
19930301	45	0.001124 (0.000114)	0.110168 (0.006272)	594620 (63445)	0.122583 (0.137197)	0.214989 (0.009744)	-5.935779 (6.747127)
19930601	46	0.001031 (0.000125)	0.105193 (0.007010)	685026 (96403)	0.185758 (0.140721)	0.224907 (0.010547)	-4.184962 (3.292957)
19930901	47	0.001028 (0.000117)	0.105033 (0.006528)	688246 (84070)	0.188464 (0.130702)	0.222509 (0.010213)	-4.136407 (3.043578)
19931201	48	0.001004 (0.000112)	0.103784 (0.006333)	714484 (84117)	0.196494 (0.129314)	0.221145 (0.009983)	-4.031699 (2.824267)
19940301	49	0.000958 (0.000111)	0.101303 (0.006386)	770347 (96953)	0.217097 (0.129134)	0.222555 (0.010007)	-3.769056 (2.405489)
19940601	50	0.000890 (0.000117)	0.097582 (0.006801)	868034 (132332)	0.263001 (0.128655)	0.226698 (0.010278)	-3.269783 (1.763940)
19940901	51	0.000896 (0.000108)	0.097916 (0.006308)	858329 (111123)	0.257525 (0.122449)	0.224517 (0.009982)	-3.323627 (1.773738)
19941201	52	0.000895 (0.000102)	0.097877 (0.006010)	859546 (99673)	0.257743 (0.120912)	0.222349 (0.009696)	-3.322585 (1.753906)
19950301	53	0.000826 (0.000104)	0.094213 (0.006291)	973464 (137153)	0.269586 (0.124895)	0.229878 (0.010265)	-3.326151 (1.711900)
19950601	54	0.000823 (0.000098)	0.094017 (0.005950)	980033 (123459)	0.273070 (0.118712)	0.227762 (0.009983)	-3.292854 (1.628020)
19950901	55	0.000837 (0.000094)	0.094941 (0.005603)	950932 (101064)	0.268401 (0.117028)	0.225960 (0.009736)	-3.310468 (1.653057)
19951201	56	0.000859 (0.000094)	0.096466 (0.005390)	909745 (81597)	0.266959 (0.117673)	0.227332 (0.009767)	-3.261934 (1.662741)

TABLE VI
NLS ESTIMATES FOR FINNISH DATA, MODELS B AND C.

Data used up to (yyymmdd)	Num of obs.	Model B			Model C			
		p	q	m	p	q	m	X_1
19880601	26	0.0013937 (0.00124)	0.093474 (0.015891)	617589 (629566)	0.0003021 (0.023916)	0.065483 (0.277000)	5507948 (465303101)	-1415.93 (9673.20)
19880901	27	0.0014761 (0.00094)	0.094170 (0.013870)	580708 (438492)	1.052E-08 (1.97E-06)	0.468330 (0.737008)	2111152 (430379086)	359.96 (487.06)
19881201	28	0.0007855 (0.00102)	0.088044 (0.012887)	1137864 (1619605)	0.0015924 (0.011633)	0.064335 (0.206639)	1286421 (14916239)	-1821.24 (10009.94)
19890301	29	0.0001816 (0.00113)	0.088870 (0.014853)	4594021 (29223767)	0.0002999 (0.001608)	0.133234 (0.205521)	738435 (6797189)	223.79 (1275.01)
19890601	30	0.0001502 (0.00089)	0.098404 (0.017268)	4580531 (28172492)	1.155E-07 (3.76E-06)	0.372032 (0.433232)	1382910 (59496333)	359.85 (602.40)
19890901	31	0.0001558 (0.00079)	0.095557 (0.015462)	4690573 (24689198)	6.387E-07 (6.74E-06)	0.317534 (0.268760)	1275860 (21731368)	358.89 (595.25)
19891201	32	0.0001559 (0.00069)	0.093292 (0.013847)	4920974 (22683152)	8.654E-10 (3.26E-09)	0.677180 (0.133600)	55082 (19679)	376.36 (731.41)
19900301	33	0.0001531 (0.00057)	0.092596 (0.012270)	5087179 (19749749)	1.185E-06 (4.12E-06)	0.407090 (0.126827)	95383 (51681)	360.27 (851.83)
19900601	34	0.0001293 (0.00044)	0.100676 (0.014081)	4962919 (18130873)	0.0002214 (0.000544)	0.180874 (0.117753)	339794 (397598)	270.27 (937.21)
19900901	35	0.0001404 (0.00048)	0.094965 (0.013657)	5263828 (19001596)	3.925E-05 (0.000103)	0.197587 (0.179078)	738376 (2479628)	223.80 (1119.45)
19901201	36	0.0007348 (0.00015)	0.112628 (0.014583)	784080 (298598)	6.883E-06 (2.85E-05)	0.311103 (0.142206)	176610 (137698)	376.11 (1461.19)
19910301	37	0.0008184 (0.00015)	0.127604 (0.013927)	538760 (100076)	4.903E-08 (2.31E-07)	0.470503 (0.149907)	101666 (64369)	359.99 (1442.14)
19910601	38	0.0007834 (0.00012)	0.113487 (0.013394)	731677 (193159)	2.482E-06 (9.26E-06)	0.346463 (0.122156)	143543 (102607)	359.50 (1459.05)
19910901	39	0.0007719 (0.00017)	0.131574 (0.013286)	520767 (70982)	8.215E-05 (0.003425)	0.091016 (0.233932)	8324852 (399039783)	-388.94 (5469.92)
19911201	40	0.0006392 (0.00018)	0.149016 (0.013576)	429455 (39233)	4.104E-05 (0.000153)	0.245243 (0.122794)	251701 (251967)	347.06 (2136.78)
19920301	41	0.0005872 (0.00017)	0.153510 (0.012459)	416918 (30920)	2.837E-05 (9.79E-05)	0.257656 (0.113032)	256898 (232410)	350.88 (2124.24)
19920601	42	0.0008316 (0.00022)	0.130399 (0.013112)	507661 (52879)	4.481E-05 (0.000195)	0.244627 (0.144873)	211351 (257344)	347.92 (2187.34)
19920901	43	0.0007739 (0.00021)	0.135939 (0.012233)	481153 (39903)	5.734E-06 (5.87E-05)	0.191961 (0.700707)	747234 (16512891)	222.45 (2506.49)
19921201	44	0.00082 (0.00021)	0.131960 (0.011327)	498313 (38519)	0.0002982 (0.001129)	0.172965 (0.128544)	368324 (554510)	237.41 (2882.99)
19930301	45	0.0009771 (0.00024)	0.119113 (0.011211)	562846 (50738)	0.0009728 (0.003343)	0.124053 (0.123708)	509055 (927886)	-193.54 (3946.67)
19930601	46	0.0011565 (0.00024)	0.094409 (0.013144)	800883 (152150)	0.0001156 (0.006947)	0.073535 (0.433759)	5418666 (399022807)	-159.32 (10637.81)
19930901	47	0.0011564 (0.00024)	0.093908 (0.012168)	808820 (135517)	6.249E-05 (0.000628)	0.131673 (0.481428)	746917 (8464813)	222.83 (3204.03)
19931201	48	0.0011573 (0.00023)	0.091637 (0.011340)	845714 (134260)	0.0017519 (0.005979)	0.076210 (0.145025)	1088205 (3113087)	-1692.84 (12544.44)
19940301	49	0.0011353 (0.0002)	0.084539 (0.010958)	991373 (189322)	0.001898 (0.005961)	0.063395 (0.155775)	1460824 (5423660)	-2585.83 (19043.66)
19940601	50	0.0010026 (0.00019)	0.072686 (0.011233)	1426715 (479667)	8.779E-05 (0.00091)	0.121654 (0.397786)	746342 (5926483)	234.37 (3528.19)
19940901	51	0.0010295 (0.00016)	0.076479 (0.010475)	1270814 (309864)	3.965E-05 (0.000474)	0.135070 (0.431241)	730257 (6134896)	227.03 (3380.43)
19941201	52	0.0010719 (0.00018)	0.079791 (0.009784)	1145260 (206159)	0.00123 (0.004502)	0.079212 (0.130745)	1161284 (2944355)	-1168.26 (9700.91)
19950301	53	0.0009476 (0.00016)	0.069883 (0.010160)	1590743 (482326)	0.0010546 (0.003774)	0.088097 (0.115296)	1006813 (2132727)	-798.24 (7133.65)
19950601	54	0.0009499 (0.00015)	0.071416 (0.009496)	1522001 (378064)	3.647E-05 (0.000523)	0.129811 (0.464843)	746668 (6630994)	229.21 (3945.55)
19950901	55	0.0010084 (0.00016)	0.076123 (0.009004)	1300566 (222425)	0.0002123 (0.001508)	0.115700 (0.186896)	700004 (1848218)	212.55 (4341.50)
19951201	56	0.001015 (0.0002)	0.083254 (0.008962)	1098795 (131437)	3.621E-08 (4.37E-07)	0.308631 (0.253765)	201411 (337724)	359.99 (4040.88)

TABLE VII
NLS ESTIMATES FOR UK DATA, MODEL A.

Data used up to (yyyymmdd)	Num of obs.	Model A					
		p	q	m	ψ	σ	X_1
19880601	14	0.000760 (0.007150)	0.136630 (0.039177)	10434171 (99308176)	-0.408617 (0.290717)	0.191934 (0.013924)	-0.332387 (0.598257)
19880901	15	0.000772 (0.005480)	0.133909 (0.031960)	10428199 (74956631)	-0.491986 (0.249337)	0.187394 (0.012823)	-0.249009 (0.457769)
19881201	16	0.000759 (0.003695)	0.136922 (0.024809)	10432897 (51617668)	-0.561741 (0.224761)	0.186058 (0.012239)	-0.243647 (0.381106)
19890301	17	0.000762 (0.002881)	0.136313 (0.021165)	10431126 (40191767)	-0.577624 (0.208239)	0.180713 (0.011201)	-0.231201 (0.357632)
19890601	18	0.000763 (0.002173)	0.136044 (0.017725)	10429186 (30360181)	-0.572774 (0.192509)	0.175680 (0.010288)	-0.230377 (0.344212)
19890901	19	0.000764 (0.001694)	0.135866 (0.015270)	10429192 (23731051)	-0.574684 (0.186822)	0.171024 (0.009490)	-0.227932 (0.332585)
19891201	20	0.000753 (0.001291)	0.137814 (0.013311)	10431165 (18438425)	-0.588202 (0.185842)	0.171296 (0.009279)	-0.243250 (0.324410)
19900301	21	0.000755 (0.001028)	0.137713 (0.011731)	10424310 (14735716)	-0.590849 (0.178580)	0.167185 (0.008626)	-0.240985 (0.314654)
19900601	22	0.001855 (0.000779)	0.146428 (0.012659)	4114793 (1952506)	-0.499534 (0.202769)	0.194113 (0.011361)	-0.330508 (0.444475)
19900901	23	0.003826 (0.000599)	0.184083 (0.023157)	1651215 (317810)	-0.087374 (0.277901)	0.287902 (0.024442)	-3.644377 (12.761904)
19901201	24	0.003974 (0.000679)	0.196259 (0.022597)	1473515 (181732)	0.076818 (0.203570)	0.286851 (0.023753)	4.985643 (13.477714)
19910301	25	0.003996 (0.000719)	0.203269 (0.021246)	1402095 (133595)	0.109864 (0.197555)	0.284635 (0.022915)	3.824838 (7.211447)
19910601	26	0.003646 (0.001116)	0.237351 (0.031163)	1210256 (126266)	0.246815 (0.243886)	0.352342 (0.034432)	2.532503 (2.644185)
19910901	27	0.003142 (0.001265)	0.258187 (0.034313)	1163434 (144548)	0.404552 (0.190408)	0.353697 (0.034048)	1.956467 (1.297014)
19911201	28	0.003515 (0.001121)	0.243449 (0.026427)	1189221 (126641)	0.318881 (0.172825)	0.355830 (0.033839)	2.108910 (1.703996)
19920301	29	0.003989 (0.001339)	0.221592 (0.027923)	1237471 (140801)	0.330391 (0.186078)	0.385463 (0.039019)	1.599579 (1.708569)
19920601	30	0.003901 (1.075819)	0.003080 (1.444568)	9949523 (2777531563)	0.770752 (0.181976)	0.435222 (0.048908)	-1.707701 (3.862282)
19920901	31	0.004451 (0.001670)	0.202083 (0.029354)	1281198 (160232)	0.418413 (0.168359)	0.385433 (0.037734)	0.966054 (1.280864)
19921201	32	0.005693 (0.003042)	0.159337 (0.046279)	1412815 (260106)	0.550952 (0.179625)	0.428364 (0.045874)	0.189238 (1.244948)
19930301	33	0.002428 (0.050207)	0.039104 (0.213969)	9259769 (205907249)	0.746919 (0.153484)	0.441483 (0.047983)	-1.081086 (2.098604)
19930601	34	0.002202 (0.038422)	0.040194 (0.174500)	10012715 (187872470)	0.750445 (0.131732)	0.435158 (0.045927)	-1.051631 (1.881739)
19930901	35	0.001945 (0.026031)	0.045599 (0.144214)	10452289 (152069481)	0.759101 (0.121077)	0.431215 (0.044450)	-0.940632 (1.765309)
19931201	36	0.001585 (0.013529)	0.058995 (0.120260)	10462755 (100779318)	0.781424 (0.115317)	0.436693 (0.044949)	-0.670109 (1.768120)
19940301	37	0.001478 (0.009882)	0.063381 (0.109621)	10462755 (80901049)	0.792473 (0.106720)	0.431779 (0.043345)	-0.578490 (1.751014)
19940601	38	0.001406 (0.007643)	0.066595 (0.102053)	10430459 (67235930)	0.799777 (0.101748)	0.426654 (0.041761)	-0.510726 (1.732406)
19940901	39	0.001379 (0.006443)	0.067692 (0.095659)	10430459 (58779391)	0.802114 (0.098627)	0.421226 (0.040180)	-0.486812 (1.690042)
19941201	40	0.001118 (0.003737)	0.078829 (0.092919)	10430459 (45036483)	0.821028 (0.098977)	0.424130 (0.040224)	-0.234134 (1.809274)
19950301	41	0.001181 (0.003672)	0.076045 (0.086734)	10430459 (41927466)	0.814168 (0.096514)	0.419416 (0.038852)	-0.300095 (1.703200)
19950601	42	0.001143 (0.003036)	0.077680 (0.082501)	10430459 (36770229)	0.817190 (0.095390)	0.414612 (0.037512)	-0.260900 (1.677350)
19950901	43	0.001250 (0.003250)	0.073242 (0.077150)	10430459 (35480172)	0.809071 (0.094467)	0.411656 (0.036547)	-0.368652 (1.573644)
19951201	44	0.001394 (0.003687)	0.067838 (0.074124)	10430459 (35498266)	0.804638 (0.093578)	0.410193 (0.035873)	-0.499170 (1.511446)

TABLE VIII
NLS ESTIMATES FOR UK DATA, MODELS B AND C.

Data used up to (yyyymmdd)	Num of obs.	Model B			Model C			
		p	q	m	p	q	m	X_1
19880601	14	0.0004265 (0.00889)	0.158829 (0.085044)	15408518 (327126201)	5.98596E-05 (0.00017374)	0.846778 (0.296257)	112543 (79544)	10381.57 (4211.89)
19880901	15	0.0048692 (0.00365)	0.176194 (0.074198)	1393424 (1504552)	2.21799E-07 (1.2925E-05)	0.776775 (1.407389)	5081724 (380630208)	8989.06 (8202.65)
19881201	16	0.0004325 (0.00543)	0.162551 (0.076446)	13934315 (181323638)	1.06121E-05 (7.5152E-05)	0.787388 (0.564850)	215595 (283540)	10401.58 (13341.80)
19890301	17	0.0033675 (0.00217)	0.174415 (0.063021)	1934017 (1908285)	4.32028E-05 (0.00113013)	0.369844 (1.126110)	5064287 (200869818)	1337.52 (17060.86)
19890601	18	0.0017927 (0.00273)	0.155653 (0.054941)	3933918 (7371585)	6.17894E-11 (7.004E-10)	1.470003 (0.731586)	144741 (140333)	10400.00 (16891.48)
19890901	19	0.0026348 (0.00104)	0.172555 (0.044213)	2393508 (1587367)	3.20442E-05 (0.0014588)	0.333251 (1.528106)	5064541 (335837585)	1297.48 (19820.81)
19891201	20	0.0005672 (0.002)	0.150691 (0.044563)	11658787 (45314994)	0.001207672 (0.00604632)	0.270806 (0.418591)	1137444 (2794104)	8601.89 (22485.93)
19900301	21	0.0016002 (0.00105)	0.154727 (0.037927)	4404039 (4193980)	1.56693E-06 (3.8836E-05)	0.462869 (1.294311)	5867958 (265343236)	8366.15 (19322.35)
19900601	22	0.0023239 (0.00069)	0.203997 (0.035554)	1937630 (424569)	1.06683E-06 (9.6714E-06)	0.656271 (0.500393)	434818 (591225)	10415.87 (20704.03)
19900901	23	0.0015498 (0.00078)	0.267022 (0.039976)	1373051 (166576)	3.93961E-12 (4.6731E-11)	1.312660 (0.613664)	202395 (169500)	10400.33 (19225.14)
19901201	24	0.001366 (0.00067)	0.279085 (0.036025)	1319686 (129634)	0.000272695 (0.01029521)	0.352781 (2.255274)	123276 (1582100)	9777.46 (23298.02)
19910301	25	0.0012724 (0.0006)	0.285148 (0.032789)	1297573 (111257)	0.000276523 (0.00963082)	0.365327 (2.145523)	123457 (1464250)	9775.80 (23010.55)
19910601	26	0.000982 (0.00048)	0.305361 (0.032646)	1235802 (99390)	0.000331017 (0.0111591)	0.357614 (2.091902)	123359 (1403017)	9762.53 (22630.45)
19910901	27	0.0008593 (0.00042)	0.315055 (0.031719)	1210463 (92952)	0.000275449 (0.01050311)	0.351208 (2.256528)	116595 (1454911)	9488.69 (22912.68)
19911201	28	0.000868 (0.00041)	0.314314 (0.030131)	1212384 (89331)	0.000375307 (0.01295002)	0.346602 (2.094366)	115059 (1286550)	9490.25 (22534.49)
19920301	29	0.0010437 (0.00048)	0.301458 (0.029266)	1243388 (92578)	0.000300929 (0.00960717)	0.348481 (1.863133)	123678 (1214887)	9149.97 (22140.64)
19920601	30	0.001147 (0.0005)	0.294826 (0.028165)	1259704 (92386)	6.62434E-08 (6.9525E-05)	0.373450 (7.435187)	5000918 (6161152076)	1397.67 (22274.51)
19920901	31	0.001283 (0.00055)	0.286918 (0.027418)	1279343 (93724)	0.000558095 (0.01560875)	0.329848 (1.710596)	124894 (1161393)	8840.82 (22016.22)
19921201	32	0.0017588 (0.00081)	0.264078 (0.029920)	1338925 (114265)	0.000308113 (0.00786619)	0.359156 (1.499078)	129628 (974770)	8850.39 (21437.97)
19930301	33	0.0032257 (0.00162)	0.215820 (0.034266)	1485617 (166317)	1.77635E-07 (6.172E-06)	0.349647 (1.809756)	5060151 (400175839)	1417.46 (21975.32)
19930601	34	0.0063472 (0.00274)	0.143805 (0.033638)	1815448 (254111)	1.31218E-07 (1.5098E-06)	0.367877 (0.963710)	5065839 (114146517)	1927.68 (21703.49)
19930901	35	0.0050857 (0.00436)	0.055552 (0.043399)	4531755 (5266848)	1.96157E-07 (2.7547E-06)	0.341575 (1.218872)	4413352 (141500235)	2501.11 (22121.48)
19931201	36	9.04E-06 (2.7E-05)	0.248316 (0.183915)	4055030 (11443857)	1.67577E-07 (2.1155E-06)	0.338826 (0.971939)	4978008 (117035104)	1574.66 (21831.75)
19940301	37	2.954E-07 (1.4E-06)	0.351525 (0.197850)	3397868 (5423810)	7.93482E-10 (1.1464E-08)	0.496699 (0.645981)	5068613 (42390532)	7432.30 (22342.45)
19940601	38	1.925E-06 (6.3E-06)	0.298027 (0.128173)	3568961 (3272963)	4.86306E-10 (4.7065E-09)	0.553452 (0.290784)	1614273 (1384420)	10401.45 (23112.63)
19940901	39	2.518E-06 (6.9E-06)	0.289957 (0.100005)	3604568 (2137558)	5.7802E-10 (3.7537E-09)	0.547999 (0.187726)	1639534 (939568)	10220.82 (22813.17)
19941201	40	3.755E-07 (1.1E-06)	0.342587 (0.092906)	3438788 (1256691)	1.59784E-09 (7.9554E-09)	0.516784 (0.141327)	1772347 (899314)	9853.20 (22592.95)
19950301	41	2.868E-07 (7E-07)	0.349660 (0.074262)	3438788 (802590)	7.80878E-09 (1.0513E-07)	0.401750 (0.446951)	5065244 (19274920)	7521.15 (28733.31)
19950601	42	2.595E-07 (5.4E-07)	0.351202 (0.061076)	3479748 (576599)	3.85975E-07 (1.4583E-06)	0.339221 (0.106278)	5056014 (2728942)	1337.39 (33609.13)
19950901	43	3.419E-07 (6E-07)	0.344722 (0.050412)	3479748 (432717)	3.70184E-07 (1.3079E-06)	0.336781 (0.097067)	4938866 (2575721)	2536.26 (33341.09)
19951201	44	2.694E-07 (4.2E-07)	0.352972 (0.044052)	3397828 (335874)	2.92355E-08 (1.314E-07)	0.408157 (0.120820)	3023079 (1833798)	10387.03 (33284.26)

TABLE IX
NLS ESTIMATES FOR US DATA, MODEL A.

Data used up to (yyymmdd)	Num of obs.	Model A					
		p	q	m	Ψ	σ	X_1
19900901	17	0.004654 (0.002483)	0.138121 (0.024778)	16350008 (10162649)	0.148518 (0.244562)	0.114885 (0.004527)	0.022913 (1.047152)
19901201	18	0.005566 (0.001563)	0.145576 (0.021498)	13317066 (4764180)	0.177099 (0.235931)	0.112893 (0.004248)	0.132083 (0.868992)
19910301	19	0.005294 (0.001238)	0.142955 (0.018353)	14141084 (4250052)	0.165907 (0.226238)	0.110099 (0.003933)	0.094700 (0.882264)
19910601	20	0.004834 (0.001109)	0.137610 (0.016576)	15862230 (4642415)	0.170112 (0.222516)	0.108344 (0.003712)	-0.013013 (0.828247)
19910901	21	0.004730 (0.000866)	0.137164 (0.014538)	16199985 (3863588)	0.171941 (0.215154)	0.105733 (0.003450)	-0.015450 (0.788891)
19911201	22	0.004668 (0.000689)	0.136435 (0.012884)	16469514 (3226166)	0.172984 (0.210105)	0.103339 (0.003220)	-0.029995 (0.757323)
19920301	23	0.004443 (0.000597)	0.133572 (0.011719)	17531357 (3132557)	0.176631 (0.206704)	0.101725 (0.003051)	-0.088830 (0.727045)
19920601	24	0.004154 (0.000550)	0.129670 (0.011047)	19113992 (3362375)	0.195156 (0.203366)	0.100796 (0.002933)	-0.159915 (0.660491)
19920901	25	0.003102 (0.000876)	0.116704 (0.013811)	27354018 (9617805)	0.281871 (0.218362)	0.110014 (0.003423)	-0.304454 (0.529973)
19921201	26	0.001567 (0.001747)	0.104066 (0.020817)	57562220 (71350582)	0.472164 (0.213838)	0.122304 (0.004149)	-0.288384 (0.402231)
19930301	27	0.001606 (0.001250)	0.102981 (0.016233)	56764932 (49623467)	0.423482 (0.173113)	0.120907 (0.003979)	-0.344028 (0.420735)
19930601	28	0.001753 (0.000995)	0.103421 (0.014458)	52101708 (34041349)	0.413418 (0.170039)	0.119725 (0.003831)	-0.356845 (0.420018)
19930901	29	0.001377 (0.000937)	0.101223 (0.013358)	66682564 (50954471)	0.392394 (0.172221)	0.121404 (0.003871)	-0.384327 (0.436919)
19931201	30	0.001480 (0.000747)	0.106653 (0.013900)	58894164 (34880183)	0.451603 (0.177944)	0.128459 (0.004261)	-0.230960 (0.414245)
19940301	31	0.001488 (0.000626)	0.106145 (0.012457)	58894164 (29376381)	0.430623 (0.164286)	0.126568 (0.004069)	-0.254346 (0.420153)
19940601	32	0.001436 (0.000533)	0.109035 (0.012990)	58894164 (26754631)	0.465698 (0.172144)	0.132546 (0.004392)	-0.165271 (0.413869)
19940901	33	0.001395 (0.000494)	0.112847 (0.016414)	57826753 (26860987)	0.573443 (0.174273)	0.137301 (0.004641)	-0.057808 (0.400860)
19941201	34	0.001268 (0.000459)	0.118972 (0.026293)	57750759 (31418107)	0.713725 (0.178904)	0.144540 (0.005067)	0.079587 (0.504167)
19950301	35	0.001317 (0.000421)	0.116802 (0.021615)	57750759 (27571615)	0.661925 (0.169233)	0.143425 (0.004917)	0.031929 (0.447054)
19950601	36	0.001337 (0.000375)	0.115906 (0.019160)	57750829 (23873206)	0.645651 (0.166540)	0.141929 (0.004748)	0.011102 (0.422913)
19950901	37	0.001269 (0.000307)	0.117389 (0.018308)	59061469 (20945033)	0.653107 (0.162901)	0.144498 (0.004854)	0.054196 (0.421591)
19951201	38	0.001264 (0.000191)	0.117789 (0.010052)	59061497 (12496905)	0.286755 (0.191814)	0.170498 (0.006669)	0.229984 (0.701004)
19960301	39	0.001149 (0.000249)	0.113337 (0.013253)	68236554 (20484951)	0.566184 (0.165322)	0.147907 (0.004954)	-0.010645 (0.403061)
19960601	40	0.001193 (0.000247)	0.111233 (0.012183)	68236527 (18342323)	0.557217 (0.164835)	0.155185 (0.005385)	-0.073607 (0.411743)
19960901	41	0.001262 (0.000135)	0.118745 (0.007842)	58406183 (7438114)	0.286326 (0.157078)	0.171388 (0.006488)	0.235656 (0.694015)
19961201	42	0.001044 (0.000163)	0.108906 (0.008509)	78722366 (14996101)	0.450372 (0.144716)	0.161738 (0.005708)	-0.107225 (0.457830)
19970301	43	0.001185 (0.000146)	0.111603 (0.007648)	68236524 (9601595)	0.294602 (0.167759)	0.190562 (0.007832)	-0.111392 (0.757863)
19970601	44	0.001186 (0.000135)	0.111600 (0.007199)	68236557 (8043154)	0.313333 (0.145425)	0.188281 (0.007558)	-0.117672 (0.700991)
19970901	45	0.001201 (0.000138)	0.112657 (0.007307)	66729546 (7034925)	0.353681 (0.143309)	0.188492 (0.007490)	-0.078532 (0.628221)
19971201	46	0.001267 (0.000131)	0.118063 (0.006523)	58406131 (4453665)	0.273592 (0.146018)	0.195876 (0.008000)	0.230748 (0.813654)
19980301	47	0.001252 (0.000153)	0.118102 (0.007410)	60262195 (4826438)	0.340174 (0.156448)	0.209888 (0.009087)	0.081920 (0.714924)

TABLE X
NLS ESTIMATES FOR US DATA, MODELS B AND C.

Data used up to (yyymmdd)	Num of obs.	Model B			Model C			
		p	q	m	p	q	m	X_1
19900901	17	0.005956468 (0.0006666)	0.162048 (0.016655)	11254428 (2044071)	0.000896 (0.001066)	0.401104 (0.104807)	3128212 (1411945)	83019.24 (24085.59)
19901201	18	0.006205648 (0.0004523)	0.167509 (0.013893)	10487452 (1227842)	0.0032519 (0.002953)	0.256994 (0.092013)	5790965 (3242125)	62957.04 (40211.32)
19910301	19	0.005806906 (0.0004181)	0.156689 (0.012603)	11936524 (1402806)	0.0038801 (0.003091)	0.231749 (0.078719)	6681941 (3559442)	54665.13 (45614.92)
19910601	20	0.005091663 (0.0004703)	0.141168 (0.012486)	14917966 (2260481)	0.0057511 (0.00351)	0.109929 (0.123849)	21534850 (40105096)	-50142.58 (275585.42)
19910901	21	0.004861326 (0.0003914)	0.136777 (0.010890)	16044758 (2134059)	0.0015059 (0.023854)	0.062522 (0.229705)	152765944 (3050468530)	-157330.79 (1013591.53)
19911201	22	0.00474627 (0.0003169)	0.134395 (0.009467)	16685353 (1843696)	0.0051961 (0.002994)	0.106034 (0.098107)	24390820 (36392046)	-52943.31 (244905.52)
19920301	23	0.004463927 (0.0002914)	0.128272 (0.008659)	18482746 (2000020)	0.0052314 (0.003069)	0.112780 (0.079253)	21864768 (23367819)	-40091.91 (188349.00)
19920601	24	0.004089019 (0.0002917)	0.120722 (0.008241)	21255548 (2502254)	0.0045227 (0.003012)	0.076662 (0.099435)	45209873 (105800918)	-132490.29 (446456.92)
19920901	25	0.002197143 (0.0009282)	0.097644 (0.012963)	45773718 (23825143)	0.0024802 (0.011313)	0.052755 (0.137645)	126783652 (966778666)	-243440.87 (1041625.93)
19921201	26	0.000860887 (0.0015067)	0.097953 (0.020859)	105691571 (204466216)	5.05E-07 (6.12E-06)	0.400685 (0.408249)	45330886 (922036306)	87557.81 (35825.93)
19930301	27	0.000817305 (0.0012035)	0.097312 (0.018171)	112037309 (184081418)	3.303E-09 (1.5E-08)	0.595182 (0.337278)	45357289 (521392581)	87432.53 (33748.14)
19930601	28	0.001562531 (0.0006147)	0.107557 (0.016786)	53347039 (29537680)	9.6E-06 (4.31E-05)	0.387842 (0.202835)	7187838 (6127037)	87605.80 (75905.54)
19930901	29	0.001026127 (0.000692)	0.100596 (0.015275)	85671158 (71472213)	1.67E-05 (5.16E-05)	0.367376 (0.134458)	8337999 (5327242)	89145.13 (75632.18)
19931201	30	0.000665058 (0.0006373)	0.103763 (0.016624)	119297287 (137532090)	0.0001931 (0.000812)	0.152099 (0.244100)	80374001 (704573906)	75577.25 (137158.95)
19940301	31	0.000672418 (0.0005525)	0.100857 (0.014735)	125059000 (124990991)	3.43E-07 (1.58E-06)	0.374298 (0.444359)	45329198 (449176196)	87337.30 (93100.09)
19940601	32	0.000574278 (0.000408)	0.104869 (0.014103)	132941229 (118951945)	2.049E-06 (1E-05)	0.395778 (0.185940)	9797991 (7708194)	87191.97 (113094.72)
19940901	33	0.000504416 (0.0002672)	0.112617 (0.014447)	126514357 (92878897)	6.07E-05 (0.000238)	0.208069 (0.234830)	45343744 (145700095)	89989.97 (126996.53)
19941201	34	0.000397152 (0.0001211)	0.160900 (0.023117)	54735983 (15952249)	0.0001248 (0.000427)	0.186920 (0.181077)	45328174 (98063436)	89999.76 (131570.28)
19950301	35	0.000404426 (0.0001727)	0.117236 (0.015265)	138622595 (91989621)	7.885E-05 (0.000375)	0.185136 (0.277788)	45328171 (191023859)	89999.76 (146283.18)
19950601	36	0.000582137 (0.0001652)	0.113167 (0.015929)	112407830 (56955493)	2.706E-07 (1.09E-06)	0.415940 (0.132510)	17347784 (9683163)	79572.90 (166525.80)
19950901	37	0.000650801 (0.0001067)	0.121608 (0.014203)	86267157 (23982078)	0.0001736 (0.000376)	0.202839 (0.080444)	45306833 (31034214)	90009.37 (181584.33)
19951201	38	0.000511829 (0.0001442)	0.106848 (0.013267)	143870004 (69777755)	8.667E-05 (0.000281)	0.213602 (0.114000)	36729117 (32968317)	83326.26 (194382.19)
19960301	39	0.000645663 (0.0001298)	0.128572 (0.013400)	75707422 (13050009)	2.631E-05 (0.000175)	0.208388 (0.259586)	45337098 (120363252)	89847.33 (207731.83)
19960601	40	0.000460075 (0.0001483)	0.156186 (0.014858)	55445277 (5586178)	2.841E-05 (9.35E-05)	0.250114 (0.105969)	34850389 (28430621)	85752.38 (245435.96)
19960901	41	0.000444607 (0.0001385)	0.158212 (0.013478)	54550036 (4491939)	0.0001347 (0.000353)	0.204714 (0.088668)	45328158 (39083434)	89999.77 (254552.42)
19961201	42	0.000607055 (0.0001817)	0.137909 (0.014074)	65221662 (7197963)	3.767E-05 (0.00015)	0.243405 (0.128473)	29443774 (32052510)	85911.17 (269179.62)
19970301	43	0.000396506 (0.0001474)	0.160856 (0.014645)	54735940 (4320485)	1.52E-05 (0.000241)	0.186175 (0.679468)	46013021 (522577054)	89502.30 (313305.94)
19970601	44	0.000398067 (0.0001406)	0.160862 (0.013455)	54735940 (3793505)	7.557E-05 (0.000311)	0.214036 (0.131666)	45328160 (57461955)	89999.72 (399096.39)
19970901	45	0.000401301 (0.0001349)	0.160877 (0.012443)	54735940 (3386089)	0.0001153 (0.000561)	0.202385 (0.158558)	39158308 (62839448)	81788.05 (404521.38)
19971201	46	0.000392095 (0.0001295)	0.160834 (0.011899)	54735940 (3173968)	0.0001115 (0.000528)	0.203147 (0.153542)	39859364 (60857146)	81133.45 (399972.45)
19980301	47	0.000402879 (0.0001296)	0.160919 (0.011363)	54735980 (2967884)	1.73E-05 (0.000127)	0.270601 (0.231692)	20068128 (31777951)	86912.77 (399783.25)

TABLE XI
NLS ESTIMATES FOR AUSTRALIAN DATA, MODEL A.

Data used up to (yyyymmdd)	Num of obs.	Model A						X_1
		p	q	m	ψ	σ		
19890601	10	0.001093 (0.013064)	0.184698 (0.081625)	2865479 (34884964)	0.022619 (0.135431)	0.121853 (0.006640)	-53.4704922 (314.4152327)	
19890901	11	0.007184 (0.007302)	0.198636 (0.076918)	440772 (543945)	0.000096 (0.173572)	0.158925 (0.010770)	-12860.19188 (23358212.74)	
19891201	12	0.001178 (0.012724)	0.173747 (0.092741)	2779716 (30871804)	0.016981 (0.199762)	0.183448 (0.013739)	-73.16356156 (849.2858729)	
19900301	13	0.001916 (0.007694)	0.170091 (0.063897)	1754459 (7414453)	0.000178 (0.166440)	0.183346 (0.013185)	-7122.934901 (6664382.82)	
19900601	14	0.003773 (0.004511)	0.177987 (0.052357)	882756 (1220137)	0.000163 (0.163488)	0.180297 (0.012286)	-7738.427317 (7750523.269)	
19900901	15	0.004687 (0.002455)	0.188256 (0.041002)	684958 (462035)	0.000155 (0.154837)	0.171935 (0.010794)	-7996.20309 (8010806.67)	
19901201	16	0.005405 (0.001421)	0.194574 (0.032940)	584047 (218160)	0.000241 (0.148093)	0.168752 (0.010068)	-5069.652846 (3108701.939)	
19910301	17	0.006595 (0.000846)	0.217466 (0.026071)	440010 (68214)	0.000120 (0.147481)	0.169061 (0.009803)	-9687.577233 (11918633.64)	
19910601	18	0.006169 (0.000752)	0.206266 (0.025704)	491107 (80987)	0.000366 (0.150497)	0.172377 (0.009905)	-3264.519259 (1343331.428)	
19910901	19	0.005260 (0.000834)	0.185330 (0.027139)	626652 (151683)	0.000315 (0.158742)	0.191487 (0.011896)	-3990.444767 (2008653.099)	
19911201	20	0.004931 (0.000738)	0.178336 (0.022957)	688909 (146027)	0.000759 (0.142241)	0.189290 (0.011331)	-1689.086069 (316488.4046)	
19920301	21	0.004762 (0.000636)	0.174478 (0.020238)	726025 (131491)	0.001734 (0.136086)	0.185633 (0.010634)	-747.3593412 (58618.1322)	
19920601	22	0.004458 (0.000586)	0.167124 (0.018880)	803972 (144984)	0.008889 (0.133414)	0.184653 (0.010281)	-149.0555204 (2230.397755)	
19920901	23	0.003874 (0.000664)	0.152875 (0.019672)	1001176 (245872)	0.051465 (0.133930)	0.189842 (0.010628)	-27.017648 (69.072597)	
19921201	24	0.002589 (0.001224)	0.130098 (0.023439)	1712644 (1027406)	0.132842 (0.136243)	0.201890 (0.011766)	-11.316830 (11.066267)	
19930301	25	0.002542 (0.000951)	0.129383 (0.019935)	1752481 (836169)	0.135723 (0.124454)	0.197824 (0.011069)	-11.103841 (9.798371)	
19930601	26	0.001626 (0.001254)	0.117904 (0.020509)	2925092 (2615021)	0.159208 (0.124636)	0.200219 (0.011118)	-9.813652 (7.328055)	
19930901	27	0.001080 (0.001296)	0.112948 (0.019608)	4527172 (5994304)	0.183035 (0.120979)	0.201324 (0.011031)	-8.635890 (5.447848)	
19931201	28	0.000621 (0.001249)	0.111836 (0.018940)	7823043 (16723546)	0.200905 (0.122973)	0.211925 (0.012003)	-7.854464 (4.633432)	
19940301	29	0.000524 (0.001045)	0.111049 (0.016774)	9314659 (19625264)	0.207143 (0.114467)	0.207844 (0.011345)	-7.636467 (4.123995)	
19940601	30	0.000553 (0.000821)	0.116113 (0.016374)	8249902 (13167273)	0.261813 (0.116538)	0.212973 (0.011711)	-5.190946 (2.348245)	
19940901	31	0.000469 (0.000687)	0.115853 (0.014558)	9853359 (15408717)	0.222585 (0.117056)	0.216080 (0.011859)	-6.887062 (3.618319)	
19941201	32	0.000352 (0.000601)	0.115513 (0.013536)	13092485 (23582747)	0.236670 (0.113565)	0.213583 (0.011404)	-6.467758 (3.138671)	
19950301	33	0.000358 (0.000494)	0.116944 (0.012628)	12683928 (18618150)	0.246751 (0.112663)	0.212776 (0.011146)	-6.148578 (2.867950)	
19950601	34	0.000328 (0.000418)	0.116934 (0.011624)	13837626 (18802040)	0.247061 (0.110706)	0.209219 (0.010616)	-6.133272 (2.811658)	
19950901	35	0.000350 (0.000351)	0.117086 (0.010832)	12977964 (14055991)	0.248660 (0.109492)	0.206987 (0.010242)	-6.093977 (2.742943)	
19951201	36	0.000255 (0.000296)	0.118377 (0.010216)	17271803 (21442655)	0.222587 (0.114109)	0.216451 (0.011043)	-6.682130 (3.486147)	
19960301	37	0.000300 (0.000324)	0.114880 (0.010885)	15304753 (17749736)	0.143046 (0.137562)	0.264569 (0.016274)	-10.626210 (10.204138)	
19960601	38	0.001429 (0.000619)	0.141316 (0.028591)	2852908 (890774)	0.545639 (0.184400)	0.394176 (0.035645)	-2.612670 (1.317629)	
19960901	39	0.001429 (0.000567)	0.141310 (0.025943)	2852908 (656234)	0.545640 (0.133108)	0.389090 (0.034283)	-2.612668 (1.121837)	
19961201	40	0.001427 (0.000553)	0.141910 (0.024581)	2852948 (565612)	0.545588 (0.123907)	0.384529 (0.033063)	-2.612714 (1.070665)	

TABLE XII
NLS ESTIMATES FOR AUSTRALIAN DATA, MODELS B AND C.

Data used up to (yyymmdd)	Num of obs.	Model B			Model C			X_1
		p	q	m	p	q	m	
19890601	10	0.007437881 (0.001869363)	0.287187 (0.055584)	275853 (114098)	0.0064115 (0.229455)	0.049426 (1.399425)	5987721 (384652329)	-38943.75 (1102423.54)
19890901	11	0.008750358 (0.001900459)	0.379149 (0.050779)	163730 (19464)	0.0024363 (0.249242)	0.051499 (1.282976)	12711618 (1582089407)	-31399.56 (693589.59)
19891201	12	0.000829277 (0.009697562)	0.190918 (0.102061)	3395241 (41102397)	0.001842 (0.002682)	0.609528 (0.196541)	97355 (61125)	791.17 (2110.42)
19900301	13	0.005846989 (0.001981915)	0.240622 (0.076199)	418473 (245726)	6.017E-08 (5.84E-06)	1.028401 (2.454641)	1617920 (198029482)	1695.25 (3723.74)
19900601	14	0.006248546 (0.001539829)	0.256317 (0.058782)	363940 (106547)	8.699E-06 (5.96E-05)	0.931855 (0.625434)	76021 (90221)	1121.28 (5361.04)
19900901	15	0.005789591 (0.001230377)	0.227918 (0.049283)	454768 (139042)	0.0065074 (0.016766)	0.264043 (0.356414)	361864 (742489)	-1920.72 (13620.61)
19901201	16	0.00590308 (0.001231467)	0.237541 (0.039786)	422803 (77223)	0.0057775 (0.084245)	0.061871 (0.905047)	4615062 (144268309)	-27068.10 (466329.21)
19910301	17	0.005775589 (0.001343739)	0.258542 (0.033937)	377832 (41487)	0.0073034 (0.015276)	0.219490 (0.265144)	492841 (912430)	-3322.53 (16499.49)
19910601	18	0.006183799 (0.001305613)	0.220099 (0.033123)	457126 (62484)	0.0037669 (0.008178)	0.304970 (0.218563)	333522 (443252)	-574.39 (7413.54)
19910901	19	0.004977685 (0.001356417)	0.153439 (0.039866)	845510 (383546)	3.141E-07 (9.22E-05)	0.552481 (3.993649)	1617405 (572136123)	2124.69 (5709.12)
19911201	20	0.004670384 (0.001123987)	0.147666 (0.034192)	929996 (373952)	2.947E-07 (1.19E-05)	0.570768 (1.733979)	1616290 (108872375)	2120.95 (5573.17)
19920301	21	0.004746648 (0.00085215)	0.149304 (0.029070)	906320 (262479)	0.0056405 (0.057619)	0.039215 (0.625713)	8505213 (245993249)	-48479.73 (929837.40)
19920601	22	0.004388608 (0.000759878)	0.140770 (0.025689)	1033795 (295736)	0.0083977 (0.011422)	0.077130 (0.298753)	2478121 (14538838)	-21138.65 (139527.17)
19920901	23	0.00304568 (0.001204841)	0.117942 (0.025411)	1720953 (979015)	0.003111 (0.127219)	0.023173 (0.835528)	26276547 (2047295952)	-82275.49 (3066774.58)
19921201	24	0.000358414 (0.002671427)	0.099251 (0.029429)	15505647 (119234594)	5.491E-08 (2.52E-06)	0.539831 (1.916063)	1618015 (133077899)	2115.61 (5785.15)
19930301	25	0.001156281 (0.0016176)	0.104984 (0.025258)	4690146 (7572293)	4.577E-08 (3.12E-07)	0.555057 (0.929760)	1622935 (31966428)	2055.78 (5688.04)
19930601	26	0.000330132 (0.001647482)	0.101660 (0.023746)	16099800 (83953208)	6.564E-11 (4.68E-10)	0.966816 (0.311244)	154560 (93802)	1236.81 (6080.72)
19930901	27	0.000276893 (0.001125506)	0.110256 (0.022945)	16407388 (70751124)	2.655E-08 (2.79E-07)	0.529364 (1.309306)	1618035 (49325298)	2095.61 (7484.58)
19931201	28	0.000135152 (0.00034289)	0.161978 (0.038610)	10591824 (32944141)	5.245E-06 (3.23E-05)	0.338373 (0.448268)	1756160 (9806510)	1108.93 (7294.32)
19940301	29	0.000194248 (0.000545527)	0.130747 (0.026754)	15492135 (49216477)	1.899E-08 (1.56E-07)	0.543845 (0.432086)	1618036 (5615607)	2095.61 (7514.67)
19940601	30	0.000170994 (0.000362913)	0.134749 (0.023519)	16023661 (39593268)	1.807E-12 (8.73E-12)	0.953598 (0.179381)	397198 (137251)	1123.37 (8633.30)
19940901	31	0.000124179 (0.000148377)	0.157573 (0.026157)	12681866 (20707833)	3.19E-05 (0.000148)	0.270232 (0.221771)	2264999 (3706082)	1033.28 (11178.82)
19941201	32	0.000115182 (3.1734E-05)	0.189053 (0.026630)	6365169 (3705000)	2.283E-09 (2.93E-08)	0.567897 (0.546074)	1618122 (5647344)	1381.15 (11900.84)
19950301	33	9.38145E-05 (3.6526E-05)	0.213365 (0.024820)	4405486 (1176808)	1.191E-07 (6.03E-07)	0.465277 (0.182530)	1373911 (876164)	1437.52 (12464.09)
19950601	34	0.000110197 (3.88989E-05)	0.208026 (0.020183)	4410514 (825257)	3.165E-06 (1.1E-05)	0.344751 (0.126273)	2099365 (1254512)	1107.80 (12247.85)
19950901	35	0.000187902 (5.29549E-05)	0.185464 (0.016595)	4883589 (802424)	2.526E-06 (7.1E-06)	0.352886 (0.097722)	2025063 (950924)	1116.30 (12066.99)
19951201	36	0.000101624 (9.45593E-05)	0.142953 (0.023352)	21453195 (29757712)	1.997E-06 (4.73E-06)	0.361500 (0.079848)	1962188 (807886)	1114.48 (11898.27)
19960301	37	8.10099E-05 (6.8656E-05)	0.225123 (0.034953)	3922129 (658449)	7.832E-10 (1.43E-08)	0.538389 (0.586086)	1606691 (3579926)	2097.19 (26582.11)
19960601	38	2.6068E-07 (4.62035E-07)	0.430117 (0.058097)	2178950 (228662)	1.096E-09 (1.05E-08)	0.569086 (0.290301)	1618003 (1664930)	1765.18 (49551.73)
19960901	39	2.42829E-07 (4.01765E-07)	0.433221 (0.053786)	2176470 (211926)	9.115E-12 (7.95E-11)	0.703641 (0.257956)	1595033 (1168902)	1134.29 (48074.77)
19961201	40	3.50948E-06 (4.1123E-06)	0.348466 (0.039149)	2565510 (225391)	3.347E-09 (3.11E-08)	0.539549 (0.283665)	1617985 (1629749)	2124.89 (49620.20)

TABLE XIII
NLS ESTIMATES FOR KOREAN DATA, MODEL A.

Data used up to (yyyymmdd)	Num of obs.	Model A					
		p	q	m	Ψ	σ	X_1
19890301	21	0.000035 (0.006984)	0.151887 (0.165570)	3886295 (776021468)	0.635799 (0.247207)	0.266215 (0.021871)	2.888273 (1.386678)
19890601	22	0.000035 (0.006248)	0.149750 (0.150518)	4074213 (740590526)	0.631475 (0.238952)	0.260126 (0.020402)	2.857124 (1.261744)
19890901	23	0.000037 (0.005040)	0.146655 (0.121947)	3965151 (544361975)	0.621817 (0.215363)	0.254733 (0.019135)	2.850168 (1.057268)
19891201	24	0.000039 (0.003952)	0.142521 (0.094279)	3965151 (404932985)	0.612389 (0.186854)	0.250152 (0.018064)	2.814481 (0.897882)
19900301	25	0.000033 (0.002152)	0.156009 (0.081141)	3965147 (262354112)	0.634598 (0.165297)	0.256140 (0.018557)	2.962896 (0.898408)
19900601	26	0.000026 (0.001251)	0.172787 (0.085442)	3886285 (189443090)	0.678234 (0.159990)	0.259283 (0.018646)	3.115575 (1.032581)
19900901	27	0.000028 (0.001165)	0.168037 (0.075774)	3886285 (163714062)	0.664540 (0.156730)	0.254984 (0.017695)	3.072678 (0.917633)
19901201	28	0.000034 (0.001294)	0.154456 (0.061108)	3954452 (152789062)	0.626557 (0.154542)	0.260687 (0.018162)	2.952238 (0.789908)
19910301	29	0.000035 (0.000941)	0.153381 (0.048403)	3945239 (109202147)	0.625814 (0.132636)	0.256309 (0.017252)	2.935222 (0.742821)
19910601	30	0.000027 (0.000485)	0.172607 (0.051482)	3886297 (71671223)	0.641021 (0.140709)	0.298935 (0.023073)	3.239600 (0.849220)
19910901	31	0.000031 (0.000512)	0.164415 (0.044739)	3886297 (65644595)	0.604069 (0.140770)	0.297148 (0.022427)	3.220223 (0.809446)
19911201	32	0.000034 (0.000511)	0.156650 (0.038123)	3965150 (60444914)	0.577473 (0.139002)	0.300198 (0.022530)	3.174048 (0.824794)
19920301	33	0.000034 (0.000346)	0.160105 (0.032868)	3886297 (41374221)	0.577949 (0.126591)	0.298542 (0.021942)	3.245568 (0.814185)
19920601	34	0.000032 (0.000249)	0.164211 (0.031129)	3886297 (32021807)	0.589181 (0.122956)	0.297342 (0.021443)	3.284739 (0.792540)
19920901	35	0.000118 (0.000237)	0.159695 (0.029427)	1152622 (2671169)	0.553627 (0.133597)	0.322263 (0.024826)	3.318161 (0.901808)
19921201	36	0.000038 (0.000217)	0.153966 (0.023256)	3886297 (23363468)	0.522943 (0.119981)	0.320128 (0.024155)	3.381206 (0.955981)
19930301	37	0.000078 (0.000161)	0.156140 (0.021886)	1829267 (4157044)	0.527860 (0.117238)	0.317344 (0.023414)	3.385256 (0.934808)
19930601	38	0.000034 (0.000147)	0.153289 (0.019538)	4366124 (19795387)	0.517787 (0.114092)	0.315292 (0.022806)	3.397143 (0.952192)
19930901	39	0.000031 (0.000116)	0.154786 (0.018572)	4687734 (18622827)	0.522571 (0.112860)	0.312581 (0.022126)	3.410934 (0.929180)
19931201	40	0.000031 (0.000097)	0.154676 (0.017497)	4700449 (15701180)	0.522069 (0.111433)	0.308651 (0.021302)	3.410637 (0.913748)
19940301	41	0.000034 (0.000060)	0.160366 (0.015204)	3886337 (7380203)	0.463264 (0.112055)	0.311637 (0.021450)	3.651776 (1.031698)
19940601	42	0.000036 (0.000054)	0.158934 (0.014615)	3847689 (6276841)	0.461164 (0.112976)	0.314062 (0.021524)	3.663328 (1.053796)
19940901	43	0.000035 (0.000042)	0.160788 (0.014124)	3847089 (5187332)	0.459166 (0.111963)	0.311731 (0.020958)	3.664407 (1.034547)
19941201	44	0.000035 (0.000035)	0.160802 (0.013493)	3847089 (4388660)	0.458096 (0.111492)	0.311035 (0.020626)	3.667464 (1.035706)
19950301	45	0.000035 (0.000030)	0.160796 (0.012832)	3847089 (3742317)	0.458459 (0.110082)	0.307693 (0.019959)	3.666545 (1.022306)
19950601	46	0.000036 (0.000025)	0.160842 (0.012320)	3827535 (3123455)	0.457330 (0.110270)	0.313192 (0.020453)	3.670467 (1.045520)
19950901	47	0.000036 (0.000021)	0.160843 (0.011771)	3827535 (2685863)	0.457330 (0.109031)	0.309936 (0.019816)	3.670467 (1.033573)
19951201	48	0.000036 (0.000018)	0.160843 (0.011093)	3827535 (2261138)	0.457330 (0.106719)	0.306699 (0.019201)	3.670467 (1.022655)
19960301	49	0.000036 (0.000016)	0.160844 (0.011261)	3827515 (2027929)	0.457327 (0.112226)	0.328401 (0.021788)	3.670468 (1.095881)
19960601	50	0.000035 (0.000013)	0.162538 (0.011064)	3828135 (1746576)	0.457720 (0.111094)	0.325773 (0.021226)	3.670480 (1.070093)
19960901	51	0.000035 (0.000012)	0.162102 (0.011028)	3830055 (1605830)	0.457948 (0.114683)	0.337707 (0.022584)	3.670621 (1.111304)

TABLE XIV
NLS ESTIMATES FOR KOREAN DATA, MODELS B AND C.

Data used up to (yyyymmdd)	Num of obs.	Model B			Model C			
		p	q	m	p	q	m	X_1
19890301	21	2.0549E-05 (0.001381635)	0.189185 (0.075227)	3657237 (248857482)	3.304E-05 (0.018728)	0.147933 (0.704037)	3476839 (2005024307)	867.01 (1429.35)
19890601	22	2.14766E-05 (0.000965418)	0.186575 (0.060111)	3657250 (166927946)	1.845E-07 (5.85E-05)	0.415935 (0.676150)	3583057 (1178535658)	749.51 (462.86)
19890901	23	2.99246E-05 (0.001016656)	0.167852 (0.049283)	3657174 (126320906)	1.312E-05 (7.49E-05)	0.452188 (0.328734)	37784 (46138)	999.24 (510.06)
19891201	24	4.16457E-05 (0.001189777)	0.149282 (0.043381)	3657006 (106276491)	2.507E-07 (1.26E-06)	0.660143 (0.248596)	21480 (14213)	1000.02 (495.36)
19900301	25	2.77919E-05 (0.000584904)	0.170711 (0.043096)	3657140 (79067374)	1.843E-05 (6E-05)	0.436601 (0.165081)	38297 (25929)	998.91 (491.98)
19900601	26	9.24654E-06 (0.000161995)	0.222449 (0.056875)	3656725 (67478180)	3.896E-07 (5.19E-05)	0.340492 (0.506221)	3625422 (520667797)	672.04 (675.63)
19900901	27	1.28595E-05 (0.000148947)	0.207891 (0.044270)	3657197 (45072765)	1.406E-07 (6.83E-06)	0.386047 (0.321826)	3757603 (209020985)	640.71 (681.14)
19901201	28	3.55901E-05 (0.000489188)	0.162603 (0.048994)	3657029 (53030135)	2.934E-10 (1.23E-09)	0.814510 (0.170082)	49053 (16950)	1011.14 (756.79)
19910301	29	5.33562E-05 (0.000641105)	0.144207 (0.043946)	3656919 (46365236)	1.879E-09 (5.5E-09)	0.743800 (0.117519)	58036 (18451)	939.89 (815.07)
19910601	30	4.20265E-06 (5.24451E-05)	0.245569 (0.099606)	3676284 (53444130)	1.524E-11 (9.5E-11)	0.942929 (0.248199)	28467 (13622)	990.35 (1006.06)
19910901	31	1.11443E-05 (9.28325E-05)	0.209556 (0.073605)	3657140 (35911659)	2.898E-08 (4.39E-07)	0.424353 (0.460156)	3775003 (100203880)	459.60 (2798.67)
19911201	32	4.12639E-05 (0.000308602)	0.162587 (0.064785)	3422489 (29740030)	9.81E-06 (9.02E-05)	0.316762 (0.378311)	321308 (657447)	987.35 (4069.14)
19920301	33	9.12337E-05 (7.13183E-05)	0.211140 (0.052338)	544575 (334599)	1.219E-11 (8.22E-11)	0.815417 (0.232608)	124860 (72259)	1019.77 (3637.54)
19920601	34	2.69363E-05 (0.000100555)	0.175276 (0.049530)	3657141 (17450087)	2.482E-06 (0.000171)	0.232633 (0.911854)	3863637 (359545932)	47.44 (4488.79)
19920901	35	2.42227E-05 (2.97413E-05)	0.276532 (0.049924)	370962 (74239)	3.264E-08 (9.72E-07)	0.366513 (0.777324)	3775773 (194434873)	815.00 (4483.68)
19921201	36	4.38577E-05 (4.47003E-05)	0.246315 (0.042252)	446557 (90375)	3.353E-10 (3.29E-09)	0.634964 (0.310211)	170818 (164650)	995.22 (5757.61)
19930301	37	6.36183E-05 (5.51165E-05)	0.228832 (0.035692)	492907 (88700)	0.0002535 (0.001852)	0.163903 (0.312418)	827657 (2678179)	770.54 (6774.54)
19930601	38	0.000132884 (7.09681E-05)	0.145798 (0.035990)	1770386 (2019707)	0.0001285 (0.00085)	0.201038 (0.237367)	587762 (1163332)	913.08 (6270.00)
19930901	39	4.00485E-05 (5.66602E-05)	0.155730 (0.038883)	3714632 (8554403)	3.976E-07 (7.05E-06)	0.268013 (0.742106)	3780297 (147872502)	309.97 (6470.72)
19931201	40	3.27206E-05 (3.11142E-05)	0.160657 (0.034094)	3830674 (6673564)	6.591E-08 (4.73E-07)	0.316904 (0.550435)	3806643 (76721576)	503.73 (6388.90)
19940301	41	1.65483E-05 (1.1647E-05)	0.176850 (0.034198)	4265447 (6439514)	4.332E-06 (2.36E-05)	0.211067 (0.379691)	3439830 (37353468)	921.65 (6559.30)
19940601	42	1.68244E-06 (1.26962E-06)	0.241693 (0.047465)	3657450 (4124634)	1.496E-08 (1.06E-07)	0.346818 (0.418097)	3780197 (40315088)	892.43 (6650.15)
19940901	43	2.18079E-06 (1.55622E-06)	0.235129 (0.037084)	3659690 (2642104)	2.513E-06 (1.24E-05)	0.235235 (0.182728)	3775705 (8967718)	3.33 (7887.73)
19941201	44	1.59606E-06 (1.14254E-06)	0.242720 (0.030538)	3657140 (1711898)	7.512E-09 (4.06E-08)	0.407011 (0.141811)	1274314 (725038)	1007.15 (8382.24)
19950301	45	2.40312E-06 (2.0679E-06)	0.245544 (0.028124)	2518321 (559400)	2.951E-07 (1.53E-06)	0.292476 (0.146232)	2589380 (2586763)	971.18 (8714.51)
19950601	46	2.75939E-06 (2.0977E-06)	0.226551 (0.027621)	3948162 (1314556)	1.711E-08 (5.89E-08)	0.384983 (0.085803)	1528601 (610514)	966.05 (9579.14)
19950901	47	3.10475E-06 (2.2611E-06)	0.227667 (0.023249)	3523990 (711520)	1.492E-06 (1.19E-05)	0.234395 (0.234477)	3783686 (8738450)	9.12 (14674.95)
19951201	48	1.21154E-06 (9.07696E-07)	0.259504 (0.020506)	2696175 (258989)	6.913E-08 (3.9E-07)	0.329974 (0.136243)	2017941 (1409084)	1000.67 (16037.42)
19960301	49	5.8092E-07 (9.37072E-07)	0.261847 (0.043979)	3704490 (961948)	3.611E-08 (1.7E-07)	0.346674 (0.111093)	1908728 (1145770)	999.95 (15892.18)
19960601	50	2.71115E-07 (4.0613E-07)	0.277722 (0.038380)	3707685 (673356)	1.087E-06 (1.55E-05)	0.234223 (0.364582)	3773780 (10987140)	9.95 (32371.49)
19960901	51	1.108E-06 (1.59319E-06)	0.250891 (0.036369)	3577962 (542689)	1.086E-06 (1.37E-05)	0.235479 (0.307426)	3778715 (8362409)	29.81 (35438.08)