

SIMPLE MULTIVARIATE TIME SERIES FOR SIMULATIONS OF COMPLEX SYSTEMS

Peter A. W. Lewis  
Department of Operations Research  
Naval Postgraduate School  
Monterey, California  
93940

ABSTRACT

Recent work has made the generation of univariate time series for inputs to stochastic systems quite simple. The time series are all random linear combinations of i.i.d. random variables with Exponential, Gamma and hyperexponential marginal distributions. The simplicity of structure of these time series models makes it practical to combine them to model multivariate situations. Thus one can model, for example, alternating sequences of response and think times at a terminal in which response and think times are not only autocorrelated, but also crosscorrelated.

Recent work by Gaver and Lewis (1980), Jacobs and Lewis (1977), Lawrance and Lewis (1977, 1980, 1981a, 1981b, 1981c) and others has made the generation of univariate time series for inputs to stochastic systems quite simple. These time series are all random linear combinations of i.i.d. random variables with Exponential, Gamma and hyperexponential marginal distributions. Thus the range of marginal distributions and correlation structures which can be modelled is very broad.

The original formulations of these models was as first-order autoregressive processes. They can, however, be extended to moving average and mixed autoregressive-moving average structures. The key to this extension is that for each marginal distribution  $F(x)$ , a way has been found to obtain a random linear combination of independent random variables with distribution  $F(x)$  so as to give a new random variable with the distribution  $F(x)$ . This is, of course, well known for normal random variables since linear, constant coefficient combinations of normal random variables are normal; this result is the basis for the widely used ARMA(p,q) models. The results are new for exponential, Gamma and hyperexponential distributions.

We describe here only the Gamma process, since it is new. Thus let  $G_n(k,\mu)$ ,  $n=0,1,2,\dots$  be an i.i.d. sequence of Gamma variates with shape parameter  $k(>0)$  and mean  $\mu$ . Let  $B_n(q,k-q)$  and  $C_n(k-q,q)$ ,  $n=1,2,\dots$  be mutually independent

sequences of i.i.d. Beta variables ( $q \leq k$ ), independent of  $\{G_n(k,\mu)\}$ . Set  $X_0 = G_0(k,\mu)$ . Then

$$X_n = B_n(q,k-q)X_{n-1} + C_n(k-q,q)G_n(k,\mu) \quad n=1,2,\dots$$

is a one parameter, autoregressive Markovian process with  $\text{Gamma}(k,\mu)$  marginal distributions for the  $X_n$ 's. It is more regular than the original, defective Gamma process given by Gaver and Lewis (1980). It is a special case, after reformulation, of the Gamma GBAR(1) process introduced by Lawrance and Lewis (1981b) and Fishman (private communication). However the broader GBAR(1) process is not as useful since it cannot always be expressed as a random linear combination of independent Gamma variables.

The corresponding first-order moving average Gamma process is

$$X_n = B_n(q,k-q)G_n(k,\mu) + C_n(k-q,q)G_{n+1}(k,\mu) \quad n=1,2,\dots$$

and combined autoregressive-moving average processes are easily constructed.

The simplicity of structure of these univariate series models makes it practical to combine them to model multivariate situations. This has been described for computer system modelling by Lewis and Shedler (1977) and for some simple queues by Jacobs (1978, 1980). As another example one can

model alternating sequences of response and think times at a terminal in which response and think times are not only autocorrelated but crosscorrelated.

A fairly broad scheme for generating these multivariate time series is given, and some of its properties are described.

#### REFERENCES

- Gaver, D.P. and P.A.W. Lewis (1980), "First order autoregressive Gamma sequences," Advances in Applied Probability, 12, 727-745.
- Jacobs, P.A. (1978), "A cyclic queueing network with dependent exponential service times," J. Appl. Prob., 15, 573-589.
- Jacobs, P.A. (1980), "Heavy traffic results for single server queues with dependent (EARMA) service and interarrival times," Advances in Applied Probability, 12, 517-529.
- Jacobs, P.A. and P.A.W. Lewis (1977), "A mixed autoregressive moving average exponential sequence and point process (EARMA 1,1)," Advances in Applied Probability, 9, 87-104.
- Lawrance, A.J. and P.A.W. Lewis (1977), "An exponential moving average sequence and point process (EMA1)," J. Appl. Prob., 14, 98-113.
- Lawrance, A.J. and P.A.W. Lewis (1980), "A mixed exponential time series model, NMEAR(p,q)." Naval Postgraduate School Technical Report NPS55-80-012. Submitted to Management Science.
- Lawrance, A.J. and P.A.W. Lewis (1981a), "The exponential autoregressive-moving EARMA(p,q) process," J. Roy. Stat. Soc. B., 42, 2, 150-161.
- Lawrance, A.J. and P.A.W. Lewis (1981b), "Generation of some first-order autoregressive Markovian sequences of positive random variables with given marginal distributions," Naval Postgraduate School Technical Report NPS55-81-003. To appear in Proc. Applied Probability/Computer Science Conference, R. Disney, Editor.
- Lawrance, A.J. and P.A.W. Lewis (1981c), "A new autoregressive time series model in exponential variables (NEAR(1))." To appear in J. App. Prob.