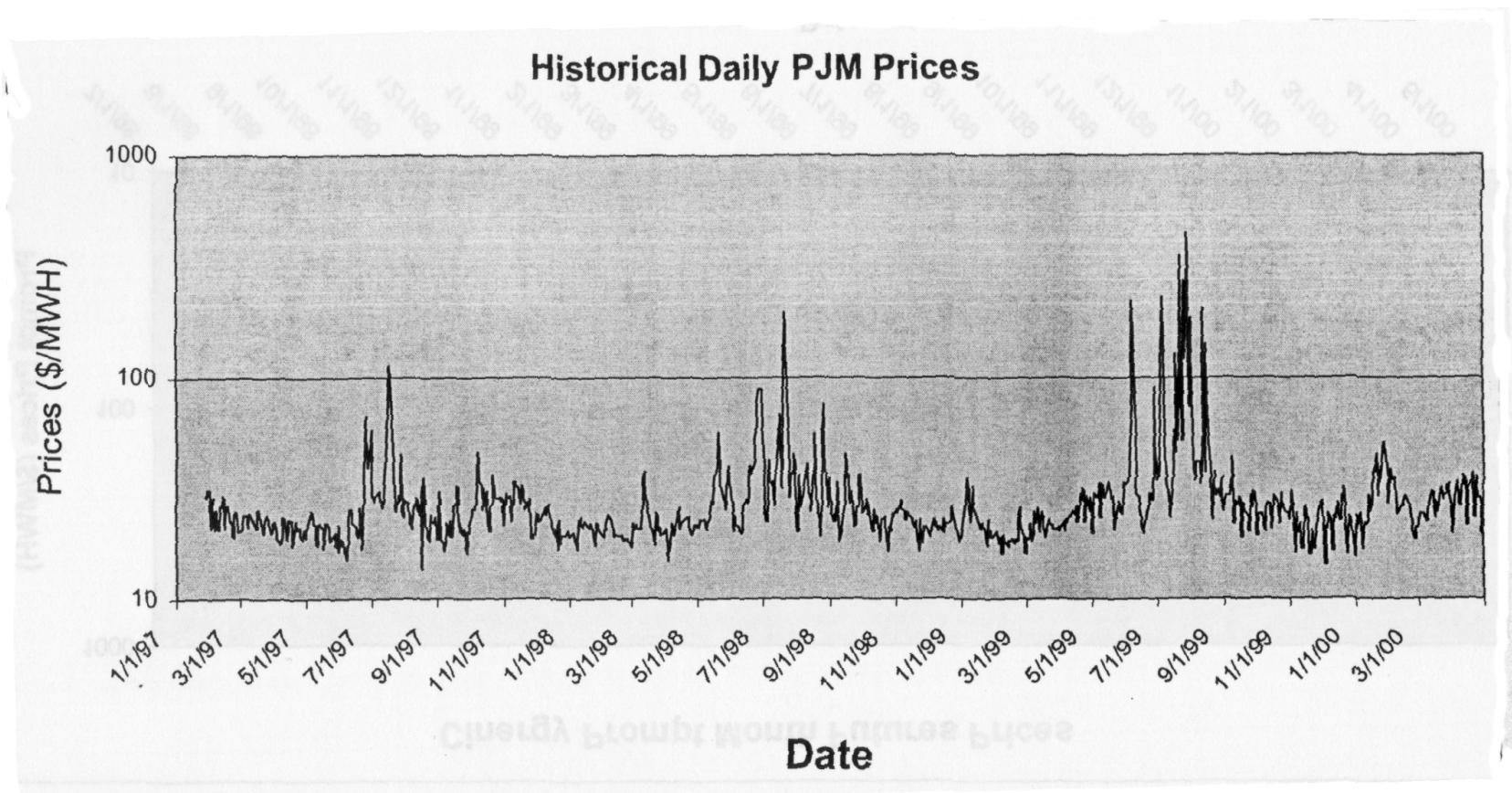


Introduction to Electricity Markets

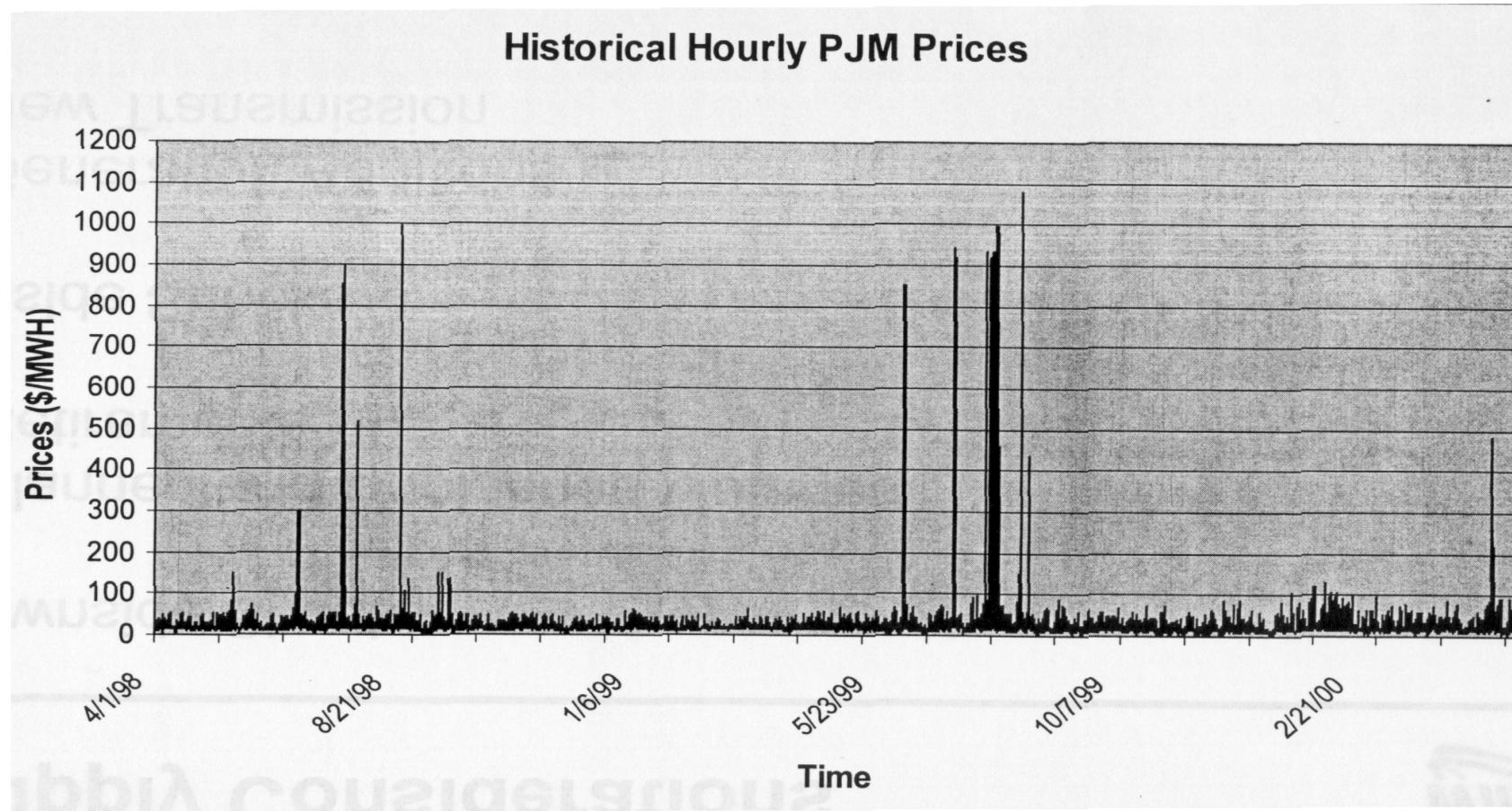
Pricing Theory

by Olivier Croissant

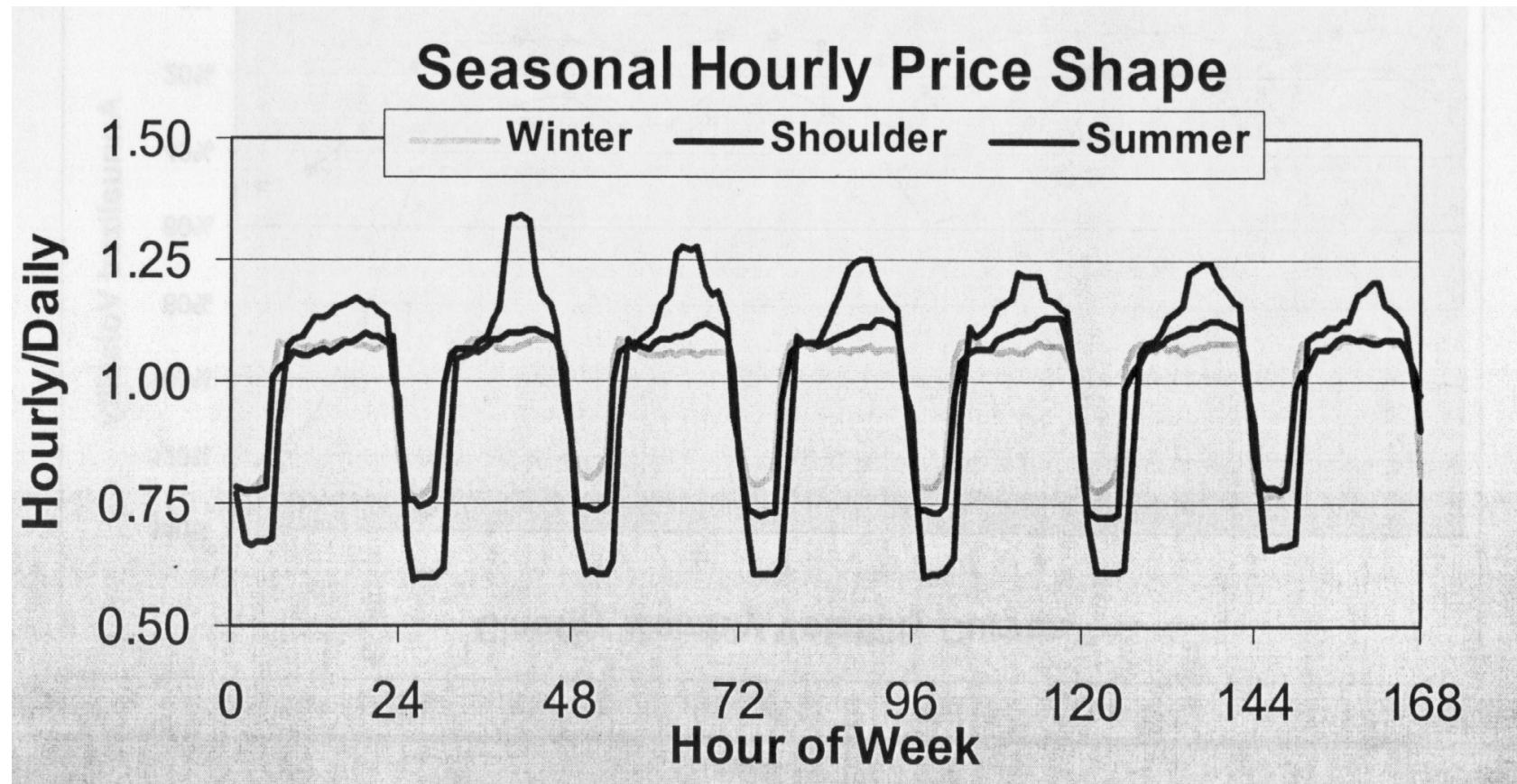
Daily Prices



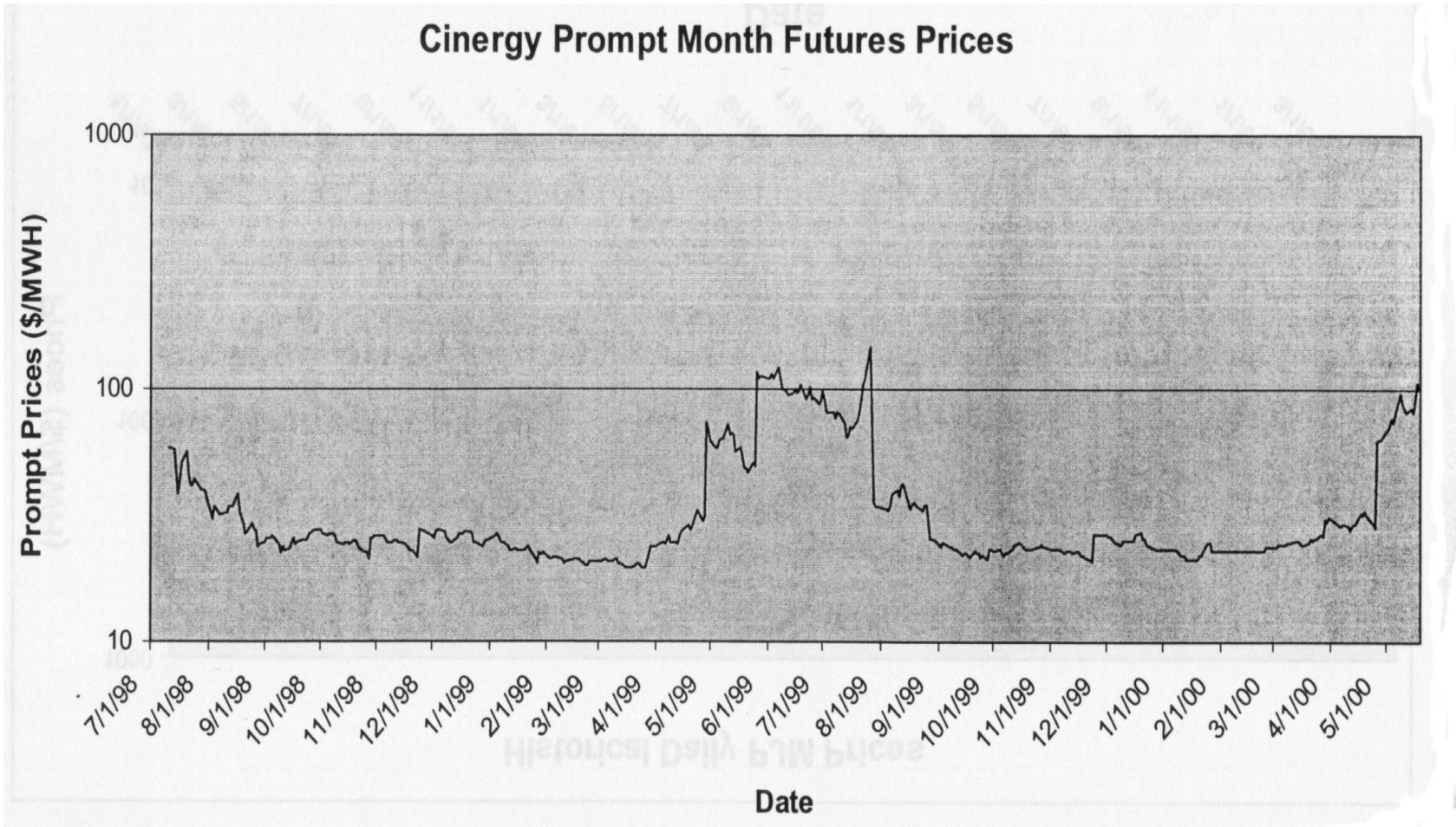
Hourly Prices



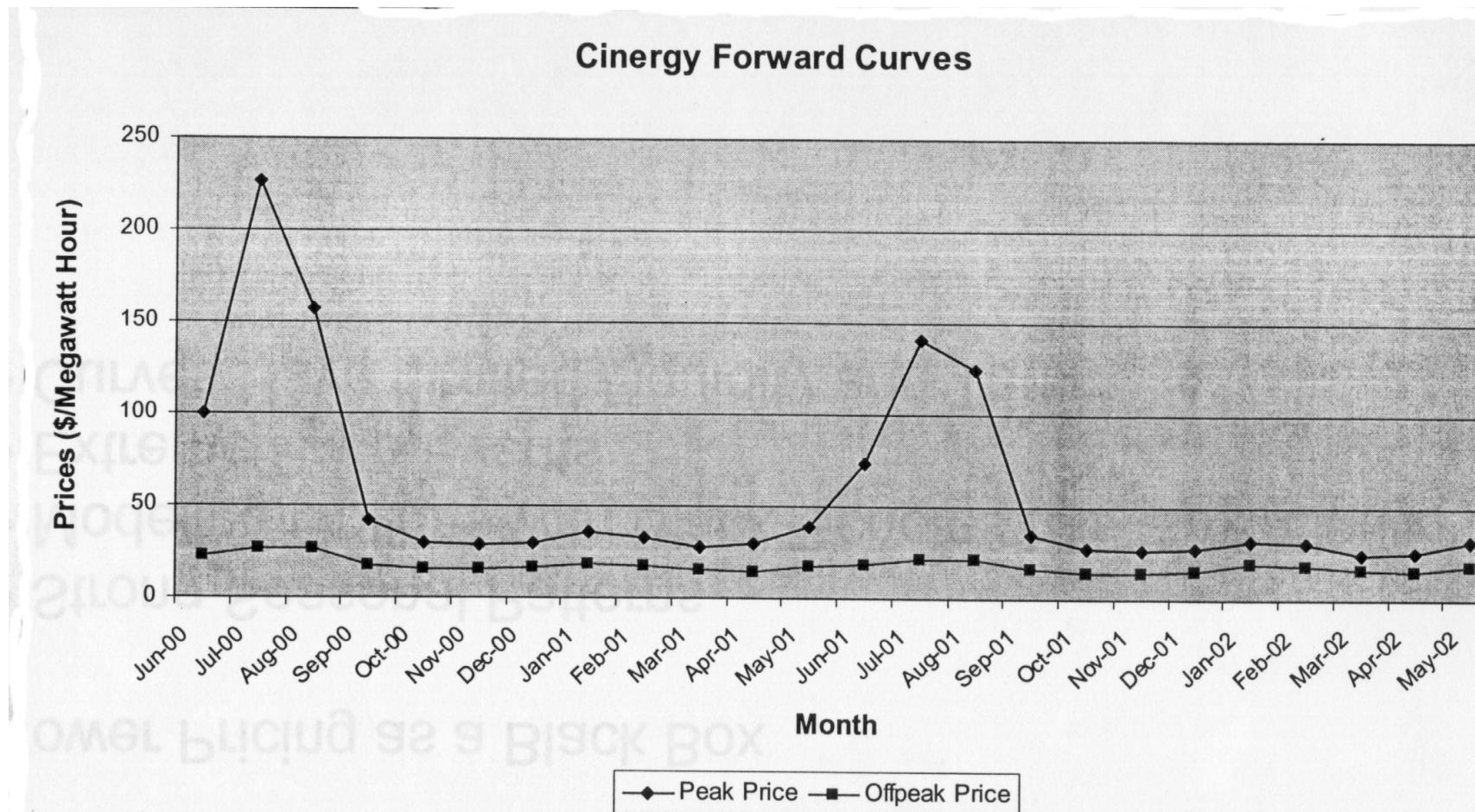
Hourly Seasonality



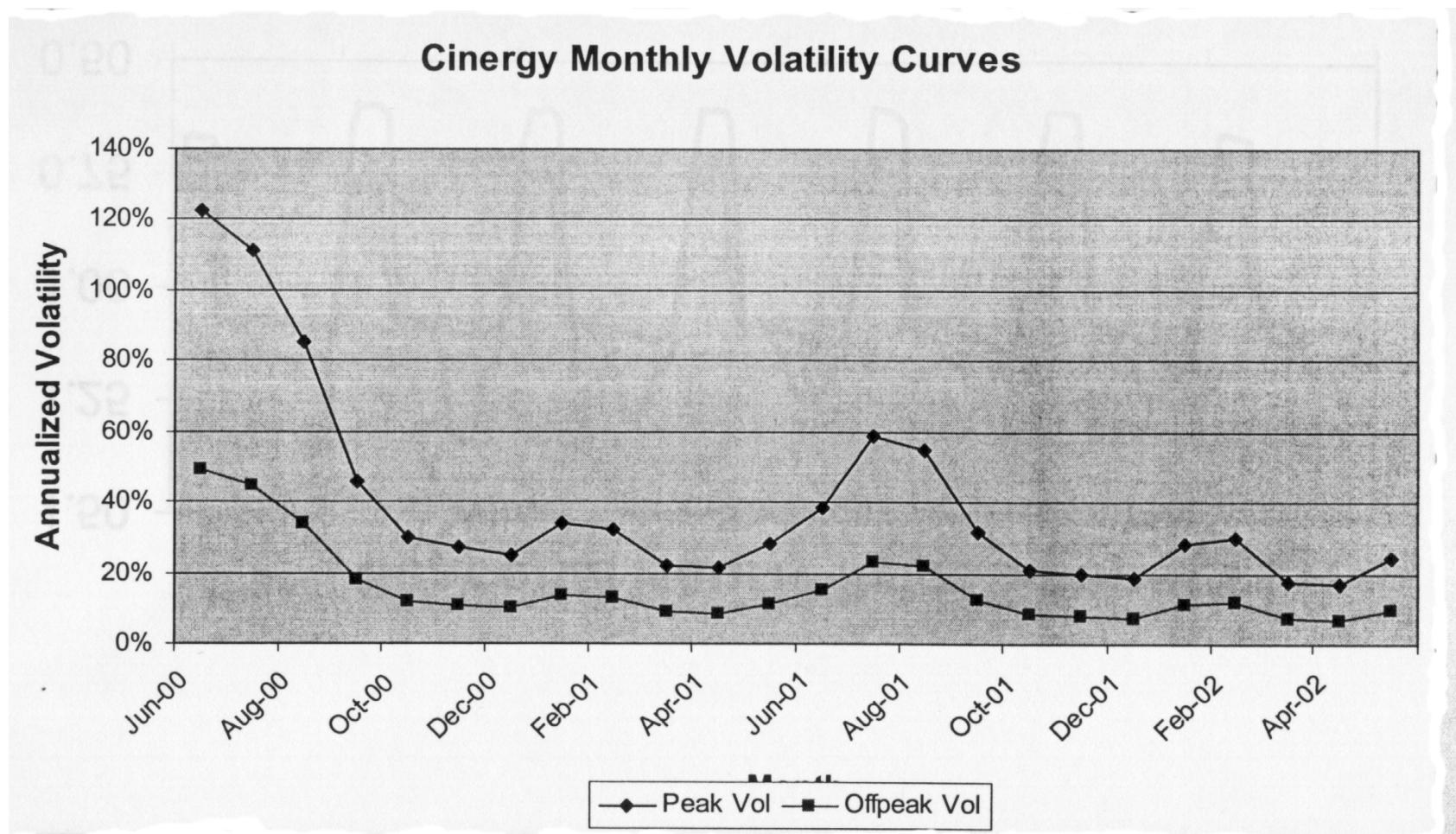
Prompt Month



Forward Curve



Historical Volatility of The next 1 month Contract



The Mathematical Opportunity

- Given a jump diffusion process:

$$dX_t = (\mu_0 + \mu_1 \cdot X_t)dt + (\sigma(X_t) \cdot X_t) \cdot dW_t + \sum_k J_k dq_k$$

$\sigma^2(x) = \Sigma_{0,t} + \Sigma_{1,t} \cdot x$ $v_k(X)$ $(\lambda_{k,0} + \lambda_{k,1} \cdot X_t)$

- Then the generalized transform :

$$\phi[u, X_t, t] = E_t \left[e^{\left(- \int_t^T (R_0 + R_1 \cdot X_s) ds \right)} u \cdot X_T \right]$$

can be

computed solving ordinary differential equations: $\phi[u, x, t] = e^{\alpha(t) + \beta(t) \cdot x}$

where $\frac{d\beta}{dt} = R_1 - \mu_1 \beta - \frac{1}{2} \beta \Sigma_1 \beta - \lambda_1 (\theta(\beta(t)) - 1)$, $\frac{d\alpha}{dt} = R_0 - \mu_0 \alpha - \frac{1}{2} \beta \Sigma_0 \beta - \lambda_0 (\theta(\beta(t)) - 1)$, $\beta(T) = u$, $\alpha(T) = 0$ $\theta(c) = \int e^{cz} dv_k(z)$

- The optional return can be computed using fourier transform :

$$G_{a,b}(y) = E \left[e^{\left(- \int_t^T (R_0 + R_1 \cdot X_s) ds \right)} a \cdot X_T \mathbf{1}_{b \cdot X_T \leq y} \right] = \frac{\phi[a, X_0, 0]}{2} - \frac{1}{\pi} \int_0^\infty \text{Im}[\phi[a + ivb, X_0, 0] e^{-ivy}] dv \quad \text{and}$$

$$E \left[e^{\left(- \int_t^T (R_0 + R_1 \cdot X_s) ds \right)} \left(e^{d \cdot X_T} - c \right)^+ \right] = G_{d,-d}(-\text{Log}[c], X_0, T) - c G_{0,-d}(-\text{Log}[c], X_0, T)$$

Extension of the preceding Framework

- We can add more terms to the expectation, to compute more options :

$$\phi[u, v, w, X_t, t] = E_t \left[e^{\left(-\int_t^T (R_0 + R_1 \cdot X_s) ds \right)} e^{u \cdot X_{T(v+w \cdot X_T)}} \right],$$

- We define the following security :

$$\tilde{G}_{a, b, d}(y) = E \left[e^{\left(-\int_t^T (R_0 + R_1 \cdot X_s) ds \right)} e^{d \cdot X_{T_a} \cdot X_T} \mathbb{1}_{b \cdot X_T \leq y} \right] = \frac{\Psi(a, d, x, 0)}{2} - \frac{1}{\pi} \int_0^\infty \text{Im}[\Psi(a, d + ibv, X_0, 0) e^{-ivy}] (dv)$$

- where $\Psi(v, u, x, t) = \phi[u, x, t](A(t) + B(t) \cdot x)$ and $\begin{cases} \frac{dB}{dt} = \mu_1 B + \frac{1}{2} B \Sigma_1 B + \lambda_1 \nabla \theta(\beta(t)) B \\ \frac{dA}{dt} = \mu_0 B + \frac{1}{2} B \Sigma_0 B + \lambda_0 \nabla \theta(\beta(t)) B \end{cases}$ with $\begin{cases} B(T) = v \\ A(T) = 0 \end{cases}$