

Keen model with Erlang distributed delay

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Keen Model

With capital assets being driven by

$$\dot{K} = \kappa(\pi_n)Y - \delta K \quad (1)$$

we get the following dynamical system

$$\dot{\omega} = \omega(\Phi(\lambda) - \alpha) \quad (2)$$

$$\dot{\lambda} = \lambda \left(\frac{\kappa(\pi_n)}{\nu} - \alpha - \beta - \delta \right) \quad (3)$$

$$\dot{d} = \kappa(\pi_n) - \pi_n - d \left(\frac{\kappa(\pi_n)}{\nu} - \delta \right) \quad (4)$$

Introducing the delay

Capital assets should be delayed from the moment of investment:

$$\dot{K}(t) = \kappa(\pi_n(t - \tau))Y(t - \tau) - \delta K(t) \quad (5)$$

To avoid complications related to Delayed-Differential Equations, we introduce investment stages:

$$\begin{aligned} \dot{\Theta}_1 &= \kappa(\pi_n)Y - \frac{n}{\tau}\Theta_1 \\ \dot{\Theta}_2 &= \frac{n}{\tau}(\Theta_1 - \Theta_2) \\ &\vdots \\ \dot{\Theta}_n &= \frac{n}{\tau}(\Theta_{n-1} - \Theta_n) \\ \dot{K} &= \frac{n}{\tau}\Theta_n - \delta K \\ \dot{D} &= (\kappa(\pi_n) - \pi_n)Y \end{aligned} \quad (6)$$

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- Dollar travels between the investment substages taking an exponential (with mean $\frac{\tau}{n}$) time in each of them.
- To see this, suppose that during investment stage k , the only process occurring was transference to stage $k + 1$. $\dot{\Theta}_k$ would then be

$$d\Theta_k/dt = -\frac{n}{\tau}\Theta_k \quad (7)$$

that is, if we start with $\$M$ dollars at time 0, $\$Me^{-\frac{n}{\tau}t}$ will remain there at time t .

- Still confused? If we start with $\$M$ at time 0, and dollars (or cents!) leave at an exponentially distributed time, at time t we can expect to still have

$$\$M \cdot \mathbb{P}[\text{exp. dist. r.v.} > t] = \$M(1 - F(t))^1 = \$Me^{-\frac{n}{\tau}t} \quad (8)$$

¹ $F(t)$ is the CDF for the exponentially distributed random variable describing the waiting time.

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- The total time it takes for each dollar invested then follows an Erlang distribution with shape parameter n and rate $\frac{n}{\tau}$, which has mean τ and variance τ^2/n .

$$\left(X_i \sim \text{Exponential}(n/\tau) \implies \sum_{i=1}^n X_i \sim \text{Erlang}(n, n/\tau) \right) \quad (9)$$

- In the limit $n \rightarrow \infty$, the distribution converges to a deterministic time delay of τ , which represents the Delayed-Differential Equation we tried to avoid.

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Dividing the Θ_k variables by Y , $\theta_k = \Theta_k/Y$, we can derive the $n + 3$ -dimensional system

$$\begin{aligned}
 \dot{\omega} &= \omega(\Phi(\lambda) - \alpha) \\
 \dot{\lambda} &= \lambda \left(\frac{n}{\tau\nu} \theta_n - (\alpha + \beta + \delta) \right) \\
 \dot{d} &= \kappa(\pi_n) - \pi_n - d \left(\frac{n}{\tau\nu} \theta_n - \delta \right) \\
 \dot{\theta}_1 &= \kappa(\pi_n) - \theta_1 \left[\frac{n}{\tau} \left(1 + \frac{1}{\nu} \theta_n \right) - \delta \right] \\
 \dot{\theta}_2 &= \frac{n}{\tau} (\theta_1 - \theta_2) - \theta_2 \left(\frac{n}{\tau\nu} \theta_n - \delta \right) \\
 &\vdots \\
 \dot{\theta}_k &= \frac{n}{\tau} (\theta_{k-1} - \theta_k) - \theta_k \left(\frac{n}{\tau\nu} \theta_n - \delta \right) \\
 &\vdots \\
 \dot{\theta}_n &= \frac{n}{\tau} (\theta_{n-1} - \theta_n) - \theta_n \left(\frac{n}{\tau\nu} \theta_n - \delta \right)
 \end{aligned} \tag{10}$$

The “good” equilibrium has become

$$\begin{aligned}
 \hat{\lambda}_1 &= \Phi^{-1}(\alpha) \\
 \hat{\theta}_{n,1} &= \frac{\tau\nu}{n}(\alpha + \beta + \delta) \\
 &\vdots \\
 \hat{\theta}_{n-k,1} &= \hat{\theta}_{n,1} \left[\frac{\tau}{n}(\alpha + \beta + n/\tau) \right]^k \\
 &\vdots \\
 \hat{\theta}_{1,1} &= \hat{\theta}_{n,1} \left[\frac{\tau}{n}(\alpha + \beta + n/\tau) \right]^{n-1} \\
 \hat{\pi}_{n,1} &= \kappa^{-1} \left[\hat{\theta}_{1,1}(\alpha + \beta + n/\tau) \right] \\
 \hat{d}_1 &= \frac{\kappa(\hat{\pi}_{n,1}) - \hat{\pi}_{n,1}}{\alpha + \beta} \\
 \hat{\omega}_1 &= 1 - \hat{\pi}_{n,1} - r\hat{d}_1
 \end{aligned} \tag{11}$$

The Jacobian matrix for the linearized system at this equilibrium is

$$\begin{bmatrix}
 0 & \bar{\omega}\Phi'(\bar{\lambda}) & 0 & 0 & 0 & \dots & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & \dots & 0 & 0 & \frac{n}{\tau\nu}\bar{\lambda} \\
 1 - \kappa'(\bar{\pi}_n) & 0 & r - r\kappa'(\bar{\pi}_n) - (\alpha + \beta) & 0 & 0 & \dots & 0 & 0 & -\frac{n}{\tau\nu}\bar{d} \\
 -\kappa'(\bar{\pi}_n) & 0 & -r\kappa'(\bar{\pi}_n) & -\frac{\alpha + \beta}{\tau} & 0 & \dots & 0 & 0 & -\frac{n}{\tau\nu}\bar{\theta}_1 \\
 0 & 0 & 0 & \frac{n}{\tau} & -\frac{\alpha + \beta}{\tau} & \dots & 0 & 0 & -\frac{n}{\tau\nu}\bar{\theta}_2 \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\
 0 & 0 & 0 & 0 & 0 & \dots & \frac{n}{\tau} & -\frac{\alpha + \beta}{\tau} & -\frac{n}{\tau\nu}\bar{\theta}_{n-1} \\
 0 & 0 & 0 & 0 & 0 & \dots & 0 & \frac{n}{\tau} & -\frac{n}{\tau} - \frac{2\alpha}{\tau} - 2\beta - \delta
 \end{bmatrix}
 \quad (12)$$

Armed with the Jacobian, we can investigate when stability is lost for each n , in terms of τ .

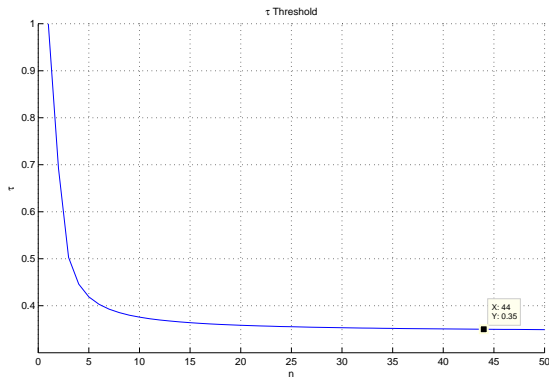


Figure 1: Stability threshold value for τ as a function of n

Using XPPAUT, we verified that there is a **supercritical** Hopf bifurcation for τ larger than the threshold seen on Figure 1. The stable equilibrium point unfolds in a stable cycle, while the equilibrium point loses its local stability, Figure 2.

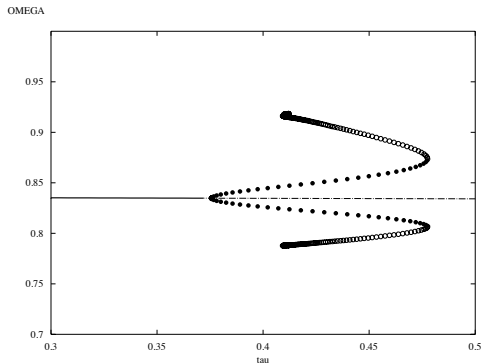


Figure 2: Supercritical Hopf bifurcation for $n = 10$

Simulations

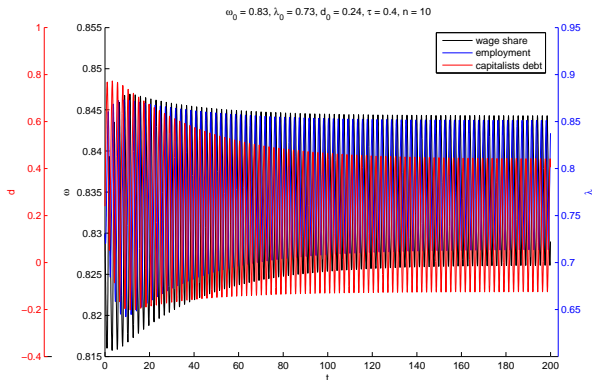


Figure 3: Solution converging to the stable cycle, $n = 10$

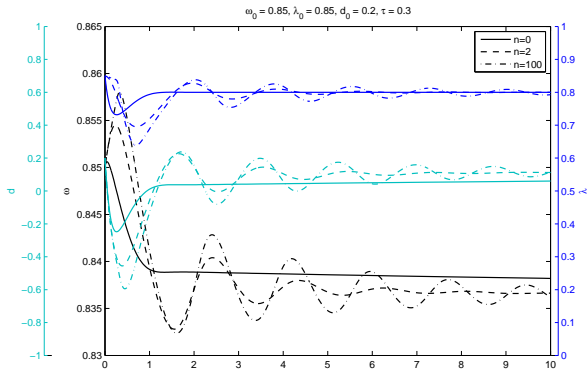


Figure 4: Solutions for different values of n

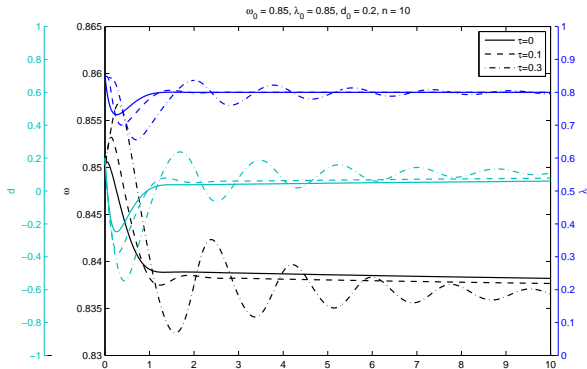


Figure 5: Solutions for different values of τ