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# NER551

## Nuclear Reactor Dynamics

### Numerical Solution of the Exact Point Kinetics Equations

3-8-2022

Prof T. Downar

# Outline

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- **Review Exact P.K. Equations (EPKE)**
  - **Derivation (See Backup Slides)**
  - **Comparison w/ “Conventional” P.K.**
- Numerical Solution of P.K. Equations
  - Theta method for temporal discretization
  - Exponential transformation
  - Precursor time integration
  - Solution of EPK
  - Linear feedback model
  - Solution of EPK with feedback
- Application to NEACRP PWR Rod Ejection

# Compare Exact and Conventional Point Kinetics

- Exact Point Kinetics

$$\frac{dp(t)}{dt} = \frac{\rho(t) - \beta^{\text{eff}}(t)}{\Lambda(t)} p(t) + \frac{1}{\Lambda_0} \sum_k \lambda_k(t) \zeta_k(t)$$

$$\frac{d\zeta_k(t)}{dt} = \frac{\Lambda_0}{\Lambda(t)} \beta_k^{\text{eff}}(t) p(t) - \lambda_k(t) \zeta_k(t), \quad k = 1, 2, \dots$$

- EPK Parameters

$$\Lambda(t) = \frac{1}{p(t)F(t)} \langle \phi_g^*(\mathbf{r}) \frac{1}{v_g} \phi_g(\mathbf{r}, t) \rangle$$

$$\beta_k^{\text{eff}}(t) = \frac{1}{p(t)F(t)} \langle \phi_g^*(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) \beta_k(\mathbf{r}) S^F(\mathbf{r}, t) \rangle,$$

$$\lambda_k(t) = \frac{\langle \phi_g^*(\mathbf{r}) \lambda_k(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) C_k(\mathbf{r}, t) \rangle}{\langle \phi_g^*(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) C_k(\mathbf{r}, t) \rangle}$$

$$\rho(t) = \frac{1}{p(t)F(t)} \langle \phi^*(r) (F - M) \phi(r, t) \rangle$$

- Conventional Point Kinetics

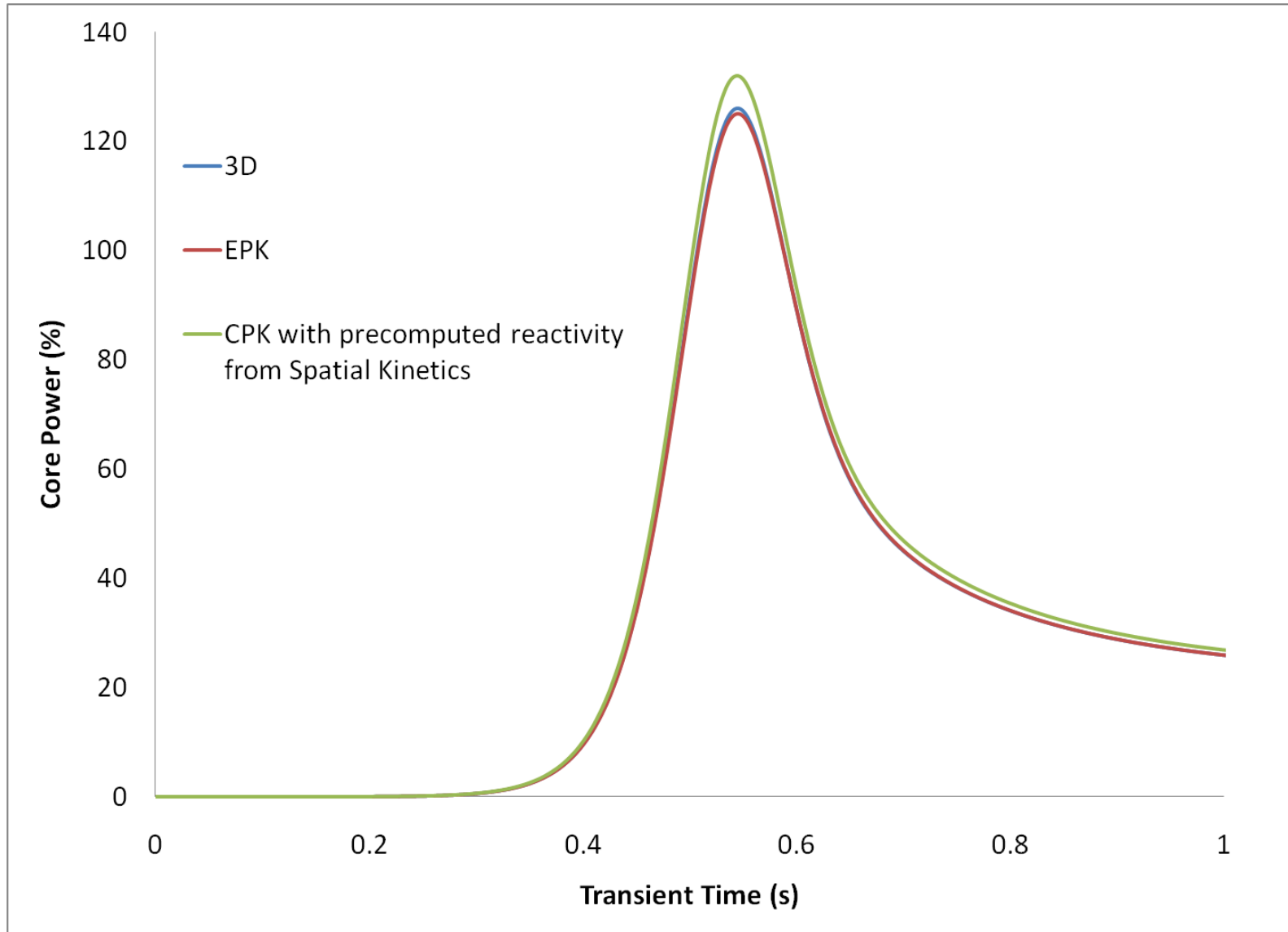
$$\frac{dp(t)}{dt} = \frac{\rho(t) - \beta^{\text{eff}}}{\Lambda_0} p(t) + \frac{1}{\Lambda_0} \sum_k \lambda_k \zeta_k(t)$$

$$\frac{d\zeta_k(t)}{dt} = \beta_k^{\text{eff}} p(t) - \lambda_k \zeta_k(t), \quad k = 1, 2, \dots$$

- CPK Parameters

- Initial  $\Lambda, \beta, \lambda$  are used for entire transient,
- The reactivity are often evaluated with core average parameters and reactivity coefficients which were evaluated near steady state condition.
- It is hard to obtained accurate reactivity for CPK.

# EPK and CPK for Rod Eject Benchmark(A1)



# Benefits from Exact Point Kinetics

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- As EPK parameters rely on solutions of spatial kinetics, in order to obtain EPK solution, the spatial kinetics solution is always needed. So, what are the Benefits from EPK?
  1. Help user to analyze the mechanism of transient results from spatial kinetics,
  2. Identify the error sources of conventional point kinetics,
  3. Accelerate the spatial kinetics solution without impacting the accuracy (e.g. improved quasi-static method).

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  - **Theta method for temporal discretization**
  - **Exponential transformation**
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  - **Solution of EPK**
  - **Linear feedback model**
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# Temporal Discretization Method for Point Kinetics

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- **Theta method** is used for flux level equation. Explicit, implicit and Crank-Nicolson methods can be realized with 0, 1 and  $\frac{1}{2}$  as theta values.
- **Exponential transformation** is used to reduce the stiffness of point kinetic equations.
- The **precursors will be analytically** integrated over time step with linear approximation of flux level during the time step.

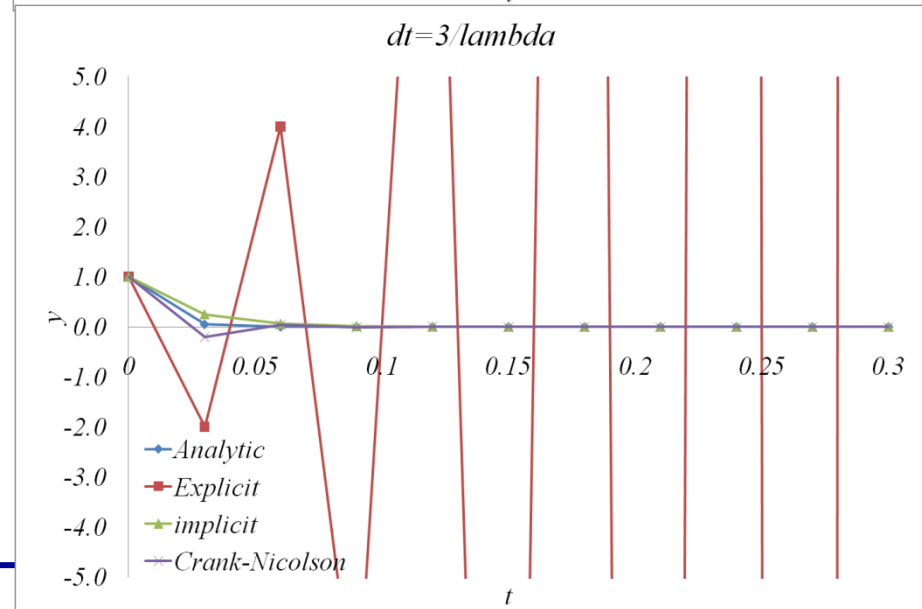
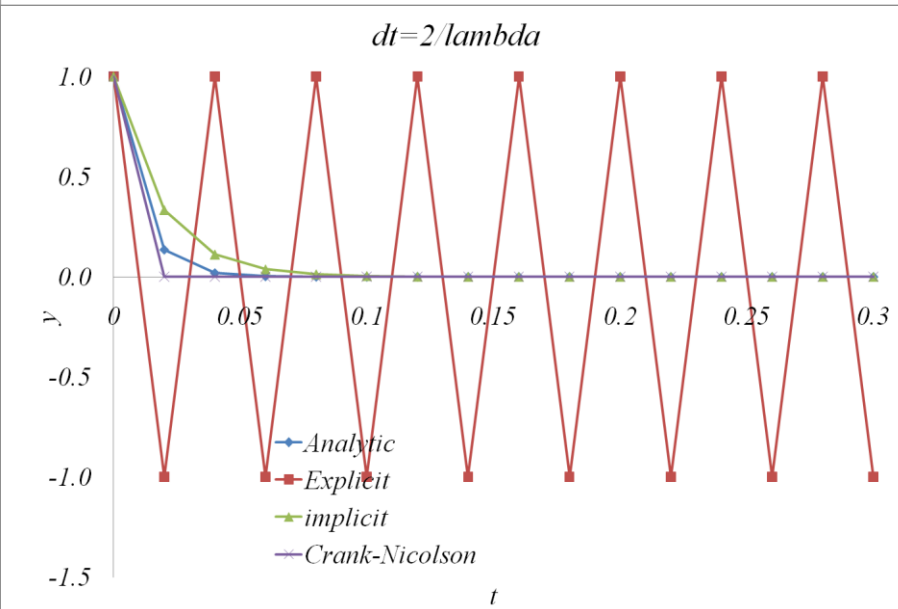
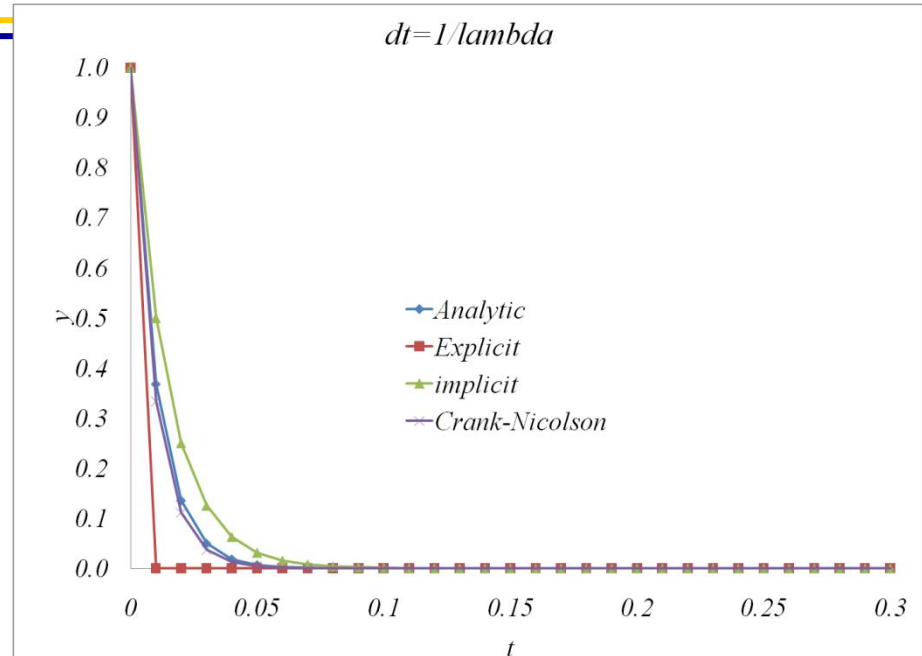
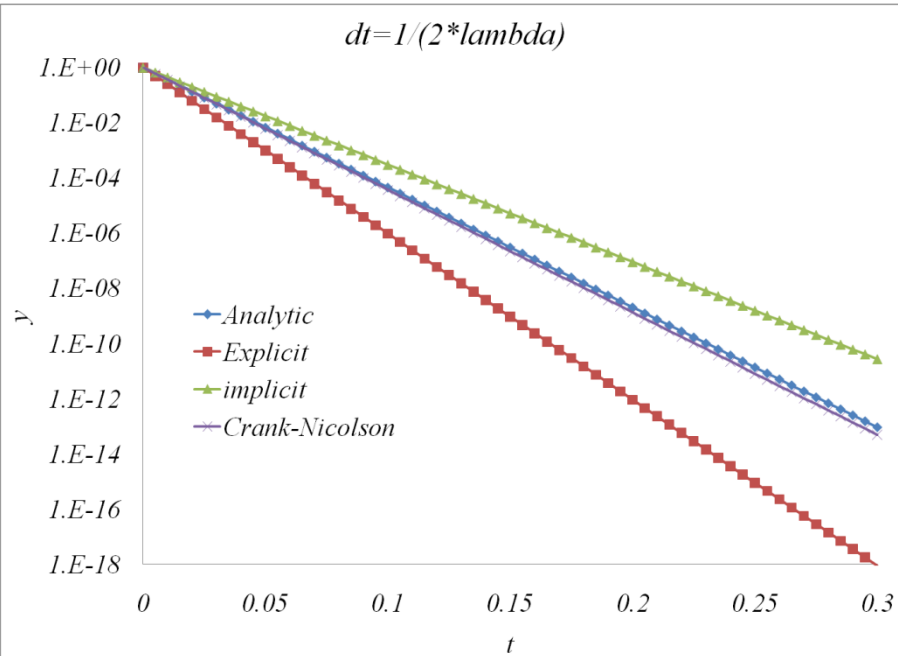
# Theta method (Review?)

- System of ordinary differential Equations  $\frac{dy(t)}{dt} = f(t, y(t))$
- Theta method  $y^n - \theta \Delta t^n f(t_n, y^n) = y^{n-1} - (1 - \theta) \Delta t^n f(t_{n-1}, y^{n-1})$
- Test problem  $\frac{dy(t)}{dt} = -\lambda y(t)$   $TE = FDE - PDE$
- Explicit method:  $\theta = 0: y^n = (1 - \lambda \Delta t^n) y^{n-1}, \quad TE = \frac{1}{2} y''(t_{n-1}) \Delta t^n + \frac{1}{6} y'''(t_{n-1}) (\Delta t^n)^2 + \dots$
- Implicit method:  $\theta = 1: y^n = \frac{1}{1 + \lambda \Delta t^n} y^{n-1}, \quad TE = -\frac{1}{2} y''(t_n) \Delta t^n + \frac{1}{6} y'''(t_n) (\Delta t^n)^2 + \dots$
- Crank-Nicolson:  $\theta = \frac{1}{2}: y^n = \frac{2 - \lambda \Delta t^n}{2 + \lambda \Delta t^n} y^{n-1}, \quad TE = \frac{1}{24} y'''(t_n + \frac{\Delta t^n}{2}) (\Delta t^n)^2 + \dots$
- Explicit and implicit methods are first order in time, Crank-Nicolson is second order in time which is more accurate for small time step size, i.e.

$$\Delta t < \left| \frac{y''(t)}{y'''(t)} \right|, \quad \text{which is } \left| \frac{1}{\lambda} \right| \text{ for test problem}$$

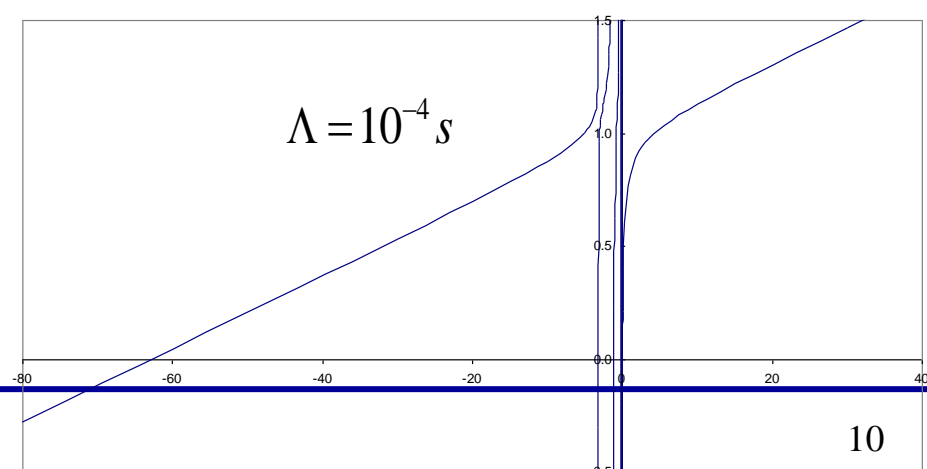
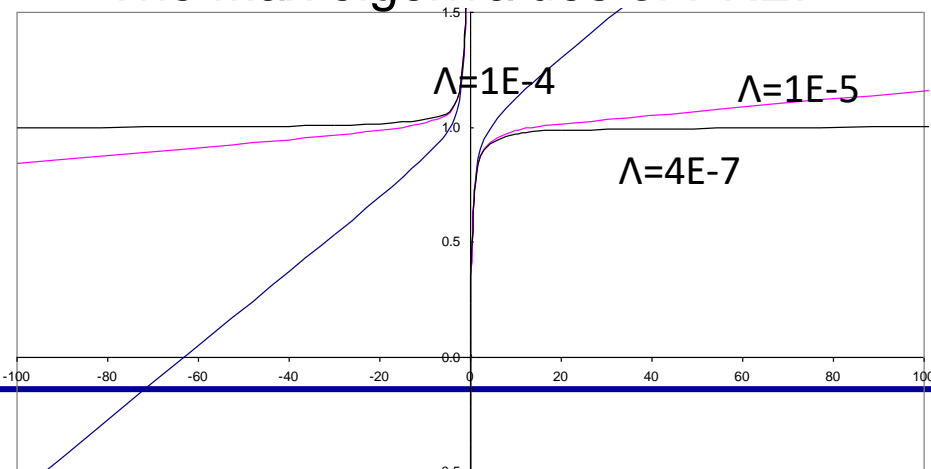


# Theta method solution for test problem



# Stability of Theta Method

- Implicit method is unconditionally stable, Crank-Nicolson is unconditionally neutral stable, explicit method is conditionally stable.
- For the test problem, explicit method is stable when  $\Delta t < \left| \frac{1}{\lambda} \right|$
- $\lambda$  is called the stiffness of test problem.
- For system of linear ordinary differential equations  $\frac{dy(t)}{dt} = Ay(t)$ , the stiffness is  $\max |\lambda_i|$ , where  $\lambda_i$  is eigenvalues of A
- The max eigenvalues of PKE:



# Theta method applied to EPKE

- Theta method for flux amplitude equation

$$e^{-\alpha^n \Delta t_n} p^n - p^{n-1} = \theta \Delta t_n e^{-\alpha^n \Delta t_n} \left( \left( \frac{\rho^n - \beta^{eff,n}}{\Lambda^n} - \alpha^n \right) p^n + \frac{1}{\Lambda_0} \sum_k \lambda_k \zeta_k^n \right) \\ + (1-\theta) \Delta t_n \left( \left( \frac{\rho^{n-1} - \beta^{eff,n-1}}{\Lambda^{n-1}} - \alpha^n \right) p^{n-1} + \frac{1}{\Lambda_0} \sum_k \lambda_k \zeta_k^{n-1} \right) \quad (7)$$

Where  $\theta$  is user input parameter.

If  $\theta=0$ , the finite differencing is explicit,

If  $\theta=1$ , the finite differencing is implicit,

If  $\theta=0.5$ , the finite differencing is called Crank-Nicolson scheme, which has high accuracy ( 2<sup>nd</sup> order)

If same method is apply to precursor equations (6), then we get K+1 equations for K+1 unknowns.

As the precursors depend only on power, and are independent from each other, we can get solution without introducing finite different error to precursor equations.

# Exponential Transformation

- Test problem:  $\frac{dy(t)}{dt} = -\lambda y(t)$
- Analytic Solution:  $y(t) = y(0)e^{-\lambda t}$
- Stiffness:  $\lambda$
- Exponential Transformation:  $y(t) = \tilde{y}(t)e^{\alpha t}$

$$\frac{dy(t)}{dt} = \frac{d\tilde{y}(t)}{dt} e^{\alpha t} + \alpha \tilde{y}(t) e^{\alpha t} = -\lambda \tilde{y}(t) e^{\alpha t}$$

$$\frac{d\tilde{y}(t)}{dt} = -(\lambda + \alpha) \tilde{y}(t)$$

- Stiffness after Exponential Transformation  $|\lambda + \alpha|$
- If estimated alpha is close to negative of lambda, then the stiffness is “softened”.

# Exponential Transformation for PKE

- Exact Point Kinetic Equations

$$\frac{dp(t)}{dt} = \frac{\rho(t) - \beta^{eff}(t)}{\Lambda(t)} p(t) + \frac{1}{\Lambda_0} \sum_k \lambda_k \zeta_k(t), \quad (1)$$

$$\frac{d\zeta_k(t)}{dt} = \frac{\Lambda_0}{\Lambda(t)} \beta_k^{eff}(t) p(t) - \lambda_k \zeta_k(t), \quad (2)$$

- Initial Condition  $\zeta_k(0) = \frac{\beta_k^{eff}(0)}{\lambda_k} p(0), \quad (3)$

- Exponential Transformation

$$p(t) = \tilde{p}(\tilde{t}) e^{\alpha^n \tilde{t}}, \quad \tilde{t} = t - t_{n-1} \in [-\Delta t_{n-1}, \Delta t_n], \quad (4) \quad \alpha^n = \frac{1}{\Delta t_{n-1}} \ln \frac{p^{n-1}}{p^{n-2}}, \quad (23)$$

$$\tilde{p}^{n-1} = p^{n-1}$$

$$\frac{d\tilde{p}(\tilde{t})}{d\tilde{t}} = \left( \frac{\rho(\tilde{t}) - \beta^{eff}(\tilde{t})}{\Lambda(\tilde{t})} - \alpha^n \right) \tilde{p}(\tilde{t}) + \frac{e^{-\alpha^n \tilde{t}}}{\Lambda_0} \sum_k \lambda_k \zeta_k(\tilde{t}), \quad (5)$$

$$\frac{d}{d\tilde{t}} \left( \zeta_k(\tilde{t}) e^{\lambda_k \tilde{t}} \right) = e^{\lambda_k \tilde{t}} \frac{d\zeta_k(\tilde{t})}{d\tilde{t}} + \lambda_k e^{\lambda_k \tilde{t}} \zeta_k(\tilde{t}) = \frac{\Lambda_0}{\Lambda(\tilde{t})} \beta_k^{eff}(\tilde{t}) p(\tilde{t}) e^{\lambda_k \tilde{t}} \quad (6)$$

# Analytic Integration of Precursors

- **Theta method** is used for flux level equation. Explicit, implicit and Crank-Nicolson methods can be realized with 0, 1 and  $\frac{1}{2}$  as theta values.
- **Exponential transformation** is used to reduce the stiffness of point kinetic equations.
- The **precursors will be analytically** integrated over time step with linear approximation of flux level during the time step.
- Exact Point Kinetics

$$\frac{dp(t)}{dt} = \frac{\rho(t) - \beta^{eff}(t)}{\Lambda(t)} p(t) + \frac{1}{\Lambda_0} \sum_k \lambda_k(t) \zeta_k(t)$$

$$\frac{d\zeta_k(t)}{dt} = \frac{\Lambda_0}{\Lambda(t)} \beta_k^{eff}(t) p(t) - \lambda_k(t) \zeta_k(t), \quad k = 1, 2, \dots$$

# Analytic Integration of Transformed Linear Precursor

$$\frac{d}{d\tilde{t}} \left( \zeta_k(\tilde{t}) e^{\lambda_k \tilde{t}} \right) = e^{\lambda_k \tilde{t}} \frac{d\zeta_k(\tilde{t})}{d\tilde{t}} + \lambda_k e^{\lambda_k \tilde{t}} \zeta_k(\tilde{t}) = \frac{\Lambda_0}{\Lambda(\tilde{t})} \beta_k^{eff}(\tilde{t}) p(\tilde{t}) e^{\lambda_k \tilde{t}} \quad (6)$$

- Assume  $\frac{\Lambda_0}{\Lambda(\tilde{t})} \beta_k^{eff}(\tilde{t}) p(\tilde{t}) e^{-\alpha \tilde{t}}$  to be linear during step

$$G(\tilde{t}) \equiv \frac{\Lambda_0}{\Lambda(\tilde{t})} \beta_k^{eff}(\tilde{t}) p(\tilde{t}) e^{-\alpha \tilde{t}} \approx G^n w + G^{n-1} (1 - w) \quad (8)$$

$$w = \tilde{t} / \Delta t_n$$

- Integrate eq. (6) over step

$$\zeta_k^n = e^{-\lambda_k \Delta t_n} \zeta_k^{n-1} + e^{\alpha \Delta t_n} \Delta t_n \{ G^n \kappa_1(\tilde{\lambda}_k^n) + G^{n-1} [\kappa_0(\tilde{\lambda}_k^n) - \kappa_1(\tilde{\lambda}_k^n)] \} \quad (9)$$

$$\tilde{\lambda}_k^n = (\lambda_k + \alpha) \Delta t_n$$

# Exponential Integration

$$\kappa_m(x) = e^{-x} \int_0^1 w^m e^{xw} dw = \frac{1 - m\kappa_{m-1}(x)}{x} \quad (10)$$

$$|x| < 0.06$$

$$\kappa_0(x) = (1 - e^{-x})/x \quad \kappa_0(x) = 1 - \frac{x}{2} + \frac{x^2}{6} - \frac{x^3}{24} + \frac{x^4}{120} - \frac{x^5}{720} + \frac{x^6}{5040} + O[x]^7 \quad (11)$$

$$\kappa_1(x) = [1 - \kappa_0(x)]/x \quad \kappa_1(x) = \frac{1}{2} - \frac{x}{6} + \frac{x^2}{24} - \frac{x^3}{120} + \frac{x^4}{720} - \frac{x^5}{5040} + \frac{x^6}{40320} + O[x]^7 \quad (12)$$

$$\kappa_0(x) - \kappa_1(x) = \frac{1}{2} - \frac{x}{3} + \frac{x^2}{8} - \frac{x^3}{30} + \frac{x^4}{144} - \frac{x^5}{840} + \frac{x^6}{5760} + O[x]^7 \quad (13)$$



# Integrated Precursor Transformed Linear Equation

- Integrated precursor  $\zeta_k^n = p^n \Omega_k^n + \hat{\zeta}_k^n, \quad (14)$

$$\Omega_k^n = \frac{\Lambda^0}{\Lambda^n} \beta_k^{eff,n} \Delta t_n \kappa_1(\tilde{\lambda}_k^n) \quad (15)$$

$$\hat{\zeta}_k^{n-1} = e^{-\lambda_k \Delta t_n} \zeta_k^{n-1} + e^{\alpha \Delta t_n} \Delta t_n G^{n-1} [\kappa_0(\tilde{\lambda}_k^n) - \kappa_1(\tilde{\lambda}_k^n)] \quad (16)$$

- Delay Neutron terms

$$\sum_k \lambda_k \zeta_k^n = p^n \sum_k \lambda_k \Omega_k^n + \sum_k \lambda_k \hat{\zeta}_k^{n-1} = \tau^n p^n + \hat{S}_d^n \quad (17)$$

$$\tau^n = \sum_k \lambda_k \Omega_k^n, \quad \hat{S}_d^n = \sum_k \lambda_k \hat{\zeta}_k^n, \quad (18)$$

$$S_d^{n-1} = \sum_k \lambda_k \zeta_k^{n-1}, \quad (19)$$

# Solution of EPKE

- Theta method for flux level equation

$$\begin{aligned} e^{-\alpha^n \Delta t_n} p^n - p^{n-1} = & \theta \Delta t_n e^{-\alpha^n \Delta t_n} \left( \left( \frac{\rho^n - \beta^{eff,n}}{\Lambda^n} - \alpha^n \right) p^n + \frac{\tau^n}{\Lambda_0} p^n + \frac{1}{\Lambda_0} \hat{S}_d^n \right) \\ & + (1-\theta) \Delta t_n \left( \left( \frac{\rho^{n-1} - \beta^{eff,n-1}}{\Lambda^{n-1}} - \alpha^n \right) p^{n-1} + \frac{S_d^{n-1}}{\Lambda_0} \right) \end{aligned} \quad (20)$$

- Flux Level Solution

$$p^n = \frac{e^{\alpha^n \Delta t_n} \left( p^{n-1} + (1-\theta) \Delta t_n \left( \left( \frac{\rho^{n-1} - \beta^{eff,n-1}}{\Lambda^{n-1}} - \alpha^n \right) p^{n-1} + \frac{S_d^{n-1}}{\Lambda_0} \right) \right) + \theta \Delta t_n \frac{\hat{S}_d^n}{\Lambda_0}}{1 - \theta \Delta t_n \left( \frac{\rho^n - \beta^{eff,n}}{\Lambda^n} - \alpha^n + \frac{\tau^n}{\Lambda_0} \right)} \quad (21)$$

- Fission Power

$$H^n = f_{fp} p^n, \quad (22)$$

# Exponential transformation parameter

- First take parameter,  $\alpha^n = \frac{1}{\Delta t_{n-1}} \ln \frac{p^{n-1}}{p^{n-2}}, \quad (23)$
- Evaluate the power.
- The transformed solution is closer to linear than original solution, and the transformation parameter is accepted, if

$$\left| p^n - e^{\alpha^n \Delta t_n} p^{n-1} \right| \leq \left| p^n - p^{n-1} - (p^{n-1} - p^{n-2}) / \gamma \right|, \quad (24)$$

- Otherwise re-evaluate solution without transformation, i.e.

$$\alpha^n = 0.$$

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# Linear Feed Back Model

- Reactivity feed back

$$\rho(t) = \rho_{im}(t) + \gamma_d \int_0^t (H(t') - P_0) \exp(\lambda_H(t' - t)) dt' \quad (25)$$

imposed reactivity      feed back constant      conduction time constant

- Doppler feedback       $\gamma_d = \frac{H \bullet 1s}{mc_p} \frac{d\rho}{dT_f} \quad (26)$

- Feed back Reactivity

$$\rho_d(t) = \rho(t) - \rho_{im}(t) = \gamma_d \int_0^t (H(t') - P_0) \exp(\lambda_H(t' - t)) dt' \quad (27)$$

# Feedback Reactivity at Next Step

- Feed back Reactivity at new step

$$\rho_d^n = \gamma_d \int_0^{t_n} (H(t') - P_0) \exp(\lambda_H(t' - t_n)) dt' \quad \hat{\lambda}_H^n = \lambda_H \Delta t_n$$

$$\rho_d^n = e^{-\hat{\lambda}_H^n} \rho_d^{n-1} - P_0 \gamma_d \Delta t_n \kappa_0(\hat{\lambda}_H^n) + \gamma_d e^{-\hat{\lambda}_H^n} \int_0^{\Delta t_n} H(\tilde{t}) \exp(\lambda_H \tilde{t}) d\tilde{t} \quad (28)$$

- Linear approximation

$$\tilde{\lambda}_H^n = (\lambda_H + \alpha) \Delta t_n$$

$$Q(\tilde{t}) = f_{fp}(\tilde{t}) p(\tilde{t}) e^{-\alpha \tilde{t}} = Q^n w + Q^{n-1} (1 - w) \quad (29)$$

$$\int_0^{\Delta t_n} Q(\tilde{t}) \exp(\lambda_H + \alpha) \tilde{t} d\tilde{t} = e^{\tilde{\lambda}_H^n} \Delta t_n \{ Q^n \kappa_1(\tilde{\lambda}_H^n) + Q^{n-1} [\kappa_0(\tilde{\lambda}_H^n) - \kappa_1(\tilde{\lambda}_H^n)] \} \quad (30)$$

$$\rho_d^n = e^{-\hat{\lambda}_H^n} \rho_d^{n-1} - \eta P_0 \gamma_d \Delta t_n \kappa_0(\hat{\lambda}_H^n) + e^{\alpha \Delta t_n} \gamma_d \Delta t_n \{ Q^n \kappa_1(\tilde{\lambda}_H^n) + Q^{n-1} [\kappa_0(\tilde{\lambda}_H^n) - \kappa_1(\tilde{\lambda}_H^n)] \}$$

# Linear Expression of Next Step Reactivity

$$\rho^n = a_1 p^n + b_1 \quad (31)$$

$$a_1 = f_{fp}^n \gamma_d \Delta t_n \kappa_1(\tilde{\lambda}_H^n) \quad (32)$$

$$b_1 = \rho_{im}^n + e^{-\hat{\lambda}_H^n} \rho_d^{n-1} - P_0 \gamma_d \Delta t_n \kappa_0(\hat{\lambda}_H^n) + e^{\alpha \Delta t_n} \gamma_d \Delta t_n f_{fp}^{n-1} p^{n-1} [\kappa_0(\tilde{\lambda}_H^n) - \kappa_1(\tilde{\lambda}_H^n)] \quad (33)$$

# Solution with Feedback

- Substitute (31) into (20)

$$e^{-\alpha^n \Delta t_n} p^n - p^{n-1} = \theta \Delta t_n e^{-\alpha^n \Delta t_n} \left( \left( \frac{a_1 p^n + b_1 - \beta^{eff,n}}{\Lambda^n} - \alpha^n \right) p^n + \frac{\tau^n}{\Lambda_0} p^n + \frac{1}{\Lambda_0} \hat{S}_d^n \right) + (1-\theta) \Delta t_n \left( \left( \frac{\rho^{n-1} - \beta^{eff,n-1}}{\Lambda^{n-1}} - \alpha^n \right) p^{n-1} + \frac{S_d^{n-1}}{\Lambda_0} \right) \quad (34)$$

- Rearrange eq. (34)  $a(p^n)^2 + bp^n + c = 0 \quad (35)$

$$a = \frac{\theta \Delta t_n a_1}{\Lambda^n} \quad (36)$$

$$b = \left[ \theta \Delta t_n \left( \left( \frac{b_1 - \beta^{eff,n}}{\Lambda^n} - \alpha^n \right) + \frac{\tau^n}{\Lambda_0} \right) - 1 \right] \quad (37)$$

$$c = \theta \Delta t_n \frac{1}{\Lambda_0} \hat{S}_d^n + e^{\alpha^n \Delta t_n} \left[ (1-\theta) \Delta t_n \left( \left( \frac{\rho^{n-1} - \beta^{eff,n-1}}{\Lambda^{n-1}} - \alpha^n \right) p^{n-1} + \frac{S_d^{n-1}}{\Lambda_0} \right) + p^{n-1} \right] \quad (38)$$

- Flux Level Solution  $p^n = \frac{-b - \sqrt{b^2 - 4ac}}{2a}, \text{ if } a < 0 \quad (39)$

- No feedback or explicit:  $a = 0, \text{ and } p^n = \frac{c}{-b}, \quad (40)$

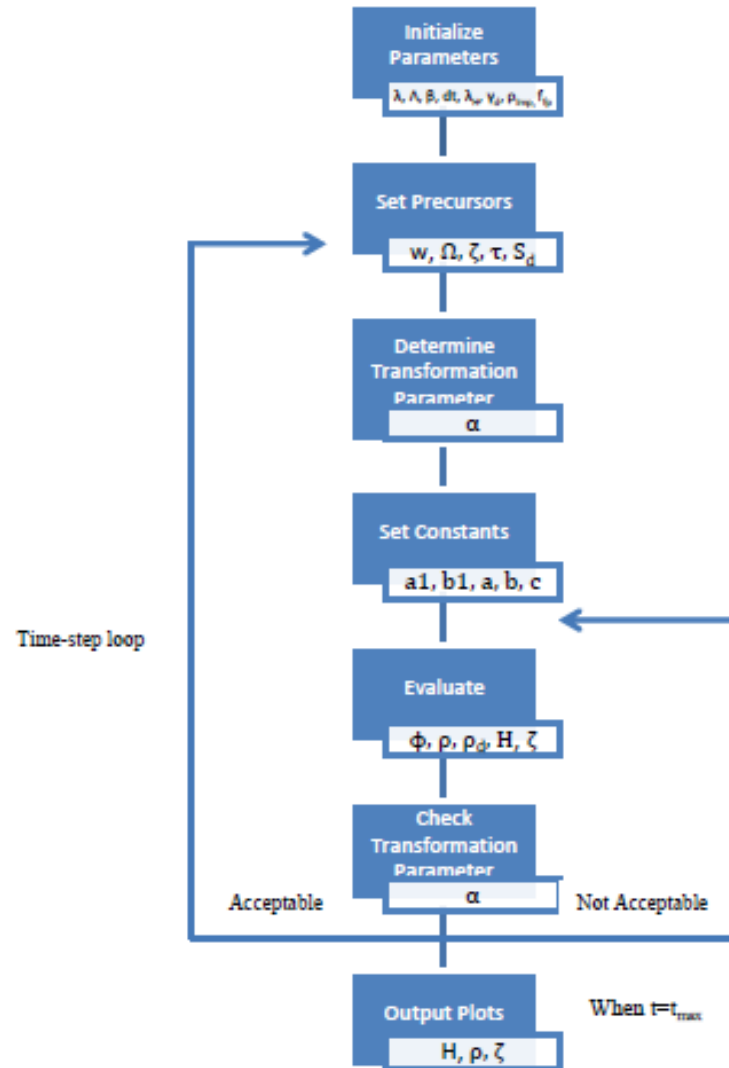


# P.K. Solution Scheme

Only formula with red numbers need to be coded, the sequence of these formula in your program should be as follows:

1. Prepare  $\Omega_k^n$  and  $\hat{\zeta}_k^n$  with eq. (15),(16)  
the exponential integration functions (10)~(13) are called
1. Prepare  $\tau^n, \hat{S}_d^n, S_d^{n-1}$  with eq. (18),(19)
2. Prepare a1,b1 with eq. (32) and (33)
3. Determine transformation parameter with eq. (23)
4. Prepare a,b,c with eq. (36), (37), and(38)
5. Evaluate flux level with eq. (39) or (40)
6. Check if exponential transformation is acceptable with eq. (24), if not, redo step 5 and 6 with  $\alpha^n = 0$ .
7. Evaluate power level, precursors and reactivity with eq. (22), (14) and (31)

# E.P.K.E. Solution Scheme



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# Kinetics options implemented in PARCS

Transient card in CNTL block

transient (tran) (kinopt)

Kinopt, integer, kinetic option, default 0

0: 3d spatial kinetics

1: point kinetics with adjoint weighted cross section change as reactivity exclude control rod reactivity component,

2: point kinetics with adjoint weighted cross section change as reactivity,

3: **Conventional** point kinetics with power weighted core average parameter feedback,

4: point kinetics with adjoint fission weighted core average parameter feedback,

5: point kinetics without feedback (user input reactivity),

6: no kinetics,(user input power levels)

Impose reactivity are acceptable for option 1 to 5

# Impose Reactivity for point kinetics

✶ IMPOSERHO cards in TRAN block

imposerho    t1   r1   t2   r2 ...

t<sub>i</sub>: real, time point for impose reactivity

r<sub>i</sub>: real, impose reactivity ( or power level if kinopt=6)

Example:

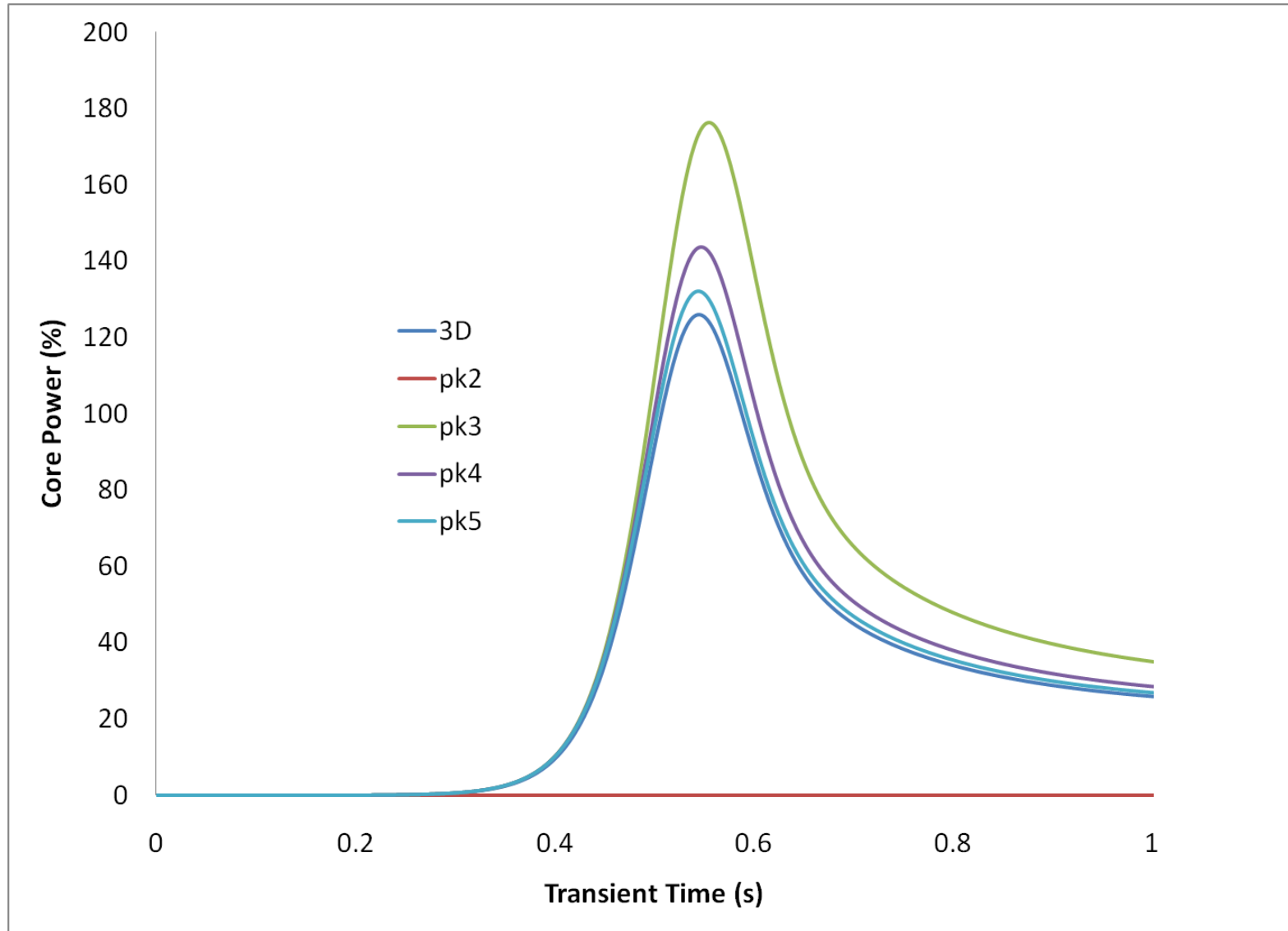
imposerho	0.000000E+00	0.000000E+00
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imposerho	1.000000E-03	2.532442E-04
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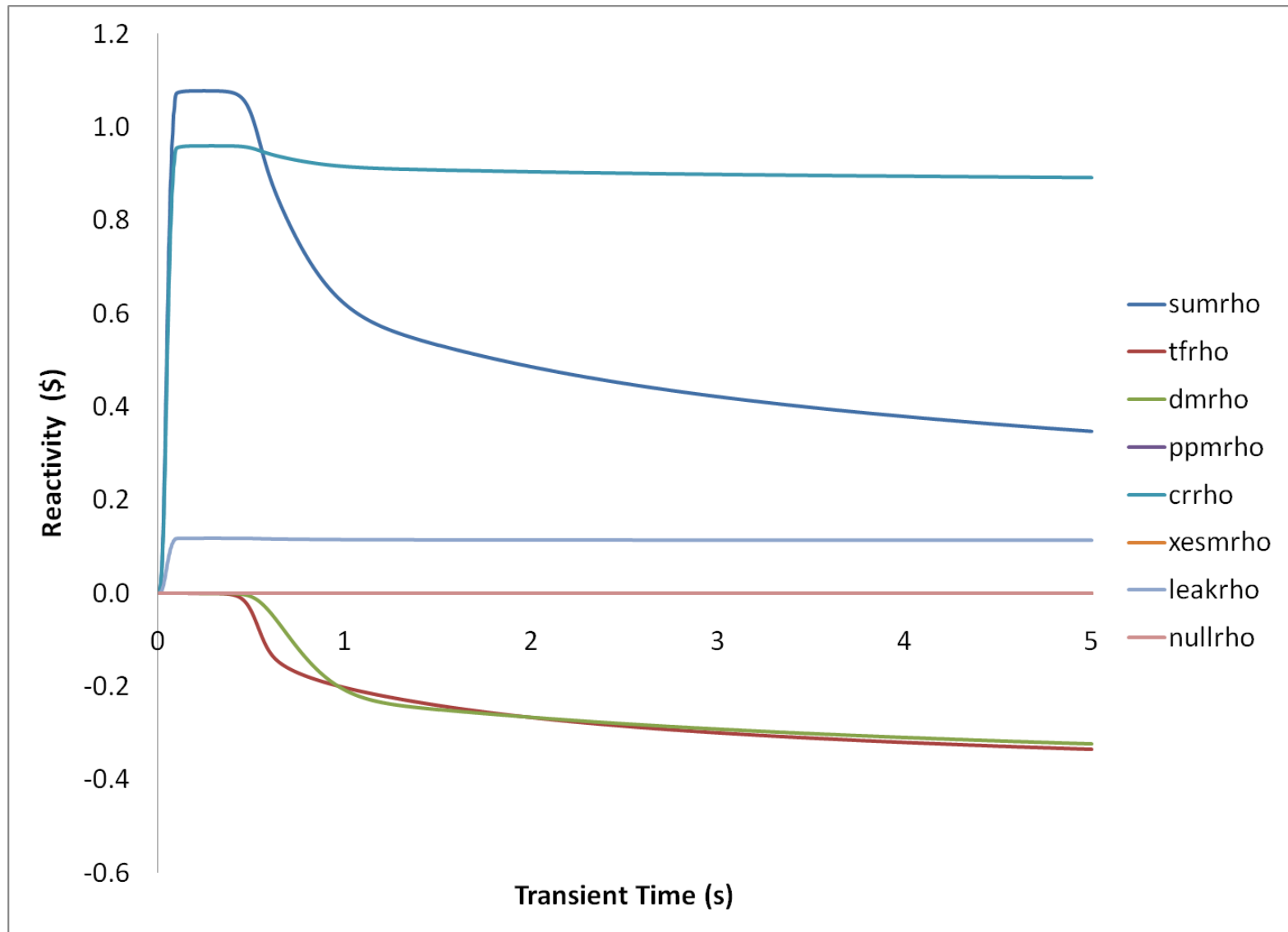
imposerho	2.000000E-03	5.697343E-04
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imposerho	3.000000E-03	9.885548E-04
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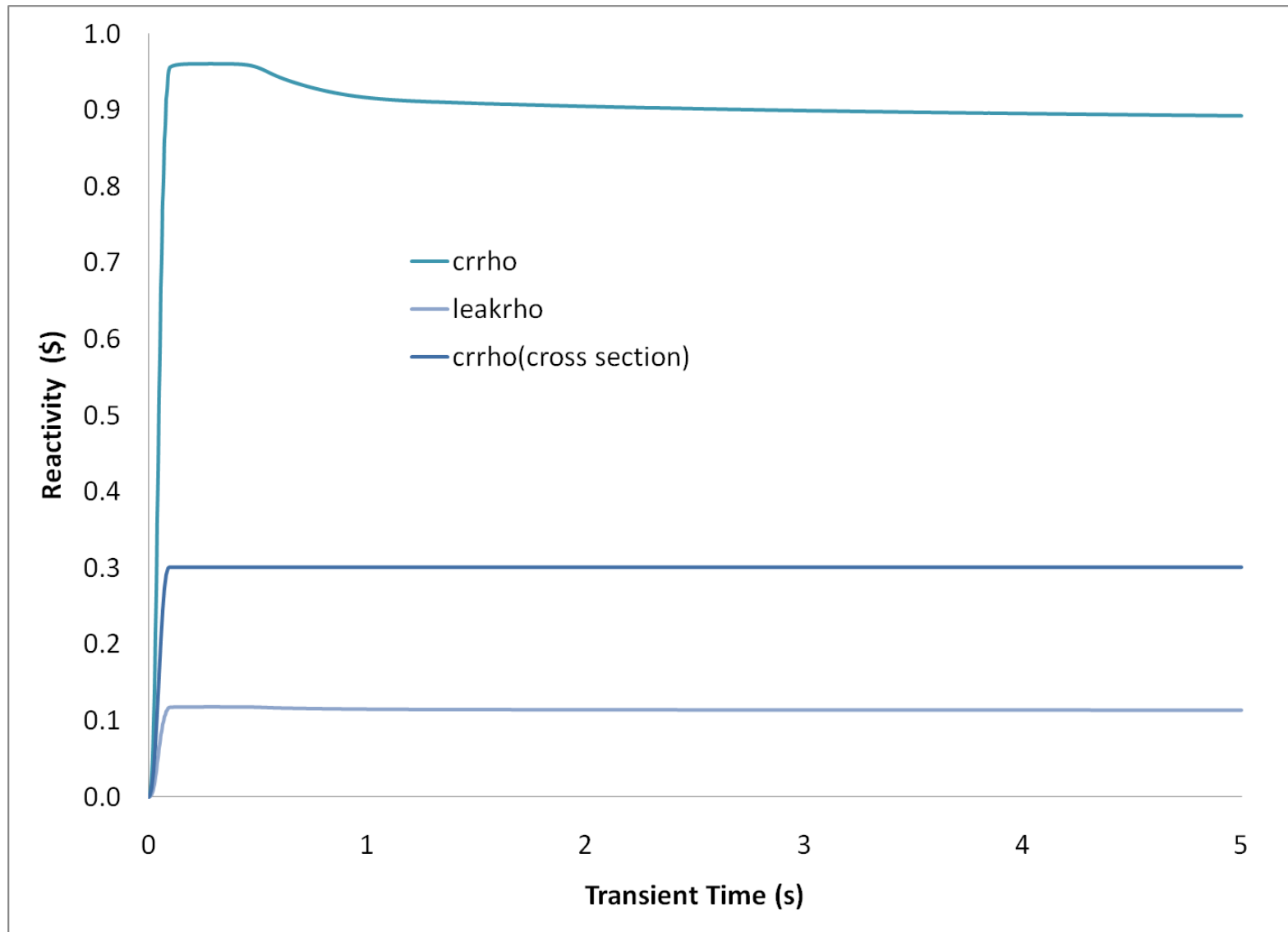
# Point kinetic solutions for NEACRP-A1



# Reactivity Components from spatial kinetics

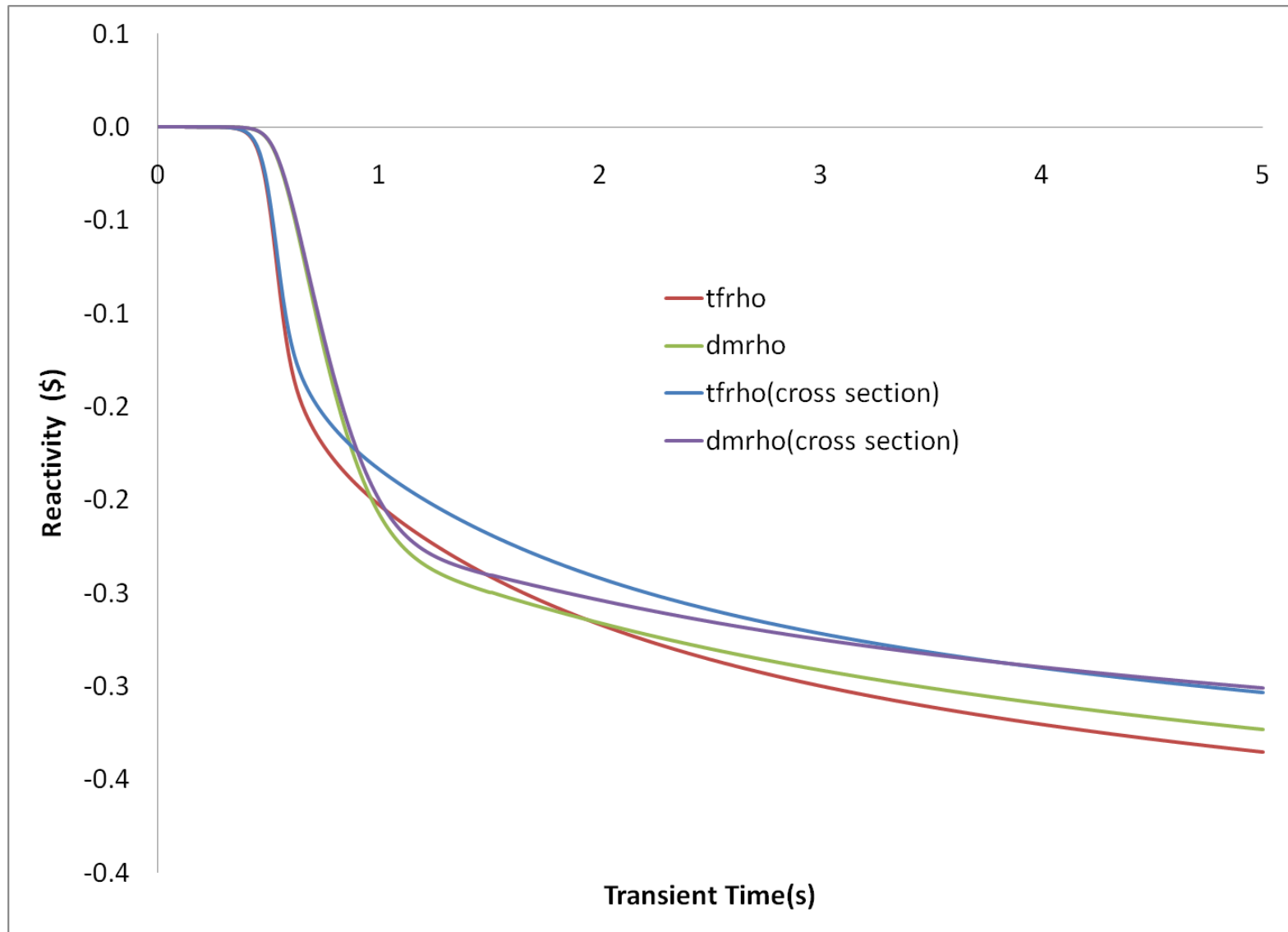


# Control Rod Reactivity from Cross Section with initial flux shape

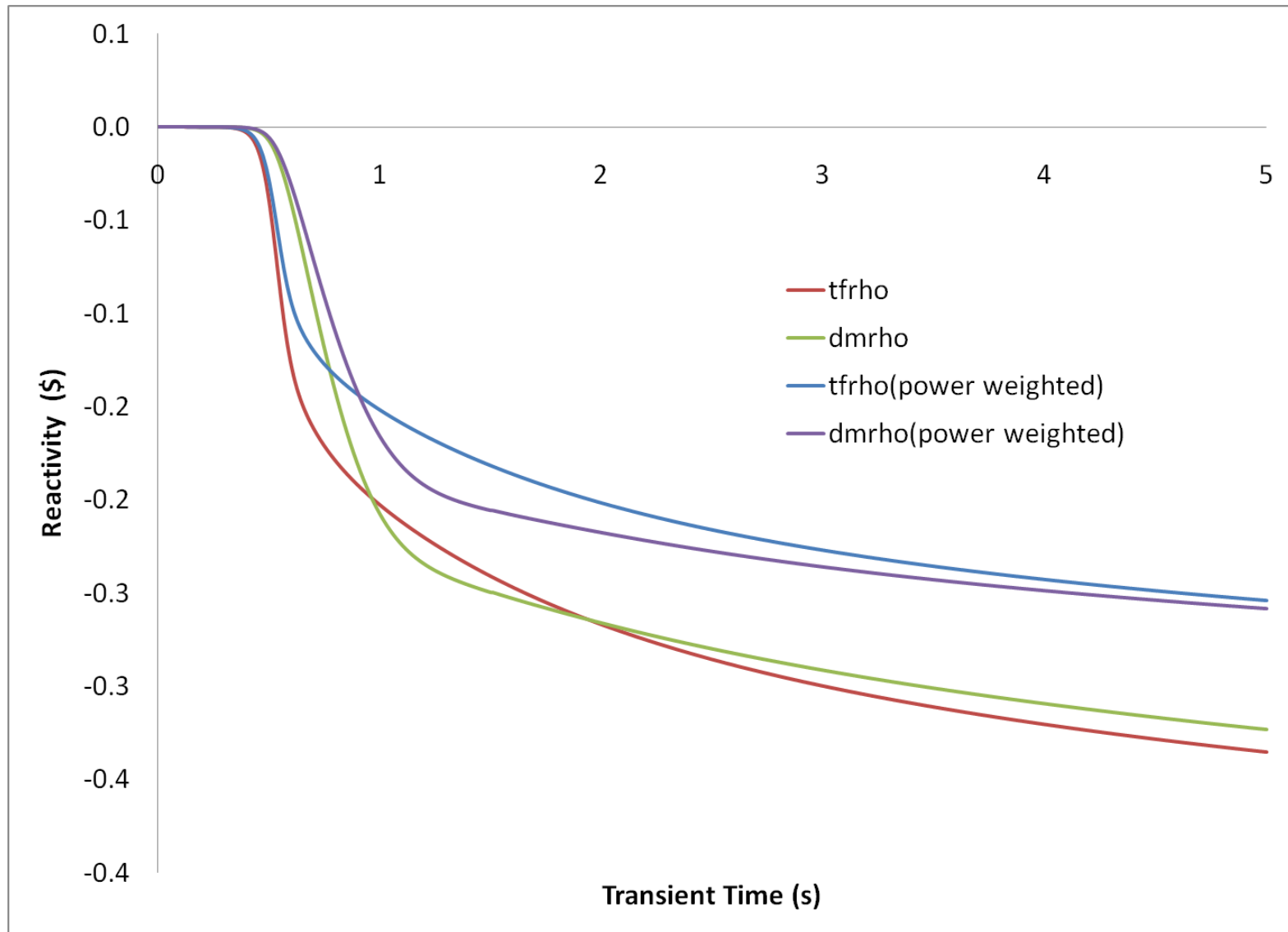




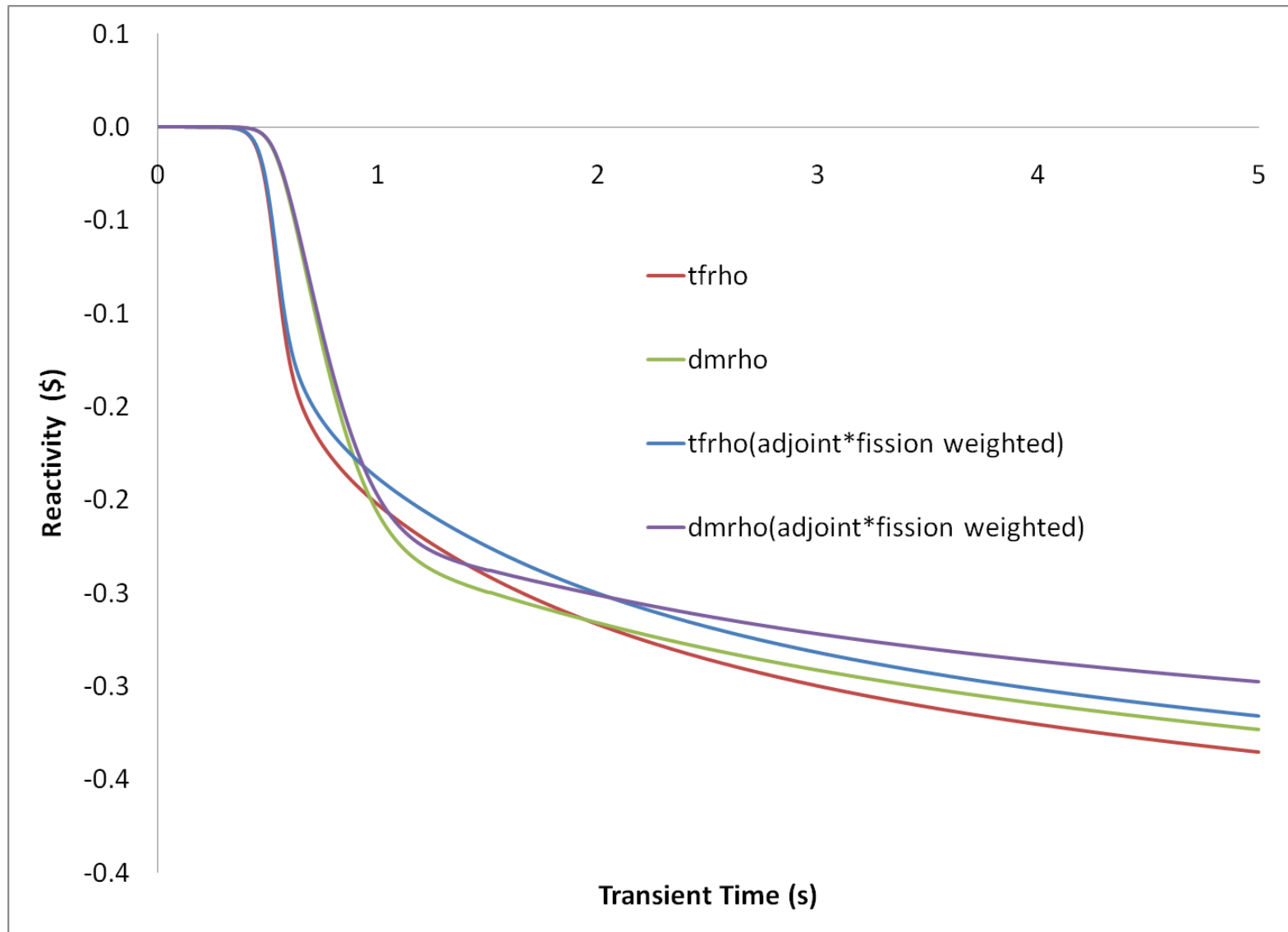
# Feed back Reactivity from Cross with initial flux shape



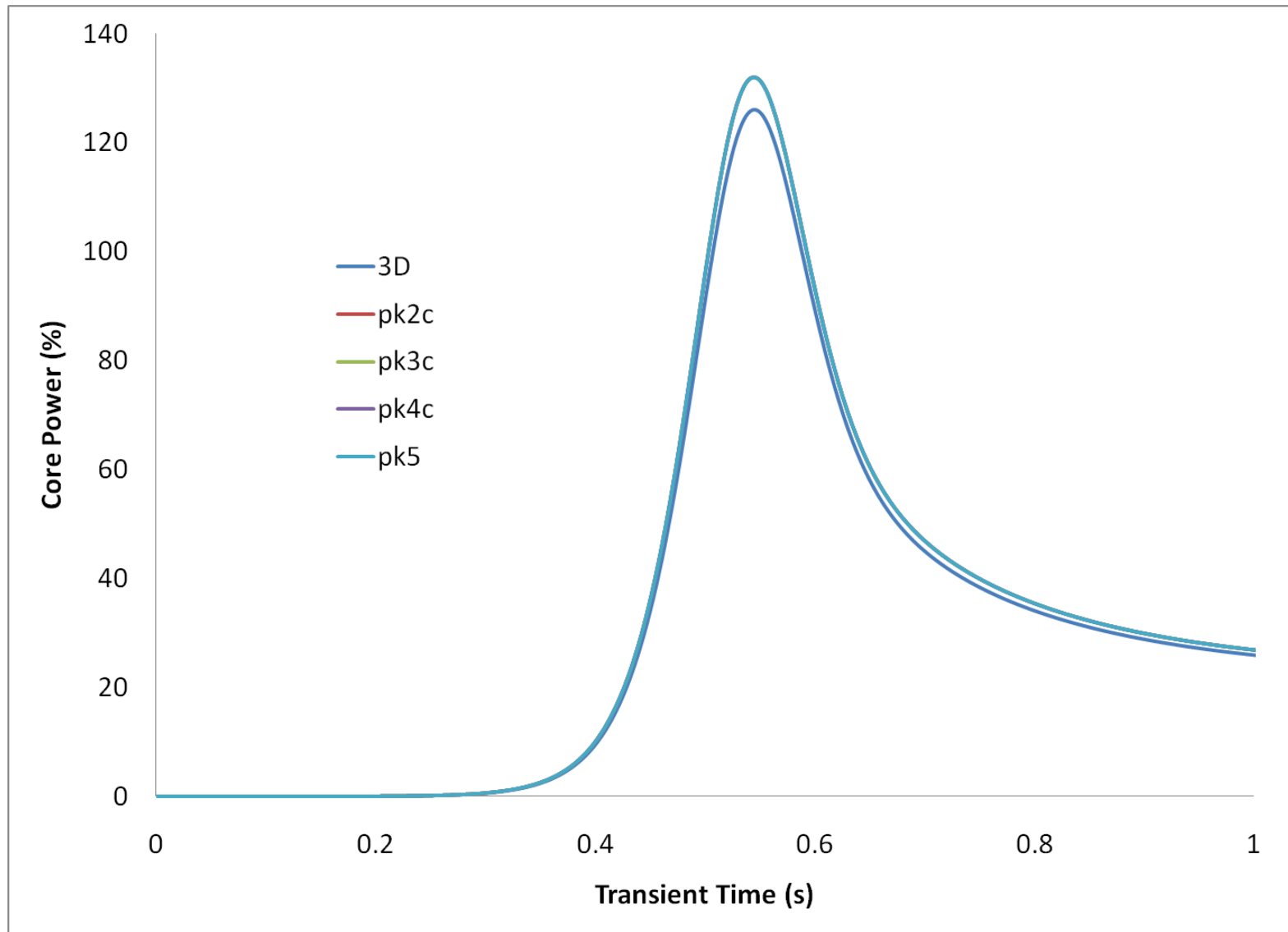
# Feed back Reactivity from power weighted core average parameters



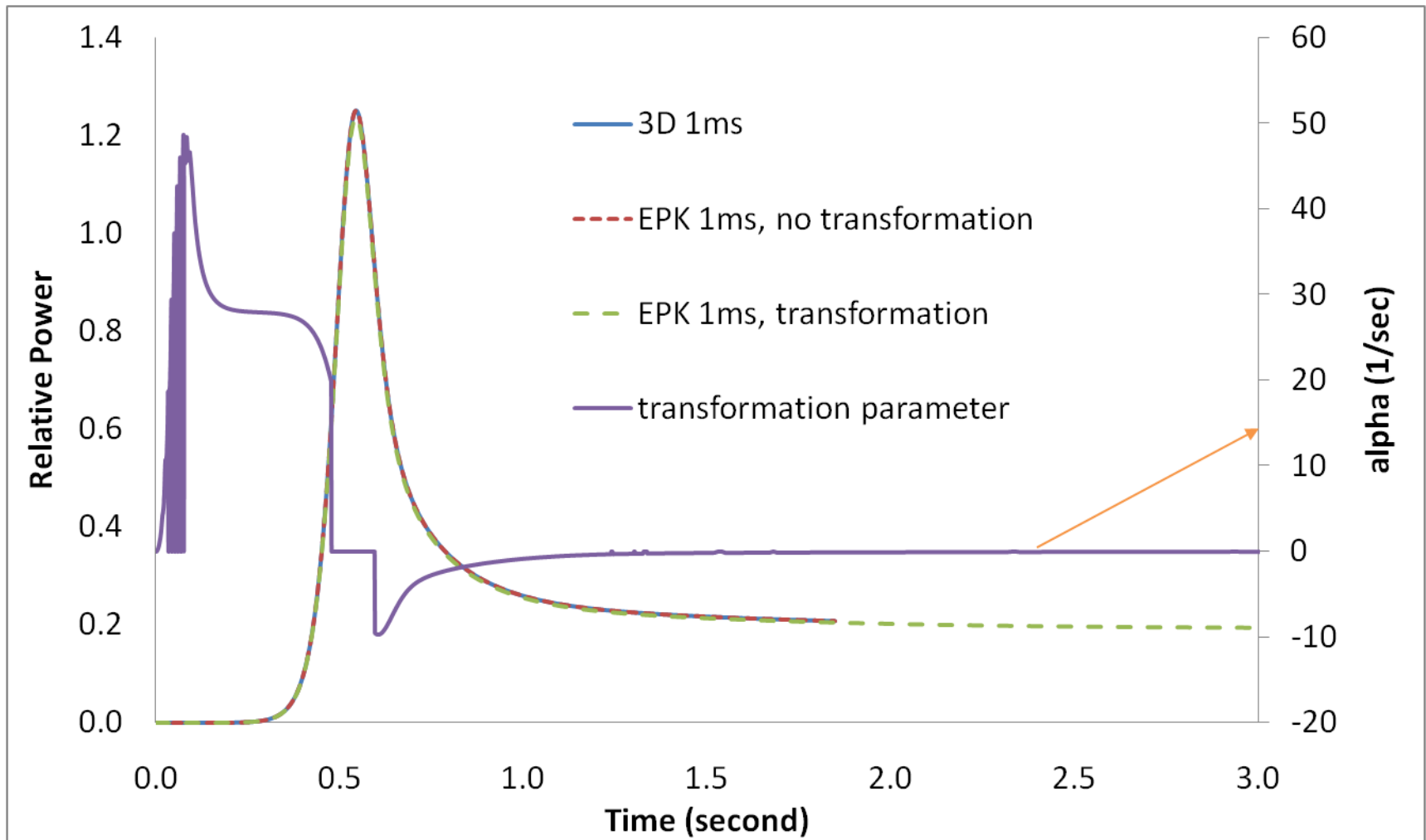
## Feed back Reactivity from Adjoint\*Fission weighted core average parameters



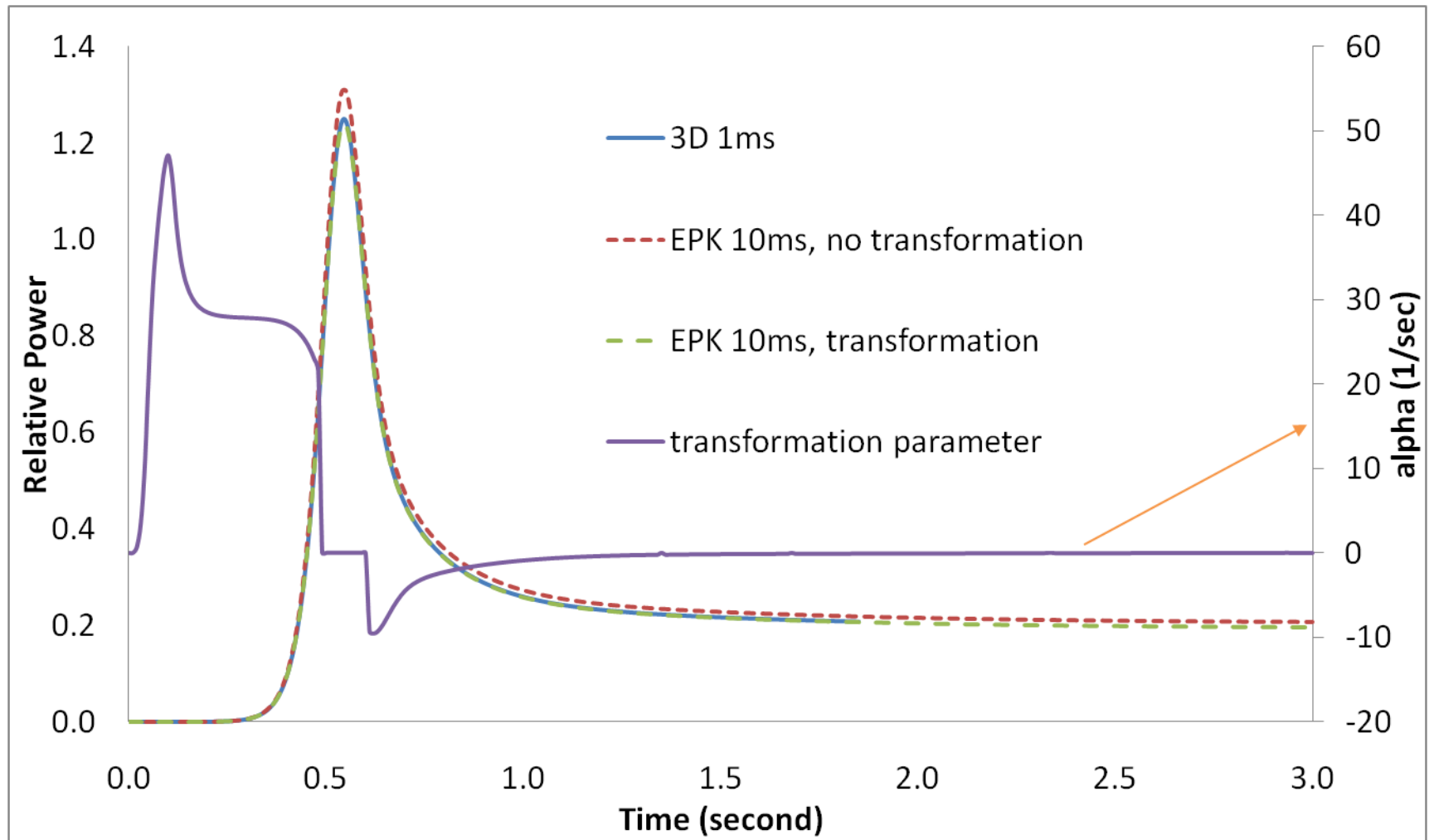
# Point kinetic solutions with corrected imposed reactivity



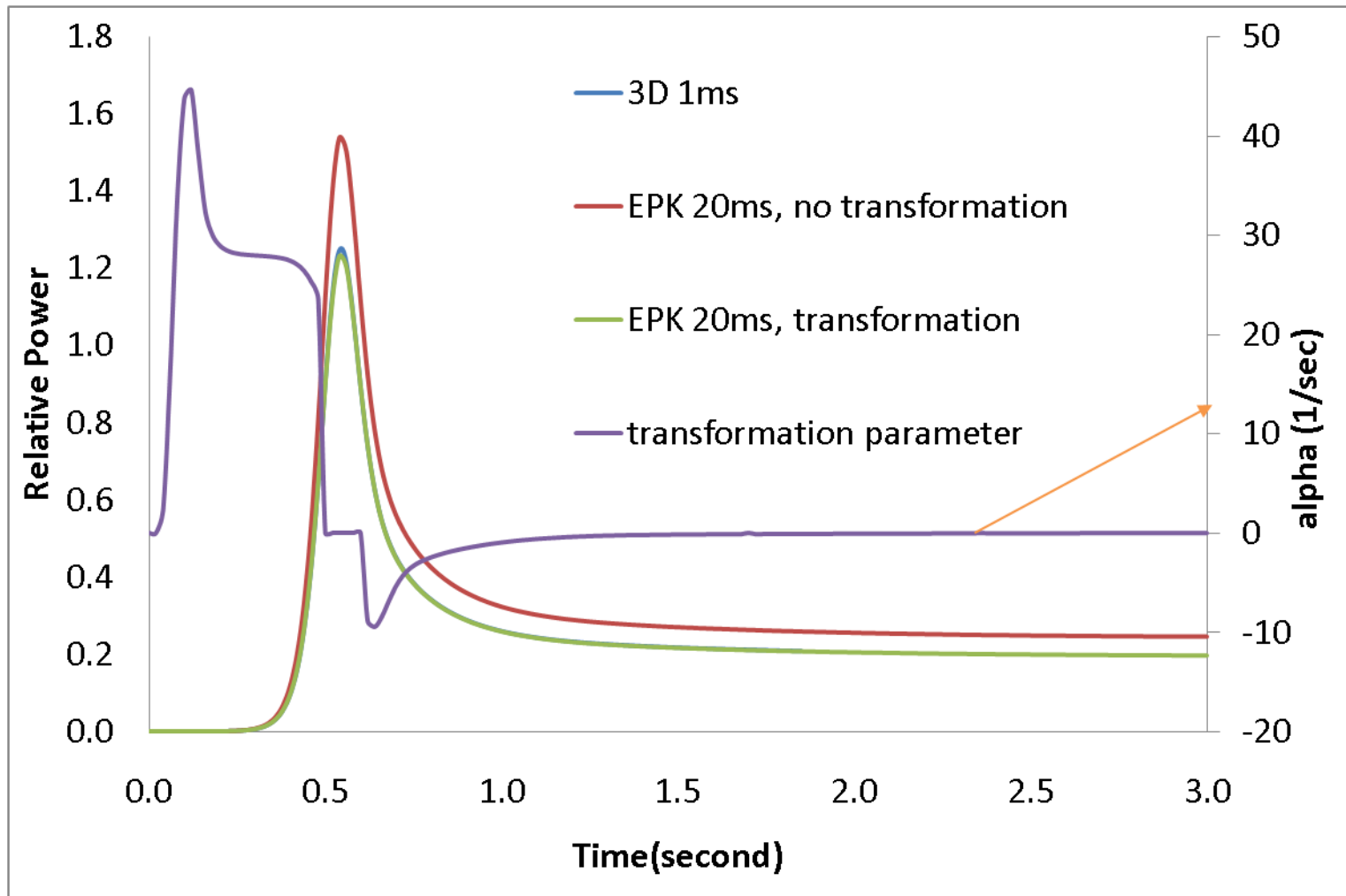
# Analysis of Exponential Transportation: EPK solution for NEACRP-A1 with 1ms steps



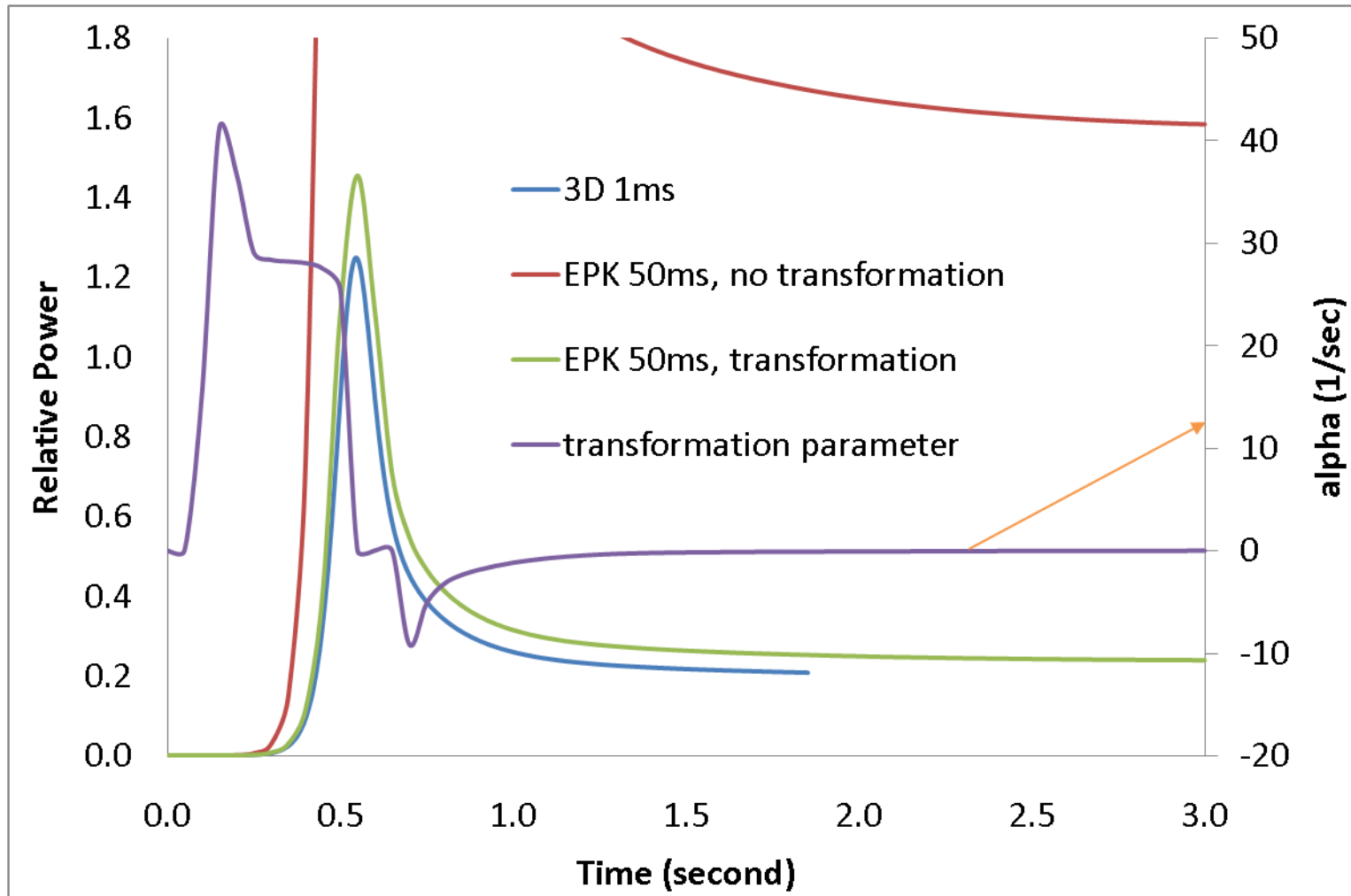
# EPK solution for NEACRP-A1 with 10ms steps



# EPK solution for NEACRP-A1 with 20ms steps



## EPK solution for NEACRP-A1 with 50ms steps





# Summary of Applications

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- Exact point kinetics can reproduce exactly the same power solution from the spatial kinetics solution with the pre-computed reactivity and time dependent kinetics parameters.
- Point kinetics with pre-computed reactivity and only the initial kinetics parameters predicts power solution slightly different from spatial kinetics solution.
- There is a large error in control rod reactivity using only the initial flux shape.
- There are errors in the feed back reactivity evaluated with cross section or core average parameters. The reactivity evaluated with power weighted core average parameters has the largest error among the three methods.

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# Backup Slides

# Spatial Kinetics Equations

- Transient Neutron Diffusion Equation

$$\frac{1}{\nu_g(\mathbf{r})} \frac{\partial \phi_g(\mathbf{r}, t)}{\partial t} = -\nabla \cdot \mathbf{J}_g(\mathbf{r}, t) + \sum_{g'} \Sigma_{g', g}(\mathbf{r}, t) \phi_{g'}(\mathbf{r}, t) - \Sigma_{tg}(\mathbf{r}, t) \phi_g(\mathbf{r}, t) + \chi_g(\mathbf{r}) S^F(\mathbf{r}, t) \\ + \sum_k \chi_{dk, g}(\mathbf{r}) (\lambda_k(\mathbf{r}) C_k(\mathbf{r}, t) - \beta_k(\mathbf{r}) S^F(\mathbf{r}, t))$$

$$S^F(\mathbf{r}, t) = \frac{1}{k_{eff}^s} \sum_{g'} \nu \Sigma_{f, g'}(\mathbf{r}, t) \phi_{g'}(\mathbf{r}, t), \quad \mathbf{J}_g(\mathbf{r}, t) = -D_g(\mathbf{r}, t) \nabla \phi_g(\mathbf{r}, t)$$

$$\chi_g = \chi_{p, g} + \sum_k \beta_k (\chi_{dk, g} - \chi_{p, g}),$$

- Precursor equations

$$\frac{\partial C_k(\mathbf{r}, t)}{\partial t} = \beta_k(\mathbf{r}) S^F(\mathbf{r}, t) - \lambda_k(\mathbf{r}) C_k(\mathbf{r}, t), \quad k = 1, 2, \dots$$

# Magnitude and Form Functions

- Magnitude and Form Functions

$$\phi_g(\mathbf{r}, t) = p(t) \psi_g(\mathbf{r}, t),$$

- Normalization of Form Function

$$\langle \phi_g^*(\mathbf{r}) \frac{1}{v_g(\mathbf{r})} \psi_g(\mathbf{r}, t) \rangle = \langle \phi_g^*(\mathbf{r}) \frac{1}{v_g(\mathbf{r})} \psi_g(\mathbf{r}, 0) \rangle$$

Adjoint flux

- Power Form Factor

$$f_{fp}(t) = \frac{\langle \kappa \Sigma_{fg}(r, t) \phi_g(r, t) \rangle p(0)}{p(t) \langle \kappa \Sigma_{fg}(r, 0) \phi_g(r, 0) \rangle} = \frac{\langle \kappa \Sigma_{fg}(r, t) \psi_g(r, t) \rangle}{\langle \kappa \Sigma_{fg}(r, 0) \psi_g(r, 0) \rangle}, \quad f_{fp}(0) = 1$$

- Fission Power Level

$$H(t) = f_{fp}(t) p(t)$$

$$\langle \rangle \equiv \sum_g \iiint_v dv \quad : \text{Integration notation}$$

- Alternative Normalization?

$$\langle \kappa \Sigma_{fg}(r, t) \psi_g(\mathbf{r}, t) \rangle = \langle \kappa \Sigma_{fg}(r, 0) \psi_g(\mathbf{r}, 0) \rangle$$

# Integrate Flux Equation with Adjoint weighting

$$\begin{aligned} \langle \phi_g^*(\mathbf{r}) \frac{1}{v_g} \frac{\partial \phi_g(\mathbf{r}, t)}{\partial t} \rangle = & \langle \phi_g^*(\mathbf{r}) \left( -\nabla \cdot \mathbf{J}_g(\mathbf{r}, t) + \sum_{g'} \Sigma_{g', g}(\mathbf{r}, t) \phi_{g'}(\mathbf{r}, t) - \Sigma_{tg}(\mathbf{r}, t) \phi_g(\mathbf{r}, t) + \chi_g(\mathbf{r}) S^F(\mathbf{r}, t) \right) \rangle \\ & + \langle \phi_g^*(\mathbf{r}) \sum_k \chi_{dk, g}(\mathbf{r}) [\lambda_k(\mathbf{r}) C_k(\mathbf{r}, t) - \beta_k(\mathbf{r}) S^F(\mathbf{r}, t)] \rangle \end{aligned}$$

- Time Derivative Term

$$\langle \phi_g^*(\mathbf{r}) \frac{1}{v_g(\mathbf{r})} \phi_g(\mathbf{r}, t) \rangle = p(t) \langle \phi_g^*(\mathbf{r}) \frac{1}{v_g(\mathbf{r})} \psi_g(\mathbf{r}, t) \rangle = p(t) \langle \phi_g^*(\mathbf{r}) \frac{1}{v_g(\mathbf{r})} \psi_g(\mathbf{r}, 0) \rangle$$

- Generation Time

$$\Lambda(t) = \frac{\langle \phi_g^*(\mathbf{r}) \frac{1}{v_g(\mathbf{r})} \psi_g(\mathbf{r}, t) \rangle}{F(t)}$$

- Adjoint weighted quasi-stationary fission source

$$F(t) = \langle \phi_g^*(\mathbf{r}) \chi_g(\mathbf{r}) S^F(\mathbf{r}, t) \rangle / p(t) = \langle \phi_g^*(\mathbf{r}) \chi_g(\mathbf{r}) \hat{S}^F(\mathbf{r}, t) \rangle, \quad \hat{S}^F(\mathbf{r}, t) = \frac{1}{k_{eff}^s} \sum_{g'} v \Sigma_{f, g'}(\mathbf{r}, t) \psi_{g'}(\mathbf{r}, t),$$

- Time invariance

$$\Lambda(t) F(t) = \Lambda(0) F(0) = \Lambda_0 F_0$$

$$\langle \phi_g^*(\mathbf{r}) \frac{1}{v_g} \frac{\partial \phi_g(\mathbf{r}, t)}{\partial t} \rangle = \frac{\partial}{\partial t} \langle \phi_g^*(\mathbf{r}) \frac{1}{v_g} \phi_g(\mathbf{r}, t) \rangle = \frac{\partial}{\partial t} \left( p(t) \langle \phi_g^*(\mathbf{r}) \frac{1}{v_g(\mathbf{r})} \psi_g(\mathbf{r}, 0) \rangle \right) = \Lambda_0 F_0 \frac{dp(t)}{dt}$$

# Integrate Flux Equation (cont.)

$$\begin{aligned} \langle \phi_g^*(\mathbf{r}) \frac{1}{v_g} \frac{\partial \phi_g(\mathbf{r}, t)}{\partial t} \rangle = & \langle \phi_g^*(\mathbf{r}) \left( -\nabla \cdot \mathbf{J}_g(\mathbf{r}, t) + \sum_{g'} \Sigma_{g, g'}(\mathbf{r}, t) \phi_{g'}(\mathbf{r}, t) - \Sigma_{tg}(\mathbf{r}, t) \phi_g(\mathbf{r}, t) + \chi_g(\mathbf{r}) S^F(\mathbf{r}, t) \right) \rangle \\ & + \langle \phi_g^*(\mathbf{r}) \sum_k \chi_{dk, g}(\mathbf{r}) [\lambda_k(\mathbf{r}) C_k(\mathbf{r}, t) - \beta_k(\mathbf{r}) S^F(\mathbf{r}, t)] \rangle \end{aligned}$$

- Reactivity Term

$$\rho(t) = \frac{\langle \phi_g^*(\mathbf{r}) \left( -\nabla \cdot \mathbf{J}_g(\mathbf{r}, t) + \sum_{g'} \Sigma_{g, g'}(\mathbf{r}, t) \phi_{g'}(\mathbf{r}, t) - \Sigma_{tg}(\mathbf{r}, t) \phi_g(\mathbf{r}, t) + \chi_g(\mathbf{r}) S^F(\mathbf{r}, t) \right) \rangle}{p(t) F(t)}$$

$$\rho(t) = \frac{1}{p(t) F(t)} \langle \phi^*(r) (F - M) \phi(r, t) \rangle$$

# Integrate Flux Equation (cont.)

$$\begin{aligned} \langle \phi_g^*(\mathbf{r}) \frac{1}{v_g} \frac{\partial \phi_g(\mathbf{r}, t)}{\partial t} \rangle = & \langle \phi_g^*(\mathbf{r}) \left( -\nabla \cdot \mathbf{J}_g(\mathbf{r}, t) + \sum_{g'} \Sigma_{g', g}(\mathbf{r}, t) \phi_{g'}(\mathbf{r}, t) - \Sigma_{tg}(\mathbf{r}, t) \phi_g(\mathbf{r}, t) + \chi_g(\mathbf{r}) S^F(\mathbf{r}, t) \right) \rangle \\ & + \langle \phi_g^*(\mathbf{r}) \sum_k \chi_{dk, g}(\mathbf{r}) [\lambda_k(\mathbf{r}) C_k(\mathbf{r}, t) - \beta_k(\mathbf{r}) S^F(\mathbf{r}, t)] \rangle \end{aligned}$$

- Beta Terms

$$\beta_k^{\text{eff}}(t) = \frac{1}{p(t)F(t)} \langle \phi_g^*(\mathbf{r}) \chi_{dk, g}(\mathbf{r}) \beta_k(\mathbf{r}) S^F(\mathbf{r}, t) \rangle, \quad \beta^{\text{eff}}(t) = \sum_k \beta_k^{\text{eff}}(t)$$

$$\langle \phi_g^*(\mathbf{r}) \sum_k \chi_{dk, g}(\mathbf{r}) \beta_k(\mathbf{r}) S^F(\mathbf{r}, t) \rangle = p(t)F(t) \sum_k \beta_k^{\text{eff}}(t) = p(t)F(t) \beta^{\text{eff}}(t)$$

- Delay neutron Terms

$$\lambda_k(t) = \frac{\langle \phi_g^*(\mathbf{r}) \lambda_k(\mathbf{r}) \chi_{dk, g}(\mathbf{r}) C_k(\mathbf{r}, t) \rangle}{\langle \phi_g^*(\mathbf{r}) \chi_{dk, g}(\mathbf{r}) C_k(\mathbf{r}, t) \rangle}, \quad \zeta_k(t) = \frac{\langle \phi_g^*(\mathbf{r}) \chi_{dk, g}(\mathbf{r}) C_k(\mathbf{r}, t) \rangle}{F_0},$$

$$\langle \phi_g^*(\mathbf{r}) \sum_k \chi_{dk, g}(\mathbf{r}) \lambda_k(\mathbf{r}) C_k(\mathbf{r}, t) \rangle = \sum_k \lambda_k(t) \zeta_k(t) F_0$$

# Integrate Flux Equation (cont.)

$$\begin{aligned} \langle \phi_g^*(\mathbf{r}) \frac{1}{v_g} \frac{\partial \phi_g(\mathbf{r}, t)}{\partial t} \rangle = & \langle \phi_g^*(\mathbf{r}) \left( -\nabla \cdot \mathbf{J}_g(\mathbf{r}, t) + \sum_{g'} \Sigma_{g', g'}(\mathbf{r}, t) \phi_{g'}(\mathbf{r}, t) - \Sigma_{tg}(\mathbf{r}, t) \phi_g(\mathbf{r}, t) + \chi_g(\mathbf{r}) S^F(\mathbf{r}, t) \right) \rangle \\ & + \langle \phi_g^*(\mathbf{r}) \sum_k \chi_{dk, g}(\mathbf{r}) [\lambda_k(\mathbf{r}) C_k(\mathbf{r}, t) - \beta_k(\mathbf{r}) S^F(\mathbf{r}, t)] \rangle \end{aligned}$$

- Put All Terms Together  $\Lambda_0 F_0 \frac{dp(t)}{dt} = \rho(t) p(t) F(t) - \beta^{\text{eff}}(t) p(t) F(t) + \sum_k \lambda_k(t) \zeta_k(t) F_0$

$$\frac{dp(t)}{dt} = \frac{\rho(t) - \beta^{\text{eff}}(t)}{\Lambda(t)} p(t) + \frac{1}{\Lambda_0} \sum_k \lambda_k(t) \zeta_k(t)$$

- Why adjoint weighting? Impact on reactivity by form function error

$$\begin{aligned} \frac{1}{F(t)} \langle \phi^*(r) (\bar{\bar{F}} - \bar{\bar{M}}) (\psi(r, t) + \delta\psi(r, t)) \rangle &= -\frac{1}{F(t)} \langle \phi^*(r) (\bar{\bar{F}} - \bar{\bar{M}}) \psi(r, t) \rangle \\ &= \frac{1}{F(t)} \langle \phi^*(r) (\bar{\bar{F}} - \bar{\bar{M}}) \delta\psi(r, t) \rangle = \frac{1}{F(t)} \langle \phi^*(r) (\bar{\bar{F}}_0 + \Delta\bar{\bar{F}} - \bar{\bar{M}}_0 - \Delta\bar{\bar{M}}) \delta\psi(r, t) \rangle \\ &= \frac{1}{F(t)} \left( \langle \phi^*(r) (\bar{\bar{F}}_0 - \bar{\bar{M}}_0) \delta\psi(r, t) \rangle + \langle \phi^*(r) (\Delta\bar{\bar{F}} - \Delta\bar{\bar{M}}) \delta\psi(r, t) \rangle \right) \end{aligned}$$

- Select weight function so that:

$$\langle \phi^*(r) (\bar{\bar{F}}_0 - \bar{\bar{M}}_0) \delta\psi(r, t) \rangle \equiv 0 \quad \text{for any } \delta\psi(r, t) \Rightarrow \left( \bar{\bar{F}}_0^* - \bar{\bar{M}}_0^* \right) \phi^*(r) = 0$$



# Integrate Precursor Eqns with Adjoint weighting

$$\langle \phi_g^*(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) \frac{\partial C_k(\mathbf{r},t)}{\partial t} \rangle = \langle \phi_g^*(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) \beta_k(\mathbf{r}) S^F(\mathbf{r},t) \rangle - \langle \phi_g^*(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) \lambda_k(\mathbf{r}) C_k(\mathbf{r},t) \rangle$$

- Time Derivative Term  $\zeta_k(t) = \frac{\langle \phi_g^*(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) C_k(\mathbf{r},t) \rangle}{F_0},$

$$\langle \phi_g^*(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) \frac{\partial C_k(\mathbf{r},t)}{\partial t} \rangle = \frac{\partial}{\partial t} \langle \phi_g^*(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) C_k(\mathbf{r},t) \rangle = \frac{d}{dt} (\zeta_k(t) F_0) = F_0 \frac{d}{dt} \zeta_k(t)$$

- Source term  $\beta_k^{\text{eff}}(t) = \frac{\langle \phi_g^*(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) \beta_k(\mathbf{r}) S^F(\mathbf{r},t) \rangle}{p(t)F(t)},$

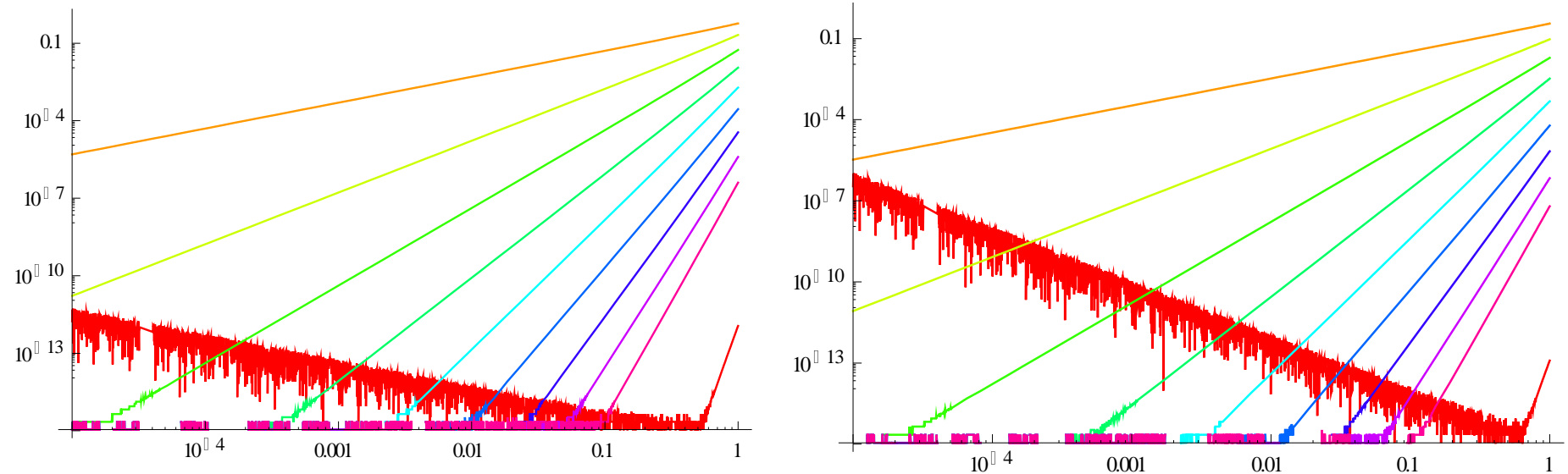
$$\langle \phi_g^*(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) \beta_k(\mathbf{r}) S^F(\mathbf{r},t) \rangle = \beta_k^{\text{eff}}(t) p(t) F(t)$$

- Decay term  $\langle \phi_g^*(\mathbf{r}) \chi_{dk,g}(\mathbf{r}) \lambda_k(\mathbf{r}) C_k(\mathbf{r},t) \rangle = \lambda_k(t) \zeta_k(t) F_0$

- Put All Terms Together  $F_0 \frac{d\zeta_k(t)}{dt} = \beta_k^{\text{eff}}(t) p(t) F(t) - \lambda_k(t) \zeta_k(t) F_0$

$$\frac{d\zeta_k(t)}{dt} = \frac{\Lambda_0}{\Lambda(t)} \beta_k^{\text{eff}}(t) p(t) - \lambda_k(t) \zeta_k(t)$$

# Numerical error of evaluating Kappa0 and Kappa1



Differences of 0-8<sup>th</sup> order Taylor expansions and direct evaluation of Kappa0 and Kappa1 from 13<sup>th</sup> order Taylor expansions value. The Taylor expansions functions have significant truncation error for large input. On the other hand, for small input, there is large round off error in numerical evaluation of kappa2 due to subtract two close numbers.