Surface Reaction Model Uncertainty Quantification Intrusive and Non-Intrusive Methods

UQTk Example

3 ODEs for a monomer (u), dimer (v), and inert species (w) adsorbing onto a surface out of gas phase.

$$\frac{du}{dt} = az - cu - 4duv$$

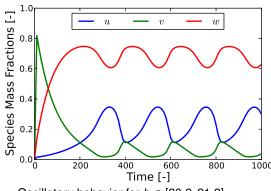
$$\frac{dv}{dt} = 2bz^2 - 4duv$$

$$\frac{dw}{dt} = ez - fw$$

$$z = 1 - u - v - w$$

$$u(0) = v(0) = w(0) = 0.0$$

$$a = 1.6$$
 $b = 20.75$ $c = 0.04$



Oscillatory behavior for $b \in [20.2, 21.2]$

$$a = 1.6$$
 $b = 20.75$ $c = 0.04$ $d = 1.0$ $e = 0.36$ $f = 0.016$

(Vigil et al., Phys. Rev. E., 1996; Makeev et al., J. Chem. Phys., 2002)

Surface Reaction Model: Intrusive Spectral Propagation (ISP) of Uncertainty

- Assume PCE for uncertain parameter b and for the output variables, u, v, w
- Substitute PCEs into the governing equations
- Project the governing equations onto the PC basis functions
 - Multiply with Ψ_k and take the expectation
- Apply pseudo-spectral approximations where necessary
- UQTk elementary operations

Surface Reaction Model: Specify PCEs for inputs and outputs

Represent uncertain inputs with PCEs with known coefficients:

$$b=\sum_{i=0}^P b_i\Psi_i(\xi)$$

Represent all uncertain variables with PCEs with unknown coefficients:

$$u = \sum_{i=0}^{P} u_i \Psi_i(\xi) \quad v = \sum_{i=0}^{P} v_i \Psi_i(\xi) \quad w = \sum_{i=0}^{P} w_i \Psi_i(\xi) \quad z = \sum_{i=0}^{P} z_i \Psi_i(\xi)$$

Surface Reaction Model: Substitute PCEs into governing equations and project onto basis functions

$$\frac{\mathrm{d}u}{\mathrm{d}t} = az - cu - 4duv$$

$$\frac{\mathrm{d}}{\mathrm{d}t} \sum_{i=0}^{P} u_i \Psi_i = a \sum_{i=0}^{P} z_i \Psi_i - c \sum_{i=0}^{P} u_i \Psi_i - 4d \sum_{i=0}^{P} u_i \Psi_i \sum_{j=0}^{P} v_j \Psi_j$$

$$\left\langle \Psi_k \frac{\mathrm{d}}{\mathrm{d}t} \sum_{i=0}^{P} u_i \Psi_i \right\rangle = \left\langle a \Psi_k \sum_{i=0}^{P} z_i \Psi_i \right\rangle - \left\langle c \Psi_k \sum_{i=0}^{P} u_i \Psi_i \right\rangle$$

$$- \left\langle 4d \Psi_k \sum_{i=0}^{P} u_i \Psi_i \sum_{i=0}^{P} v_j \Psi_j \right\rangle$$

$$\frac{\mathrm{d}}{\mathrm{d}t}u_{k}\left\langle \Psi_{k}^{2}\right\rangle = az_{k}\left\langle \Psi_{k}^{2}\right\rangle - cu_{k}\left\langle \Psi_{k}^{2}\right\rangle - 4d\sum_{i=0}^{P}\sum_{j=0}^{P}u_{i}v_{j}\left\langle \Psi_{i}\Psi_{j}\Psi_{k}\right\rangle$$

$$\frac{\mathrm{d}}{\mathrm{d}t}u_{k} = az_{k} - cu_{k} - 4d\sum_{i=0}^{P}\sum_{j=0}^{P}u_{i}v_{j}\frac{\left\langle \Psi_{i}\Psi_{j}\Psi_{k}\right\rangle}{\left\langle \Psi_{k}^{2}\right\rangle}$$

$$\frac{\mathrm{d}}{\mathrm{d}t}u_{k} = az_{k} - cu_{k} - 4d\sum_{i=0}^{P}\sum_{j=0}^{P}u_{i}v_{j}C_{ijk}$$

• Triple products $C_{ijk}=rac{\langle \psi_i \psi_j \psi_k
angle}{\langle \psi_k^2
angle}$ can be pre-computed and stored for repeated use

Surface Reaction Model: Substitute PCEs into governing equations and project onto basis functions

$$\frac{\mathrm{d}v}{\mathrm{d}t} = 2bz^{2} - 4duv$$

$$\frac{\mathrm{d}}{\mathrm{d}t} \sum_{i=0}^{P} v_{i} \Psi_{i} = 2\sum_{h=0}^{P} b_{h} \Psi_{h} \sum_{i=0}^{P} z_{i} \Psi_{i} \sum_{j=0}^{P} z_{j} \Psi_{j} - 4d\sum_{i=0}^{P} u_{i} \Psi_{i} \sum_{j=0}^{P} v_{j} \Psi_{j}$$

$$\left\langle \Psi_{k} \frac{\mathrm{d}}{\mathrm{d}t} \sum_{i=0}^{P} v_{i} \Psi_{i} \right\rangle = \left\langle 2\Psi_{k} \sum_{h=0}^{P} b_{h} \Psi_{h} \sum_{i=0}^{P} z_{i} \Psi_{i} \sum_{j=0}^{P} z_{j} \Psi_{j} \right\rangle$$

$$- \left\langle 4d\Psi_{k} \sum_{i=0}^{P} u_{i} \Psi_{i} \sum_{j=0}^{P} v_{j} \Psi_{j} \right\rangle$$

Surface Reaction Model: Reorganize terms

$$\frac{\mathrm{d}}{\mathrm{d}t} v_k \left\langle \Psi_k^2 \right\rangle = 2 \sum_{h=0}^P \sum_{i=0}^P \sum_{j=0}^P b_h z_i z_j \left\langle \Psi_h \Psi_i \Psi_j \Psi_k \right\rangle - 4d \sum_{i=0}^P \sum_{j=0}^P u_i v_j \left\langle \Psi_i \Psi_j \Psi_k \right\rangle
\frac{\mathrm{d}}{\mathrm{d}t} v_k = 2 \sum_{h=0}^P \sum_{i=0}^P \sum_{j=0}^P b_h z_i z_j \frac{\left\langle \Psi_h \Psi_i \Psi_j \Psi_k \right\rangle}{\left\langle \Psi_k^2 \right\rangle} - 4d \sum_{i=0}^P \sum_{j=0}^P u_i v_j \frac{\left\langle \Psi_i \Psi_j \Psi_k \right\rangle}{\left\langle \Psi_k^2 \right\rangle}
\frac{\mathrm{d}}{\mathrm{d}t} v_k = 2 \sum_{h=0}^P \sum_{i=0}^P \sum_{j=0}^P b_h z_i z_j D_{hijk} - 4d \sum_{i=0}^P \sum_{j=0}^P u_i v_j C_{ijk}$$

- Pre-computing and storing the quad product D_{hijk} becomes cumbersome
- Use pseudo-spectral approach instead

Surface Reaction Model: Pseudo-Spectral approach for products

• Introduce auxiliary variable $g = z^2$

$$g_k = \sum_{i=0}^{P} \sum_{j=0}^{P} z_i z_j C_{ijk}$$

$$f = 2bz^2 = 2bg$$

$$f_k = 2\sum_{i=0}^{P} \sum_{j=0}^{P} b_i g_j C_{ijk}$$

- Limits the complexity of computing product terms
 - Higher products can be computed by repeated use of the same binary product rule
- Does introduce errors if order of PCE is not large enough

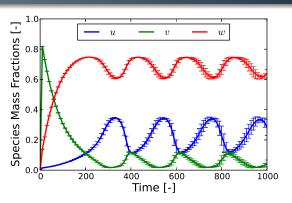
Surface Reaction Model: UQTk implementation

- All operations are replaced with their equivalent intrusive UQ counterparts
- Results in a set of coupled ODEs for the PC coefficients
 - u, v, w, z represent vector of PC coefficients
- This set of equations is integrated to get the evolution of the PC coefficients in time

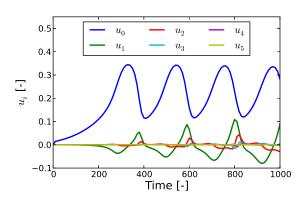
Surface Reaction Model: Second equation implementation

Dummy variables used where needed to build the terms in the equations

Surface Reaction Model: ISP results



- Assume 0.5% uncertainty in b around nominal value
- Legendre-Uniform intrusive PC
- Mean and standard deviation for u, v, and w
- Uncertainty grows in time



- Modes of u
- · Modes decay with higher order
- Amplitudes of oscillations of higher order modes grow in time

Galerkin Projection

$$u_k = \frac{\langle u\Psi_k \rangle}{\langle \Psi_k^2 \rangle} = \frac{1}{\langle \Psi_k^2 \rangle} \int u\Psi_k(\xi)w(\xi)d\xi, \quad k = 0, \dots, P$$

Evaluate projection integrals numerically

- Pick samples of uncertain parameters, e.g. $b(\xi)$ by sampling the germ ξ
- Run deterministic forward model for each of the sampled input parameter values b_i = b(ξ_i)
- Integration depends on sampling approach
 - Random Sampling: $\langle u\Psi_k \rangle = \frac{1}{N_c} \sum_{i=1}^{N_s} u(b_i) \Psi_k(\xi_i)$
 - Quadrature: $\langle u\Psi_k \rangle = \sum_{i=1}^{N_q} q_i u(b_i) \Psi_k(\xi_i)$

Reconstruct uncertain model output

$$u(x,t;\theta) = \sum_{k=0}^{P} u_k(x,t) \Psi_k(\xi(\theta))$$

Evaluate integral through sampling

$$\int u\Psi_k(\xi)w(\xi)d\xi = \frac{1}{N_s}\sum_{i=1}^{N_s}u(\xi_i)\Psi_k(\xi_i)$$

- Samples are drawn according to the distribution of ξ
 - Monte-Carlo (MC)
 - Latin-Hypercube-Sampling (LHS)
- Pros:
 - Can be easily made fault tolerant
 - Sometimes random samples is all we have
- Cons: slow convergence, but less dependent on number of stochastic dimensions

Quadrature approaches

Numerically evaluate integrals in Galerkin projection

$$\int u\Psi_k(\xi)w(\xi)d\xi = \sum_{i=1}^{N_q} q_i u(\xi_i)\Psi_k(\xi_i)$$

- Gauss quadrature rules are very efficient
 - ξ are quadrature points, with corresponding weights q_i
 - N_a quadrature points can integrate polynomial of order $2N_a 1$ exactly
 - Gauss-Hermite and Gauss-Legendre quadrature tailored to specific choices of the weight function $w(\xi)$
 - As a rule of thumb, p+1 quadrature points are needed for Galerkin projection of PCE of order p
 - If both u and Ψ_k are of order p, then integrand is of order 2p
 - $2p < 2N_a 1 \text{ or } N_a > p + \frac{1}{2}$
 - Only exact if u is indeed a polynomial of order $\leq p$
- Pros:
 - Can use existing codes as black box to evaluate $u(\xi_i)$
 - · Embarrassingly parallel
- Cons: Tensor product rule for d dimensions requires N_q^d samples

Non-Intrusive propagation of uncertainty - Collocation

- Collocation techniques minimize errors at sample points
 - $\sum_{k=0}^{P} u_k \Psi_k(\xi_i) = u(\xi_i)$, $i = 1, \dots, N_c$ Can use interpolation, e.g. Lagrange interpolants

 - Or use regression approaches: P + 1 degrees of freedom to fit N_c points
- Pros: can position points where most accuracy desired
- Cons:

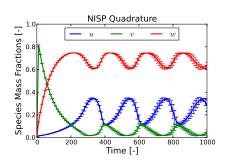
Surface Reaction Model: NISP implementation in UQTk

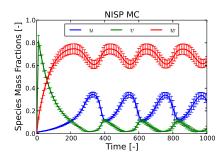
Quadrature:

```
// Get the quadrature points
int nqdpts=myPCSet.GetNQuadPoints();
double* qdpts=new double[nQdpts];
myPCSet.GetQuadPoints(qdpts);
...
// Evaluate parameter at quad pts
for(int i=0;i<nQdpts;i++) {
    bval[i]=myPCSet.EvalPC(b,&qdpts[i]);
}
...
// Run model for all samples
for(int i=0;i<nQdpts;i++) {
    u_val[i] = ...
}
// Spectral projection
myPCSet.GalerkProjection(u_val,u);
myPCSet.GalerkProjection(v_val,v);
myPCSet.GalerkProjection(w_val,v);</pre>
```

Monte-Carlo Sampling:

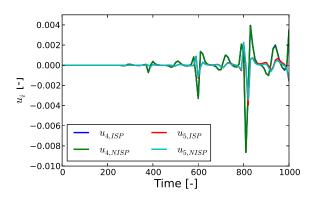
```
// Get the sample points
int nSamples=1000;
Arrav2D<double> samPts(nSamples.dim);
myPCSet.DrawSampleVar(samPts);
// Evaluate parameter at sample pts
for (int i=0:i<nSamples:i++) {
  ... // select samPt from samPts
  bval[i]=mvPCSet.EvalPC(b,&samPt)
// Run model for all samples
for(int i=0;i<nSamples;i++) {
  u \ val[i] = ...
// Spectral projection
myPCSet.GalerkProjectionMC(samPts,u_val,u);
myPCSet.GalerkProjectionMC(samPts,v val,v);
myPCSet.GalerkProjectionMC(samPts,w val,w);
```





- Mean and standard deviation for u, v, and w
- Quadrature approach agrees well with ISP approach using 6 quadrature points
- Monte Carlo sampling approach converges slowly
 - With a 1000 samples, results are quite different from ISP and NISP

Surface Reaction Model: Comparison ISP and NISP



- Lower order modes agree perfectly
- Very small differences in higher order modes
 - · Difference increases with time

Further Reading

- R. Vigil and F. Willmore, "Oscillatory dynamics in a heterogeneous surface reaction: Breakdown of the mean-field approximation.," Phys Rev E Stat Phys Plasmas Fluids Relat Interdiscip Topics, vol. 54, no. 2, pp. 1225-1231, Aug. 1996.
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- O. P. Le Maître and O. M. Knio, "Spectral Methods for Uncertainty Quantification: With Applications to Computational Fluid Dynamics" (Scientific Computation), 1st ed. Springer, 2010.