### **PCE VS RSM**

This example is with known exact solution. Compared with analytical solution, this example hopes to provide surrogate model to create

customer defined loglikelihood to do Bayesian inference

#### 1 - INITIALIZE UQLAB

```
clearvars
rng(100, 'twister')
uqlab
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This is UQLab, version 2.0
UQLab is distributed under the BSD 3-clause open source license available at:
C:\NY2023\D_document\UQLab_Rel2.0.0\LICENSE.
To request special permissions, please contact:
  - Stefano Marelli (marelli@ibk.baug.ethz.ch).
Useful commands to get started with UQLab:
                 - Access the available documentation
uqlab -doc
uqlab -help
                    - Additional help on how to get started with UQLab
uq_citation help - Information on how to cite UQLab in publications
uqlab -license
                 - Display UQLab license information
```

#### 2 - Data for FE realization and measurement

Read measurements for each points along the beam *Measurement Y* (dim 10X29): 10 sets of experiments and 29 monitored points *Z* along

the beam for each experiement

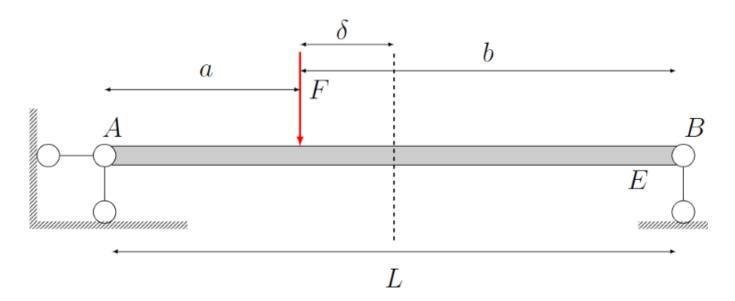
```
Measurement = xlsread("data.xlsx", 'Measurement_for_surrogate', 'A2:AC11');
```

Read FE realization from LHS sampling

```
y = \mathcal{M}(x)<br/>Dim (248X3)
```

```
FE_Realization = xlsread("data.xlsx",'FE_models','A2:AE249');
```

## 2 - PRIOR DISTRIBUTION for PCE model and Bayesian inference



```
b_b = 0.15; % beam width (m)
b_h = 0.3; % beam height (m)
a % distance from the point A (m)
b % distance from the point B (m)
L = 30; % beam length (m)
F = 43000;% Concentrated force (N)
```

## 3 - Calculate the residual r(x,z) for surrogate model $\widetilde{\mathcal{M}}$

Calculate the Root-mean-square-error data for surrogate

$$r(\overrightarrow{x}, \overrightarrow{z}) = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (Y_i - \mathcal{M}(\overrightarrow{x}, \overrightarrow{z}))^2}$$

$$\overrightarrow{x} = [E, \delta]; \overrightarrow{z} = [z_1, z_2, ..., z_{28}, z_{29}];$$

r is residual;  $\overrightarrow{x}$  is parameters of interests; E is elastic modulus;  $\delta$  is the loading postion

 $\overrightarrow{z}$  is the different measurement points along the beam;

 $\mathcal{M}$  is the FE model;  $Y_i$  is the measurement data; N is the number of experiment expNum.

```
expNum = size(Measurement,1);% experiment number N = 10

Nreal = size (FE_Realization,1);% number of FE realization = 248

Npoint = size(Measurement,2);% number of z [z1,z2,...,z28,z29]

Residual = []; % intitail value for Residual

%loop to calculate the ResidualSum
for i = 1: expNum

    error = ((Measurement(i,:) - FE_Realization(1:Nreal,3:31)) ).^2;
    Residual = [Residual, sum(error,2)];

end

%Divide the number of experiment expNum = 10 and monitored points number along
%the beam Npoint = 29
ResidualSum = sqrt(sum(Residual,2)/expNum/Npoint);
```

## 4 - Create RSM surrogate $\widetilde{\mathcal{M}}(r)$

RSM:Response surface model

```
%Get the residual data for RSM

RSM_data = [FE_Realization(1:Nreal,1),FE_Realization(1:Nreal,2),ResidualSum];
```

Response surface model fitting

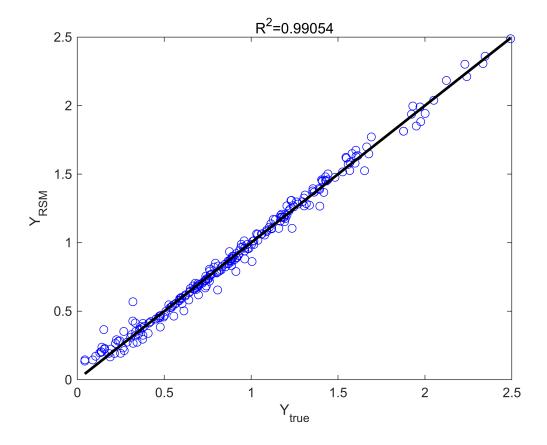
```
x(:,1) = RSM data(:,1)./10e9;% normlized the E with 10e9Pa, beacuase E is too big
to fit good results
x(:,2) = RSM data(:,2)./5;% normalized the delta with deviatoin = 5m
y_fit = RSM_data(:,3); % sum error
% response surface model-- Six order polynomial equation
model fun RSM = @(p,x) p(1) + ...
p(2) * x(:,1) + ...
p(3) * x(:,2) + ...
p(4) * x(:,1).^2 + ...
p(5) * x(:,2).^2 + ...
p(6) * x(:,1).^3 + ...
p(7) * x(:,2).^3 + ...
p(8) * x(:,1).^4 + ...
p(9) * x(:,2).^4 + ...
p(10) * x(:,1).^3 .* x(:,2) + ...
p(11) * x(:,1).^2 .* x(:,2).^2 + ...
p(12) * x(:,1) .* x(:,2).^3 + ...
p(13) * x(:,1).^4 .* x(:,2) + ...
p(14) * x(:,1) .* x(:,2).^4 + ...
p(15) * x(:,1).^5 + ...
p(16) * x(:,2).^5 + ...
p(17) * x(:,1).^4 .* x(:,2).^2 + ...
p(18) * x(:,1).^3 .* x(:,2).^3 + ...
p(19) * x(:,1).^2 .* x(:,2).^4 + ...
p(20) * x(:,1) .* x(:,2).^5 + ...
p(21) * x(:,1).^6 + ...
p(22) * x(:,2).^6 + ...
p(23) * x(:,1).^5 .* x(:,2) + ...
p(24) * x(:,1).^4 .* x(:,2).^2 + ...
p(25) * x(:,1).^3 .* x(:,2).^3 + ...
p(26) * x(:,1).^2 .* x(:,2).^4 + ...
p(27) * x(:,1) .* x(:,2).^5 + ...
p(28) * x(:,1).^6 .* x(:,2) + ...
p(29) * x(:,1).^5 .* x(:,2).^2 + ...
p(30) * x(:,1).^4 .* x(:,2).^3 + ...
p(31) * x(:,1).^3 .* x(:,2).^4 + ...
p(32) * x(:,1).^2 .* x(:,2).^5 + ...
p(33) * x(:,1) .* x(:,2).^{6};
value for the coeficients
[p,R,J,CovB,MSE,ErrorModelInfo] = nlinfit(x,y_fit,model_fun_RSM,p0);
```

警告: 解处的 Jacobian 矩阵为病态,而且某些模型参数的估计值可能不准确(不可识别)。进行预测时要谨慎。

% coeficient :https://ww2.mathworks.cn/help/stats/nlinfit.html#d124e38841

Surrogate mode: Predicted vs True

```
y_estimate = model_fun_RSM(p,x);
SSR = sum(R.^2);
SST = sum((y_fit - mean(y_fit)).^2);
R2 = 1 - SSR / SST;
str=['R^2=', num2str(R2)];
close all;
plot(y_fit,y_estimate,'bo');
hold on;
plot([min(y_fit),max(y_fit)],[min(y_fit),max(y_fit)],'k-','LineWidth',2);
title(str);
xlabel('Y_{true}')
ylabel('Y_{RSM}')
hold off;
```



# 5 - Create PCE surrogate $\widetilde{\mathcal{M}}(r)$

PCE: Polynomial Chaos expansion

```
%Get the residual data for PCE
PCE_data = [FE_Realization(1:Nreal,1),FE_Realization(1:Nreal,2),ResidualSum];
```

#### myPCE = PCE\_create(PCE\_data);

```
Calculating the PCE coefficients by regression.
警告: Warning: numerical instability!! Gamma for LAR iteration 5 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 7 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 8 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 9 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 14 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 11 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 12 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 13 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 19 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 20 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 25 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 27 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 31 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 32 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 33 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 34 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 35 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 34 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 35 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 44 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 10 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 53 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 54 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 16 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 22 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 47 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 51 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 52 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 53 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 54 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 55 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 56 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 58 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 59 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 60 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 61 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 62 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 63 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 64 was set to 0 to prevent crashes.
警告: Warning: numerical instability!! Gamma for LAR iteration 65 was set to 0 to prevent crashes.
The estimation of PCE coefficients stopped at polynomial degree 10 and qNorm 1.00 for output variable 1
Final LOO error estimate: 4.064805e-01
                   Calculation finished!
```

#### Print a summary of the resulting PCE metamodel:

#### uq\_print(myPCE)

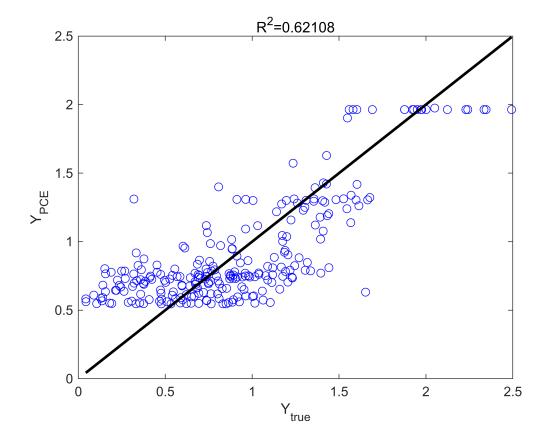
```
%-----Polynomial chaos output -----%
  Number of input variables:
                                2
  Maximal degree:
                                10
  q-norm:
                                1.00
  Size of full basis:
                                66
  Size of sparse basis:
                               5
  Full model evaluations:
                               248
  Leave-one-out error:
                               3.9285616e-01
  Modified leave-one-out error: 4.0648052e-01
```

```
Mean value:
                                 0.2572
  Standard deviation:
  Coef. of variation:
                                  31.295%
hold on
```

0.8220

#### Plot the true vs. predicted values:

```
%Obtain the X_val and YPCE/Y_val
X_val(:,1) = PCE_data(1:Nreal,1);
X_val(:,2) = PCE_data(1:Nreal,2);
YPCE = uq_evalModel(myPCE,X_val);
Y_val = PCE_data(1:Nreal,3);
R = Y_val - YPCE;
SSR = sum(R.^2);
SST = sum((Y_val - mean(Y_val)).^2);
R2 = 1 - SSR / SST;
str=['R^2=', num2str(R2)];
close all;
plot(Y_val,YPCE,'bo');
hold on;
plot([min(Y_val),max(Y_val)],[min(Y_val),max(Y_val)],'k-','LineWidth',2);
title(str);
xlabel('Y_{true}')
ylabel('Y_{PCE}')
hold off;
```



```
save myPCE
```

### 6 - Define the priors for discrepancy $\sigma$

By default, UQlab assumes an independent and identically distributed discrepancy

$$\varepsilon \sim \mathcal{N}(0, \mu_y^2)$$
, with  $\mu_y = \frac{1}{N} \sum_{i=1}^{N} y_i$ 

```
PriorOpts.Marginals(3).Name = 'sigma2'; % variance
PriorOpts.Marginals(3).Type = 'Uniform';
sigma2 = mean(Measurement(:,:),"all");
PriorOpts.Marginals(3).Parameters = [0 sigma2^2];
myPriorDist = uq_createInput(PriorOpts);
```

# 7 - Define the custom-loglikelihood and measurement data for UQlab calculation

$$\log \mathcal{L}(\Theta) = \sum_{i=1}^{N} \left( -\frac{1}{2} \left( r_i(\overrightarrow{x}, \overrightarrow{z}) \right)^T \Sigma(\epsilon)^{-1} \left( r_i(\overrightarrow{x}, \overrightarrow{z}) \right) \right) - \frac{3N}{2} \log(2\pi) - \frac{N}{2} \log\left(\det(\Sigma(\epsilon))\right)$$

Monitored data  $Y_i$  for beam deflection have been used in PCE surrogate model. No measurements available for Bayesian inference: Pass zeros to measurement  $Y_i$  just for calculation, which equals:

$$r(\overrightarrow{x}, \overrightarrow{z}) \sim (0 - r(\overrightarrow{x}, \overrightarrow{z})) \sim (Y_i - \mathcal{M}(\overrightarrow{x}, \overrightarrow{z}))$$

$$r(\overrightarrow{x}, \overrightarrow{z}) \dim 248X1 \sim (Y_i - \mathcal{M}(\overrightarrow{x}, \overrightarrow{z})) \dim 248X29$$

Also, measurement data for  $Y_{10X29}$  changed into  $Y_{10X1}$ 

```
y = zeros(1,1);
myData.y = y;
myData.Name = 'Zeros vector measurement for Bayesian inference';
```

Loglikelihood still follows the Gaussian discrepancy criteria

```
myLogLikeli = @(params,y) myLogLikeli2(params,y);
```

## 8 - Solver options

```
Solver.Type = 'MCMC';
Solver.MCMC.Visualize.Parameters = [1 2];
Solver.MCMC.Visualize.Interval = 10;
Solver.MCMC.Sampler = 'AIES';
```

```
Solver.MCMC.Steps = 500;
Solver.MCMC.NChains = 5;
Solver.MCMC.Proposal.PriorScale = 1e-3;
```