## POD-RBF

Let the response data be stored in a matrix called the snapshot U:

(1)

Where is the total number of sensor points. And is the number of snapshot vectors, each representing the response corresponding a unique set of parameters. On the other hand, the matrix stores parameters.

The main purpose of POD is to extract a set of orthogonal vectors, called POD basis vectors. Where is related to the matrix **U** by the following linear relationship:

(2)

Thus, according to the orthogonally of, the amplitude matrix A can be computed from:

(3)

These vectors can also be calculated through the POD operation [[1](#_ENREF_1), [2](#_ENREF_2)], by extracting the eigenvectors of the covariance matrix

(4)

, (5)

Where is the matrix storing the normalized eigenvectors of the covariance matrix, and the diagonal matrix storing its eigenvalues. This method gives eigenvalues that are stored in a descending order. Also called POD directions. And they represent the original data.

Because lower values POD directions hold very little representation, they can be sacrificed in order to reduce the size of the model. With negligible influence the accuracy of the representation. This step is called the truncation. Consequently, the amplitude matrix is given by:

(6)

Where is the sub-space extracted out from by preserving the first K (K ≪ S) columns of that correspond to the largest eigenvalues. We can then reconstruct the original data using the reduced model.

(7)

On the other hand, Radial Basis Function (RBF) allows the generation of the responses corresponding to new sets parameters by interpolation between values of the initial selection. This is done using the matrix containing the coefficients calculated by the following equation:

(8)

Where is already calculated by equation (3) and **G** is defined as the matrix of interpolation functions. In this work, the identity RBF been employed [[3-5](#_ENREF_3)]:

(9)

In Equation (9), the set of interpolation functions is arbitrary chosen from the matrix **G** and is the parameter corresponding to (i=1,2,…,S). The argument of the i-th RBF is the distance, and being respectively current and reference parameters.

After the evaluation of the coefficient matrix , a low-dimensional model of (8) can be put in vector form, as follows:

(10)

By defining the amplitude vector as a function of parameters **P**, the Equation (7) can be rewritten as the approximation of the snapshot corresponding to a new parameter vector:

(11)

The reduced model given by Equation (11) is defined as the POD-RBF network. It is able to reproduce unknown response fields corresponding to any set of parameter However, it should be noted that this model can lead to results with weak precision when making extrapolation outside the range of known parameters **P**.

**References**

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