


POLYNOMIAL CURVE FITTING FOR CONTROL ROD WORTH USING LEAST SQUARE NUMERICAL ANALYSIS

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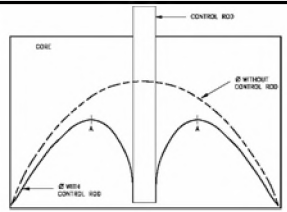
THE CONTENT

- Flux distribution & control rod effectiveness
- Problem statement
- Curve fitting
- SCILAB 5.3.3 in action
- Results & Discussion

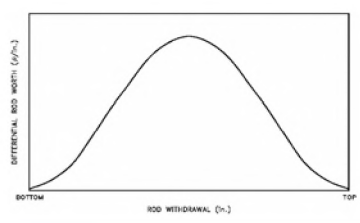
Flux Distribution & Control Rod Effectiveness

Neutron flux distribution depends on;

- Reflector effect
- Fuel burn-up
- Temperature effect
- Control rod worth




- Control rod worth is the change in reactivity caused by control rod motion.
- If additional rods are added to this simple reactor, the most effective location is where the flux is maximum, that is, at point A
- Numerous control rods are required for a reactor that has a large amount of excess reactivity (that amount of reactivity in excess of that needed to be critical)



- Differential control rod worth is the reactivity change per unit movement of a rod and is normally expressed as ρ/inch or $\Delta k/k$ per inch.
- The integral rod worth at a given withdrawal is merely the summation of all the differential rod worth up to that point of withdrawal
- It is also the area under the differential rod worth curve at any given withdrawal position

PROBLEM STATEMENT

SAFETY CONTROL ROD Differential Reactivity Curve
21 Feb 2012



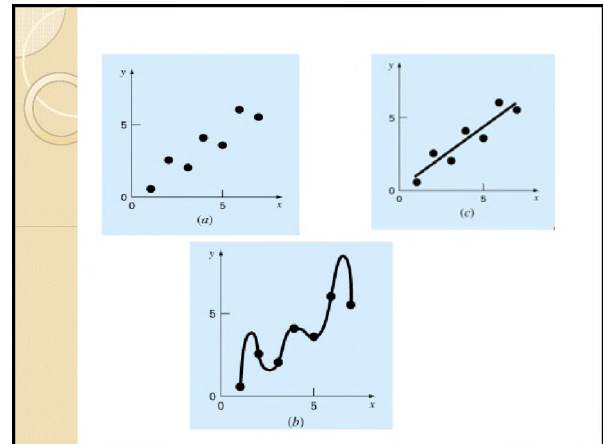
Rod Position	Differential Rod Worth (k/k)
100	0.0010
200	0.0020
300	0.0025
400	0.0045
500	0.0050
600	0.0045
700	0.0025
800	0.0015
900	0.0005

CURVE FITTING

Describes techniques to fit curves (*curve fitting*) to discrete data to obtain intermediate estimates.

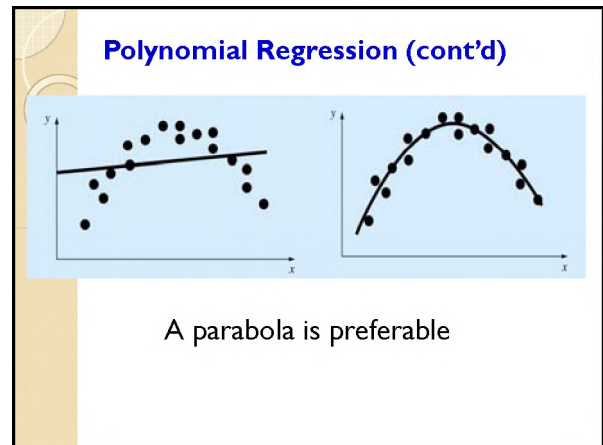
There are two general approaches for curve fitting:

- **Least Squares regression:**
Data exhibit a significant degree of scatter. The strategy is to derive a single curve that represents the general trend of the data.
- **Interpolation:**
Data is very precise. The strategy is to pass a curve or a series of curves through each of the points.



Polynomial Regression

- Some engineering data is poorly represented by a straight line.
- For these cases a curve is better suited to fit the data.
- The least squares method can readily be extended to fit the data to higher order polynomials.



Polynomial Regression (cont'd)

- A **2nd order polynomial (quadratic)** is defined by:
 $y = a_0 + a_1x + a_2x^2 + e$
- The residuals between the model and the data:
 $e_i = y_i - a_0 - a_1x_i - a_2x_i^2$
- The sum of squares of the residual:
 $S_r = \sum e_i^2 = \sum (y_i - a_0 - a_1x_i - a_2x_i^2)^2$

Polynomial Regression (cont'd)

$$\frac{\partial S_r}{\partial a_0} = -2 \sum (y_i - a_0 - a_1x_i - a_2x_i^2) = 0$$

$$\frac{\partial S_r}{\partial a_1} = -2 \sum (y_i - a_0 - a_1x_i - a_2x_i^2)x_i = 0$$

$$\frac{\partial S_r}{\partial a_2} = -2 \sum (y_i - a_0 - a_1x_i - a_2x_i^2)x_i^2 = 0$$

$$\left. \begin{aligned} \sum y_i &= n \cdot a_0 + a_1 \sum x_i + a_2 \sum x_i^2 \\ \sum x_i y_i &= a_0 \sum x_i + a_1 \sum x_i^2 + a_2 \sum x_i^3 \\ \sum x_i^2 y_i &= a_0 \sum x_i^2 + a_1 \sum x_i^3 + a_2 \sum x_i^4 \end{aligned} \right\} \begin{array}{l} \text{3 linear equations} \\ \text{with 3 unknowns} \\ (a_0, a_1, a_2), \text{ can be} \\ \text{solved} \end{array}$$

Polynomial Regression (cont'd)

- A system of 3x3 equations needs to be solved to determine the coefficients of the polynomial.

$$\begin{bmatrix} n & \sum x_i & \sum x_i^2 \\ \sum x_i & \sum x_i^2 & \sum x_i^3 \\ \sum x_i^2 & \sum x_i^3 & \sum x_i^4 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} \sum y_i \\ \sum x_i y_i \\ \sum x_i^2 y_i \end{bmatrix}$$

- The standard error

$$s_{y/x} = \sqrt{\frac{S_r}{n-3}}$$

Polynomial Regression- Example (cont'd)

x_i	y_i	y_{model}	e_i^2	$(y_i y)^2$
0	2.1	2.4786	0.14332	544.42889
1	7.7	6.6986	1.00286	314.45929
2	13.6	14.64	1.08158	140.01989
3	27.2	26.303	0.80491	3.12229
4	40.9	41.687	0.61951	239.22809
5	61.1	60.793	0.09439	1272.13489
15	152.6	3.74657	2513.39333	

- The standard error of estimate:
 $s_{y/x} = \sqrt{\frac{3.74657}{6-3}} = 1.12$
- The coefficient of determination:
 $r^2 = \frac{2513.39 - 3.74657}{2513.39} = 0.99851, \quad r = \sqrt{r^2} = 0.99925$

SCILAB 5.3.3 in action

```

--aa = X\y
a =
- 2.4440308
- 0.000274
- 0.0229468

--aa = (100:110:990)'
xx =
100.
120.
140.
160.
180.
200.
220.
240.
260.
280.
300.

--yy = a(1)*xx.^2 + a(2)*xx + a(3)
yy =
- 0.001159
- 0.002035
- 0.003024
- 0.004026
- 0.004955
- 0.004818
- 0.003422
- 0.001496
- 0.000285

--e = y-yy
e =
0.00034

```

RESULTS & DISCUSSION

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Residual e_i : 0.00034

Error : 0.5%

$f(x) = -0.003x^2 + 0.26x - 2.6E-5$ for PUSPATI TRIGA Reactor's 2012 safety rod calibration.

THANK YOU