

## Section 1

MANE 6313

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## Subsection 1

Week 13, Module D

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# Student Learning Outcome

- Select an appropriate experimental design with one or more factors,
- Select an appropriate model with one or more factors,
- Evaluate statistical analyses of experimental designs,
- Assess the model adequacy of any experimental design, and
- Interpret model results.

## Module Learning Outcome

*Employ `rsm()` to design and analyze RSM models.*

## RSM Design in Package rsm()

- `rsm()` supports two designs:
  - Box-Behnken design with 2 to 7 factors
  - Center Composite Design (ccd)
- Details and examples of using both designs will be provided in separate modules

## Model Types

- The model formula for designs in `rsm()` is different
  - `FO()` specifies first-order model
  - `TWI()` is used to generate two-way interactions
  - `PQ()` is used to add pure quadratic terms to model
  - `SO()` creates all terms (`FO()`, `TWI()`, `PQ()`) in a model

## Coded Variables

- When analyzing RSM models, variables should be coded
- The R chunk for creating coded variables in a model is shown below

```
library(rsm)
```

```
bb3.design <- bbd(3,n0=2,coding=list(x1~(Force-20)/3,x2~(Rate-
```

```
print(bb3.design)
```

## Coded Variables Output

```

14 > ```{r}
15 library(rsm)
16 bb3.design <- bbd(3,n0=2,coding=list(x1~(Force-20)/3,x2~(Rate-50)/10,x3~Polish-4))
17 print(bb3.design)
18 > ```

```

	run.order	std.order	Force	Rate	Polish
1	1	7	17	50	5
2	2	10	20	60	3
3	3	9	20	40	3
4	4	14	20	50	4
5	5	3	17	60	4
6	6	5	17	50	3
7	7	4	23	60	4
8	8	6	23	50	3
9	9	13	20	50	4
10	10	12	20	60	5
11	11	8	23	50	5
12	12	1	17	40	4
13	13	11	20	40	5
14	14	2	23	40	4

Data are stored in coded form using these coding formulas ...

```

x1 ~ (Force - 20)/3
x2 ~ (Rate - 50)/10
x3 ~ Polish - 4

```



## Adding a Response Variable

```
y <- rnorm(14)
bb3.design$y <- y
print(bb3.design)
```

## Adding a Response Variable Output

```

20 > ```{r}
21 y <- rnorm(14)
22 bb3.design$y <- y
23 print(bb3.design)
24 > ```

```

	run.order	std.order	Force	Rate	Polish	y
1	1	7	17	50	5	1.3724421
2	2	10	20	60	3	1.7480628
3	3	9	20	40	3	-0.5099157
4	4	14	20	50	4	0.8571949
5	5	3	17	60	4	-1.5198610
6	6	5	17	50	3	1.2546819
7	7	4	23	60	4	0.1785562
8	8	6	23	50	3	-0.2568965
9	9	13	20	50	4	-1.8243286
10	10	12	20	60	5	1.2297702
11	11	8	23	50	5	0.3593844
12	12	1	17	40	4	1.0656732
13	13	11	20	40	5	1.1265194
14	14	2	23	40	4	0.9289255

Data are stored in coded form using these coding formulas ...

$x1 \sim (Force - 20)/3$

$x2 \sim (Rate - 50)/10$

## Model Fitting

```
bb3.fitted <- rsm(y~S0(x1,x2,x3),data=bb3.design)  
summary(bb3.fitted)
```

# Model Fitting Output

```

26 ~```{r}
27 bb3.fitted <- rsm(y~S0(x1,x2,x3),data=bb3.design)
28 summary(bb3.fitted)
29 ~```

Call:
rsm(formula = y ~ S0(x1, x2, x3), data = bb3.design)

              Estimate Std. Error t value Pr(>|t|)
(Intercept) -0.48357      1.10443  -0.4378  0.6841
x1           -0.12037      0.55222  -0.2180  0.8381
x2           -0.12183      0.55222  -0.2206  0.8362
x3            0.23152      0.55222   0.4193  0.6966
x1:x2         0.45879      0.78095   0.5875  0.5884
x1:x3         0.12463      0.78095   0.1596  0.8809
x2:x3        -0.53868      0.78095  -0.6898  0.5283
x1^2          0.21534      0.87313   0.2466  0.8173
x2^2          0.43155      0.87313   0.4943  0.6470
x3^2          0.95063      0.87313   1.0888  0.3375

Multiple R-squared:  0.3734,    Adjusted R-squared:  -1.036
F-statistic: 0.2649 on 9 and 4 DF,  p-value: 0.9546

Analysis of Variance Table

Response: y
              Df Sum Sq Mean Sq F value Pr(>F)
F0(x1, x2, x3)  3  0.6635   0.2212   0.0907  0.9614
TWI(x1, x2, x3)  3  2.0648   0.6883   0.2821  0.8367
PQ(x1, x2, x3)   3  3.0869   1.0290   0.4218  0.7479
Residuals       4  9.7581   2.4395
Lack of fit      3  6.1628   2.0543   0.5714  0.7223
Pure error       1  3.5953   3.5953

Stationary point of response surface:
              x1          x2          x3
1.2224666 -0.7710184 -0.4203606

Stationary point in original units:
      Force      Rate      Polish
23.667400 42.289816 3.579639

Eigenanalysis:
eigen() decomposition
$values
[1] 1.06647794 0.50637516 0.02466496

$vectors
              [,1]      [,2]      [,3]
x1 -0.04164836 -0.6261646  0.7785777

```