

```
> data <- read.csv("Conversion.csv"); data
```

These are data taken from the examples in the Design-Expert manual. We are making a product that should have an activity of 63; we could accept the range 60 to 66, but 63 is more desirable. There are three process variables: time (minutes), temperature (degrees C), and catalyst (percent). When we do our experiment, we have two responses: the conversion factor (essentially the percent yield) and the activity.

Our goal is to get process variables so that we hit an activity of 63 (or very close) and maximize the conversion. In addition, we want to operate within the “cube” part of the experimental region.

	Time	Temperature	Catalyst	Conversion	Activity
1	45.00	85.00	2.50	75	60.4
2	45.00	85.00	2.50	83	60.6
3	45.00	85.00	2.50	76	59.1
4	45.00	85.00	2.50	81	59.2
5	45.00	85.00	2.50	91	58.9
6	45.00	85.00	2.50	80	60.8
7	53.41	85.00	2.50	79	65.9
8	36.59	85.00	2.50	76	53.6
9	45.00	85.00	1.66	55	57.4
10	45.00	85.00	3.34	81	63.2
11	45.00	93.41	2.50	97	60.7
12	45.00	76.59	2.50	85	60.0
13	40.00	80.00	2.00	74	53.2
14	40.00	90.00	3.00	66	59.8
15	50.00	90.00	2.00	70	62.6
16	40.00	90.00	2.00	88	53.4
17	50.00	90.00	3.00	97	67.8
18	50.00	80.00	3.00	90	67.9
19	40.00	80.00	3.00	71	57.3
20	50.00	80.00	2.00	51	62.9

```
> CD <- coded.data(data, cTime~(Time-45)/5, cTemp~(Temperature-85)/5,
  cCat~(Catalyst-2.5)/.5); CD
```

This was a central composite design with center at (45, 85, 2.5) and steps of (5, 5, and .5). So let's recode the variables.

	cTime	cTemp	cCat	Conversion	Activity
1	0.000	0.000	0.00	75	60.4
2	0.000	0.000	0.00	83	60.6
3	0.000	0.000	0.00	76	59.1
4	0.000	0.000	0.00	81	59.2
5	0.000	0.000	0.00	91	58.9
6	0.000	0.000	0.00	80	60.8
7	1.682	0.000	0.00	79	65.9
8	-1.682	0.000	0.00	76	53.6
9	0.000	0.000	-1.68	55	57.4
10	0.000	0.000	1.68	81	63.2
11	0.000	1.682	0.00	97	60.7
12	0.000	-1.682	0.00	85	60.0
13	-1.000	-1.000	-1.00	74	53.2
14	-1.000	1.000	1.00	66	59.8
15	1.000	1.000	-1.00	70	62.6
16	-1.000	1.000	-1.00	88	53.4

```

17  1.000  1.000  1.00          97      67.8
18  1.000 -1.000  1.00          90      67.9
19 -1.000 -1.000  1.00          71      57.3
20  1.000 -1.000 -1.00         51      62.9

```

Variable codings ...

```

cTime ~ (Time - 45)/5
cTemp ~ (Temperature - 85)/5
cCat  ~ (Catalyst - 2.5)/0.5

```

```
> fit1<-rsm(Conversion~SO(cTime, cTemp, cCat) , data=CD)
```

Let's fit a second order model to the conversion (yield) variable.

```
> plot(fit1);boxCox(fit1)
```

Not shown, but things look OK with the residuals, etc.

```
> summary(fit1)
```

Looking things over, there is no (significant) lack of fit for this model, and we see that there are highly significant effects at all orders. Temperature and catalyst seem to behave in a second order fashion. Time does not seem to do much by itself, but it interacts (a lot) with catalyst.

The stationary point is more or less within the region of experimentation, but the eigen analysis shows that it is a saddle point, not a maximum or minimum.

Call:

```
rsm(formula = Conversion ~ SO(cTime, cTemp, cCat), data = CD)
```

Residuals:

```

      Min       1Q   Median       3Q      Max
-6.092 -1.452 -0.289  1.474  9.908

```

Coefficients:

```

              Estimate Std. Error t value Pr(>|t|)
(Intercept)   81.092      1.923  42.163 1.35e-12 ***
cTime          1.028      1.276   0.806  0.43902
cTemp          4.040      1.276   3.166  0.01005 *
cCat           6.206      1.277   4.861  0.00066 ***
cTime:cTemp    2.125      1.667   1.275  0.23129
cTime:cCat    11.375      1.667   6.823 4.61e-05 ***
cTemp:cCat    -3.875      1.667  -2.324  0.04247 *
cTime^2       -1.831      1.242  -1.474  0.17115
cTemp^2        2.941      1.242   2.368  0.03942 *
cCat^2        -5.203      1.244  -4.181  0.00188 **
---

```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 4.716 on 10 degrees of freedom

Multiple R-squared: 0.92, Adjusted R-squared: 0.8479

F-statistic: 12.77 on 9 and 10 DF, p-value: 0.0002207

Analysis of Variance Table

Response: Conversion

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
FO(cTime, cTemp, cCat)	3	762.93	254.31	11.4357	0.001428
TWI(cTime, cTemp, cCat)	3	1191.37	397.12	17.8576	0.000243
PQ(cTime, cTemp, cCat)	3	601.51	200.50	9.0160	0.003414
Residuals	10	222.38	22.24		
Lack of fit	5	56.38	11.28	0.3397	0.869479
Pure error	5	166.00	33.20		

Stationary point of response surface:

cTime	cTemp	cCat
-1.0179845	-0.5294920	-0.3192329

Stationary point in original units:

Time	Temperature	Catalyst
39.910077	82.352540	2.340384

Eigenanalysis:

\$values

[1] 3.408437 2.322758 -9.824322

\$vectors

	[,1]	[,2]	[,3]
[1,]	0.1380685	0.7982920	0.5862312
[2,]	-0.9428308	0.2872127	-0.1690534
[3,]	0.3033270	0.5293758	-0.7923093

> **fit2<-rsm(Activity~SO(cTime,cTemp,cCat),data=CD)**

Now let's try fitting activity.

> **plot(fit2);boxCox(fit2)**

Not shown, but things are OK with this model too.

> **summary(fit2)**

The main thing we see is that the first order terms are highly significant, but the mixed and quadratic terms are not.

Call:

rsm(formula = Activity ~ SO(cTime, cTemp, cCat), data = CD)

Residuals:

Min	1Q	Median	3Q	Max
-1.2757	-0.6433	-0.1112	0.6254	1.1969

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	59.81833	0.42544	140.603	< 2e-16 ***
cTime	4.26033	0.28226	15.094	3.29e-08 ***
cTemp	0.25460	0.28226	0.902	0.388
cCat	2.23118	0.28240	7.901	1.31e-05 ***
cTime:cTemp	-0.38750	0.36881	-1.051	0.318
cTime:cCat	-0.03750	0.36881	-0.102	0.921

```

cTemp:cCat    0.31250    0.36881    0.847    0.417
cTime^2       0.06769    0.27472    0.246    0.810
cTemp^2       0.27977    0.27472    1.018    0.333
cCat^2        0.26294    0.27525    0.955    0.362

```

```
---
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Residual standard error: 1.043 on 10 degrees of freedom
```

```
Multiple R-squared:  0.9672, Adjusted R-squared:  0.9376
```

```
F-statistic: 32.74 on 9 and 10 DF,  p-value: 2.955e-06
```

Analysis of Variance Table

```
Response: Activity
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
FO(cTime, cTemp, cCat)	3	316.71	105.571	97.0169	1.079e-07
TWI(cTime, cTemp, cCat)	3	1.99	0.665	0.6107	0.6232
PQ(cTime, cTemp, cCat)	3	1.94	0.645	0.5931	0.6336
Residuals	10	10.88	1.088		
Lack of fit	5	7.23	1.446	1.9786	0.2359
Pure error	5	3.65	0.731		

```
Stationary point of response surface:
```

```

      cTime      cTemp      cCat
15.05168  17.57008 -13.61031

```

```
Stationary point in original units:
```

```

      Time Temperature      Catalyst
120.258424  172.850398   -4.305155

```

```
Eigenanalysis:
```

```
$values
```

```
[1]  0.48931661  0.18268588 -0.06160297
```

```
$vectors
```

```

      [,1]      [,2]      [,3]
[1,]  0.3691679  0.4710760  0.8011258
[2,] -0.7502855 -0.3576299  0.5560328
[3,] -0.5484402  0.8063425 -0.2214160

```

```
> fit2b<-rsm(Activity~FO(cTime, cTemp, cCat), data=CD)
```

```
Refit with just first order terms.
```

```
> summary(fit2b)
```

```
Still no lack of fit. Coefficients are the same as what we had before (except for intercept).
```

```
Call:
```

```
rsm(formula = Activity ~ FO(cTime, cTemp, cCat), data = CD)
```

```
Residuals:
```

```

      Min      1Q  Median      3Q      Max

```

```
-1.5009 -0.6843  0.1231  0.6285  1.4281
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	60.2350	0.2151	279.977	< 2e-16 ***
cTime	4.2603	0.2603	16.364	2.06e-11 ***
cTemp	0.2546	0.2603	0.978	0.343
cCat	2.2312	0.2605	8.566	2.26e-07 ***

Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 0.9621 on 16 degrees of freedom

Multiple R-squared: 0.9553, Adjusted R-squared: 0.9469

F-statistic: 114 on 3 and 16 DF, p-value: 5.193e-11

Analysis of Variance Table

Response: Activity

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
FO(cTime, cTemp, cCat)	3	316.71	105.571	114.0412	5.193e-11
Residuals	16	14.81	0.926		
Lack of fit	11	11.16	1.014	1.3883	0.3782
Pure error	5	3.65	0.731		

Direction of steepest ascent (at radius 1):

	cTime	cTemp	cCat
	0.88462865	0.05286616	0.46328968

Corresponding increment in original units:

	Time	Temperature	Catalyst
	4.4231432	0.2643308	0.2316448

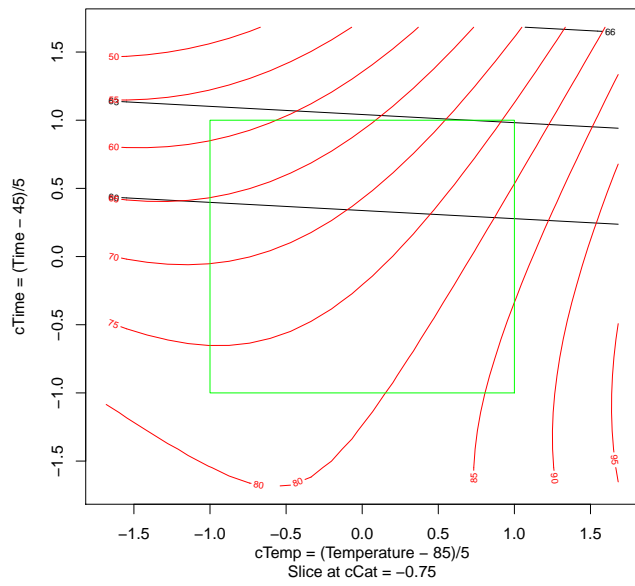
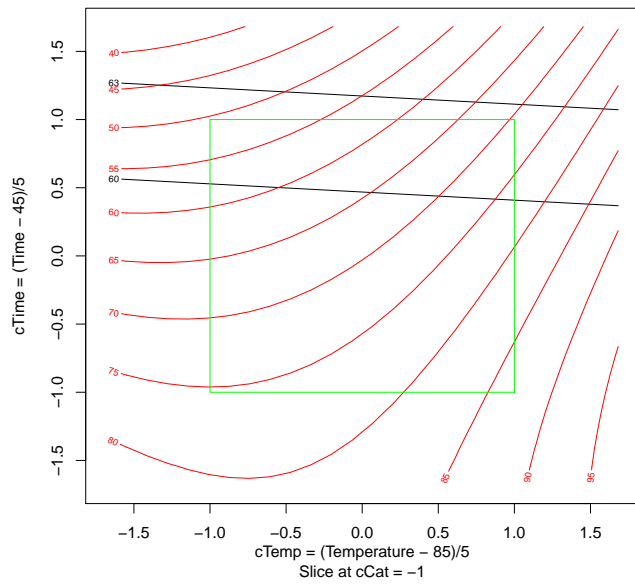
> #

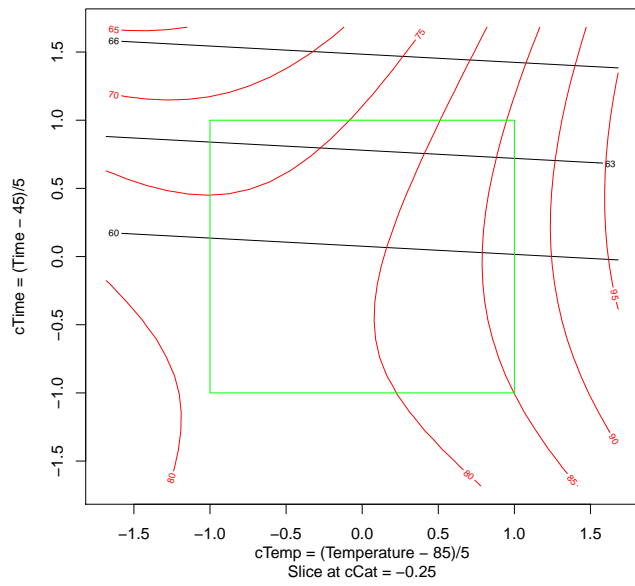
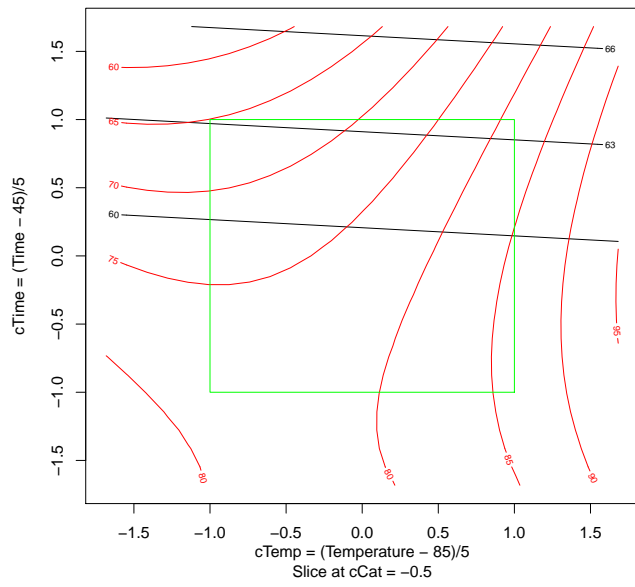
We have our models, now we work on the experimental goal. We are going to approach this graphically. For each level of cCat from -1 to 1 by increments of .25 we will:

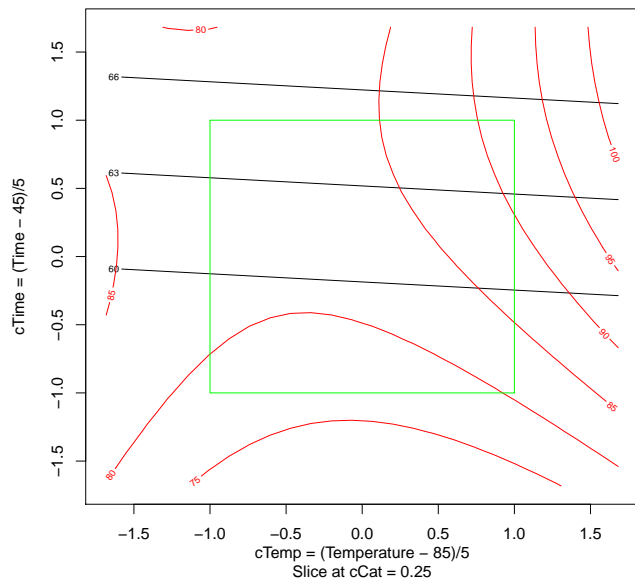
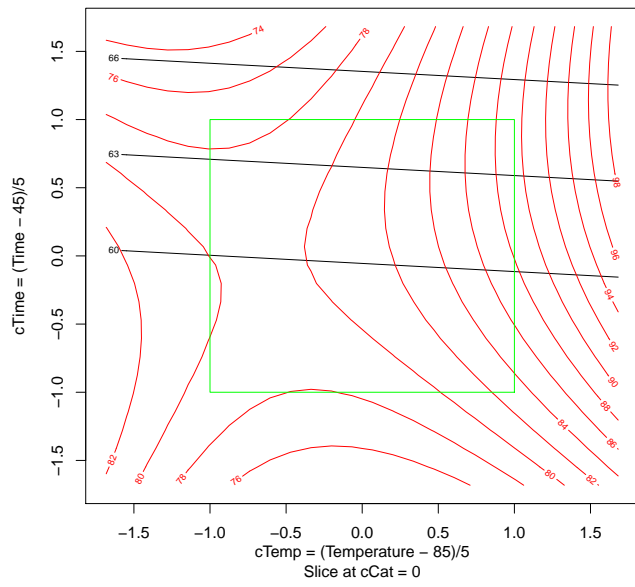
1. Make a contour plot of activity with contours at levels 60, 63, and 66 (the minimum, desired, and maximum levels);
2. Overlay a contour plot of conversion (red, to make it visible);
3. Overlay a box (in green) to show the experimental area.

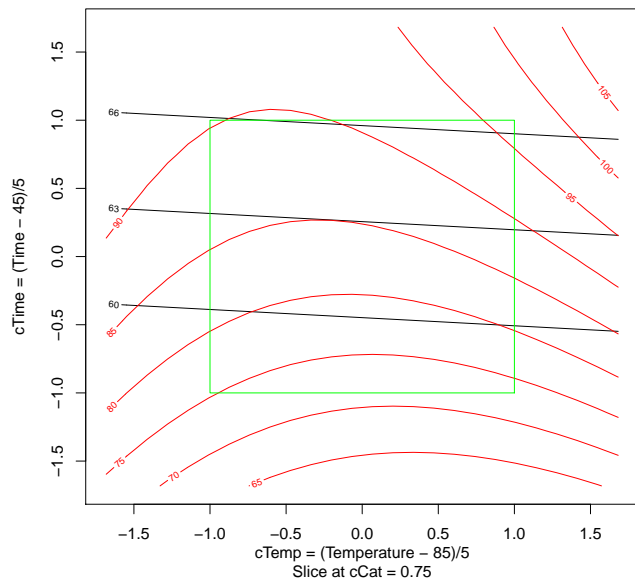
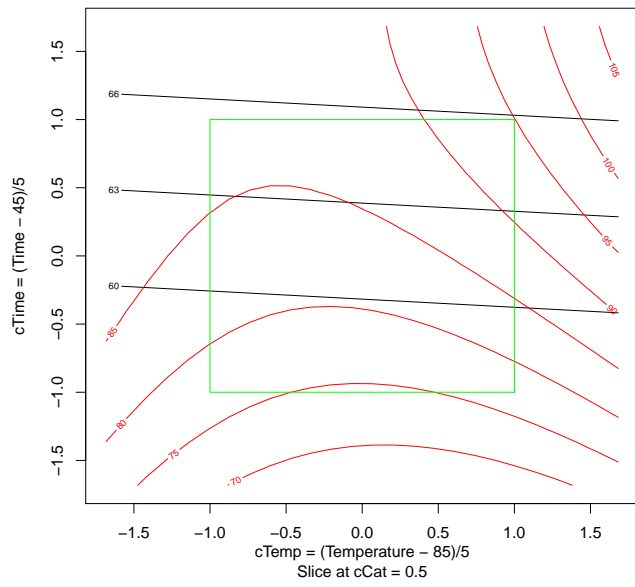
The goal is to walk through the images and find where we want to operate the system. This will involve some tradeoff between being as close as possible to 63 for activity and having the highest conversion.

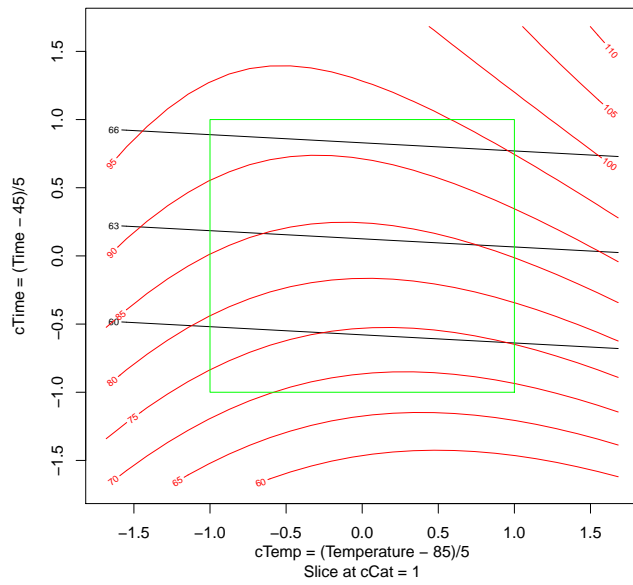
```
> for(cc in seq(-1,1,.25)) {
  contour(fit2b,cTime~cTemp,at=list(cCat=cc),levels=c(60,63,66))
  contour(fit1,cTime~cTemp,at=list(cCat=cc),add=TRUE,col="red")
  lines(c(1,1,-1,-1,1),c(-1,1,1,-1,-1),col="green")
}
```











> #

We can get 63 activity and conversion of 90% or better when cCat is about .25 to .5. However, we are right on the border for Temperature. Alternatively, if we want a high conversion with any activity from 60 to 66, then we can get just a bit over 95% with cCat at .75; again, we're on the border for temperature.