

Recall example 1 2^3 experiment (2 runs, n=16)

A: temperature

B: concentration

C: pH

Now: can only do 4 exp. (half fraction)

$$ABC = 1$$

Note $A = BC$ \leftarrow column for A = column for BC

$$B = AC$$

"aliasing"

$$C = AB$$

Example 2^{4-1} design

4 factors, $2^4 = 16$ factor combinations

$$2^{4-1} \rightarrow 2^3 = 8 \text{ experiments}$$

Effects: A B C D

AB AC AD BC BD CD

ABC ABD BCD

ABCD

only do experiments with $ABCD = 1$

which columns will be equal?

$$ABCD = 1$$

$$\underbrace{A \cdot}_{\pm} \underbrace{ABCD}_{A} = A$$

$$\underbrace{BCD}_{\pm} = A$$

$$\underbrace{B \cdot}_{\pm} \underbrace{BCD}_{A} = B \cdot A$$

$$= 1$$

$$CD = BA$$

$$\text{And } \begin{array}{l} B \cdot ABCD = B \\ ACD = B \end{array} \quad | \quad \begin{array}{l} ACD \cdot D = BD \\ AC = BD \end{array} \quad \text{etc.}$$

Here main effects aliased with 3-factor interactions
 2-factor int. aliased with 2-factor int.

Consequence?

\hat{A}^* : estimate of main effect in fraction $ABCD = 1$

$$\hat{A}^* = \frac{1}{8} \sum_{\substack{A=1 \\ ACD=1}} y - \frac{1}{8} \sum_{\substack{A=-1 \\ ACD=1}} y$$

\hat{A} and \hat{BCD} : estimates in full factorial

$$\hat{A} + \hat{BCD} = \left(\frac{1}{16} \sum_{A=1} y - \frac{1}{16} \sum_{A=-1} y \right) + \left(\frac{1}{16} \sum_{BCD=1} y - \frac{1}{16} \sum_{BCD=-1} y \right)$$

$$= \left(\frac{1}{16} \left[\sum_{\substack{A=1 \\ BCD=1}} y + \sum_{\substack{A=1 \\ BCD=-1}} y \right] - \frac{1}{16} \left[\sum_{\substack{A=-1 \\ BCD=1}} y + \sum_{\substack{A=-1 \\ BCD=-1}} y \right] \right)$$

$$+ \left(\frac{1}{16} \left[\sum_{\substack{BCD=1 \\ A=1}} y + \sum_{\substack{BCD=1 \\ A=-1}} y \right] - \frac{1}{16} \left[\sum_{\substack{BCD=-1 \\ A=1}} y + \sum_{\substack{BCD=-1 \\ A=-1}} y \right] \right)$$

$$= \frac{2}{16} \sum_{\substack{A=1 \\ BCD=1}} y - \frac{2}{16} \sum_{\substack{A=-1 \\ BCD=-1}} y = \hat{A}^*$$

$$\text{So } \hat{A}^* = \hat{A} + \hat{BCD}$$

$$\text{if } \hat{BCD} \approx 0 \text{ then } \hat{A}^* = \hat{A}$$

In the 2^{4-1} experiment A is aliased with BCD (etc),
but OK if $BCD \approx 0$.

Example 3 : Drug combinations

2^{6-1} fractional factorial

Main effects $A \ B \ C \ D \ E \ F$

Defining relation $ABCDE = F$ ← main effect
aliased with 5-factor int.

which effects are aliased?

$$A \cdot ABCDE = A \cdot F$$

$$BCDE = AF \quad \leftarrow \begin{array}{l} \text{2-factor int. aliased with} \\ \text{4-factor int.} \end{array}$$

$$B \cdot BCDE = B \cdot AF$$

$$CDE = BAF \quad \leftarrow \begin{array}{l} \text{3-factor int. aliased with} \\ \text{3-factor int.} \end{array}$$

(etc.)

Assumption 4-factor, 5-factor and 6-factor
effects ≈ 0

↳ can use the 2^{6-1} experiment with
defining relation $ABCDE = F$ to estimate
main, 2-factor int., and pairwise sums of 3-factor int.

Resolution (see def. on slide)

1-factor aliased with 5-factor

$$p=1$$

$$R-p = 5 \Rightarrow R=6$$

2-factor aliased with 4-factor

$$p=2$$

$$R-p = 4 \Rightarrow R=6$$

3-factor aliased with 3-factor

$$p=3$$

$$R-p = 3 \Rightarrow R=6$$

What if defining relation was $ABC = 1$

$$A \cdot ABC = A$$

$$BC = A$$

1-factor aliased with 2 factor

$$p=1$$

$$R-p = 2$$

$$\Rightarrow R=3.$$

[Results, note center runs]

Sums of Squares

Recall $SST = SSR + SSE$

$$SST = \sum (y - \bar{y})^2 \quad SSE = \sum (y - \hat{y})^2$$

$$SSR = n \hat{\beta}_1^2 + \underbrace{n \hat{\beta}_2^2}_{\substack{\uparrow \\ \text{2-level} \\ \text{factorial}}} + \dots$$

2-level
factorial

\downarrow
Variability explained by
factor B in the model.

[note ABC† DEF]

Factor D: $n \hat{\beta}_2^2 = 32 \cdot (-0.141)^2 = 0.636\dots$

$$\frac{n \hat{\beta}_2^2}{SST} = 0.68 \rightarrow 68\% \text{ variability explained by factor D.}$$

$$\frac{n \sum_{\text{main effects}} \hat{\beta}_j^2}{SST} = 0.815$$

$$\frac{n \sum_{\substack{\text{2-factor} \\ \text{int}}} \hat{\beta}_j^2}{SST} = 0.068$$

$$\frac{n \sum_{\text{3-factor}} \hat{\beta}_j^2}{SST} = 0.032$$

$$\frac{SSE}{SST} = 0.083$$

Plots :

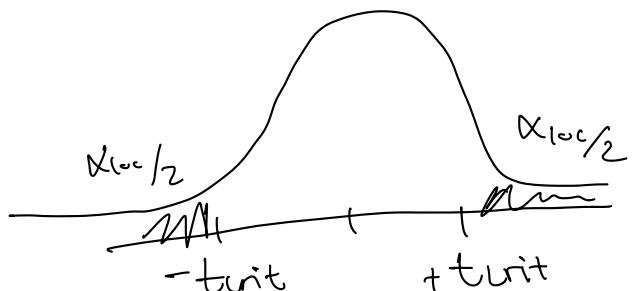
main effects : pH greatest effect,
then conc., then temp.

Int. effects :

Pareto :

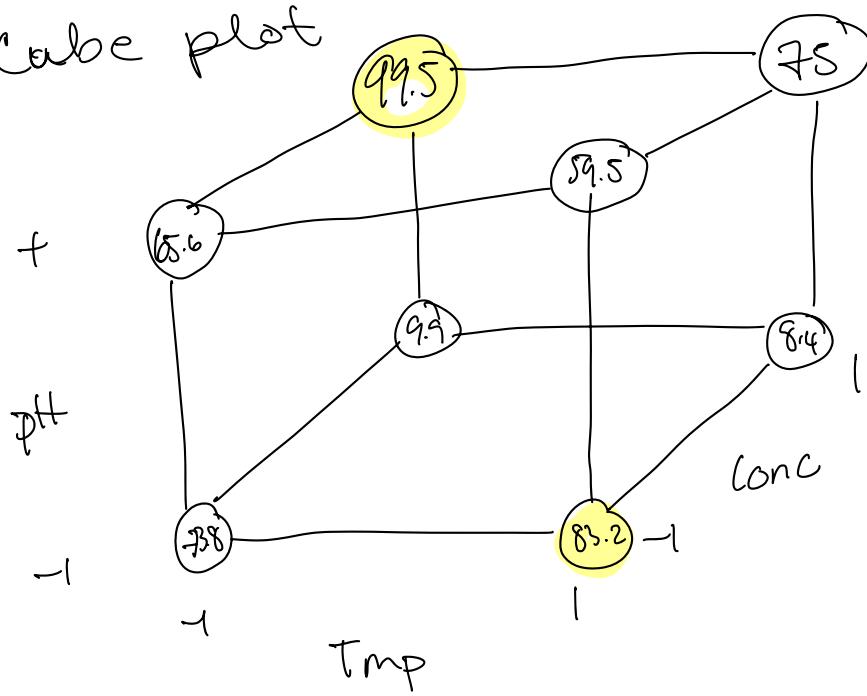
add critical value

$$T\text{-test} \quad \frac{\hat{\beta}_j}{\hat{s}(\hat{\beta}_j)} \stackrel{H_0}{\sim} t_{n-p} \quad 16^{-8}$$



$$\text{critical } |\hat{\beta}_j| \rightarrow t_{\text{crit}} \cdot \hat{s}(\hat{\beta}_j) \cdot 2$$

Cube plot



means for
each combination

shows which
factor comb.
are favorable.

same sign.

tmp -1, pH +1, conc +1

we see interaction pH · conc here

tmp -1, pH -1, conc -1