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CS 147: omputer Systems Performance Analysis Higher Designs and Other Considerations

CS 147: Computer Systems Performance Analysis Higher Designs and Other Considerations

Overview

Larger Designs

Price of More Levels Extending Confounding Fractionating Using Confounding Algebra Example Higher and Mixed Levels

Block Designs

Informal Methods

Other Considerations

Record-Keeping Randomization of Experimental Order Digression on PRNGs Types of Randomization



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The Price of More Levels



Using more levels increases no. of runs
 50% for raising a single parameter from 2 to 125% for two parameters
Extra runs could be used in other ways
 Examine more parameters Reduce variance (more replications)
Extra levels complicate experimentation

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- Using more levels increases no. of runs
 - 50% for raising a single parameter from 2 to 3
 - 125% for two parameters
- Extra runs could be used in other ways
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Deciding to Use More Levels



Must balance cost of extra runs against extra informatio Is response likely to be nonlinear? Are extreme resonases sufficient?

- Must balance cost of extra runs against extra information gained
- Is response likely to be nonlinear?
- Are extreme responses sufficient?
- Does curve have minimum/maximum between extremes?

Extending the Algebra of Confounding



- Standard 2-level confounding algebra is based on exponentiation modulo 2: A² = A⁰ = I
- This is trivial to extend to level *n*: $A^{x}A^{y} = A^{(x+y) \mod n}$

Rules of Extended Confounding Algebra



- First letter should have unit exponent
 - Achieved by raising to powers
 - Example (mod 3): $A^2B = (A^2B)^2 = A^4B^2 = AB^2$
- This works because of fractionating method
 - Sum of exponents modulo *n* is constant

Fractionating Using Confounding Algebra

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- Choose a confounding
- Divide experiment into blocks based on confounding
- Choose a particular block at random
- Execute experiments in random order

Using Confounding to Create 3^{k-p} Blocks



- Levels of each factor chosen from $\{0, 1, 2\}$
- Confounding exponents indicate multipliers
- Sum of multiplied levels modulo 3^p gives block number
- Example: AB^2 converts to $a + 2b \mod 3 = i$ where *i* is block number

Example of a 3^{2-1} Fraction

- Choose confounding: $I = AB^2$
- Divide combinations into blocks:

а	b	a + 2b	mod
0	0	()
0	1	2	2
0	2	1	l
1	0	1	l
1	1	()
1	2	2	2
2	0	2	2
2	1	1	
2	2	()

3



Running the 3^{2-1} Fraction



Choose block 2 by rolling dice
 Run combinations (a,b) = (0,1);(1,2);(2,0)
 Complete confoundings:

 I = AB²
 A = AB : B
 Cabulating effects is beyond scope of lecture
 Even in the aimple case

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Higher Prime Levels



•	Same algebra can be used for higher numbers of levels So long as prime
÷	Complexity rapidly becomes prohibitive
	 Normally use computers to do hard stuff
÷	Often simpler to use 2-level experiments
	 Figure out which effects are major
	Then use 1-factor tests to examine closely

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Mixed Levels



Factors may have different numbers of levels

 E.g., 2 CPUs, 3 dask drives, 4 memory sizes
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 - E.g., 2 CPUs, 3 disk drives, 4 memory sizes
- Possible to do fractional experiments here, too
 - Complexity is remarkable
 - No simple way to select fraction
 - Consult catalogs or software for fraction tables

Block Designs

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Block Designs

Causes of blocking
 Example of blocking
 Confounding between blocks
 Special types of blocks

- Causes of blocking
- Example of blocking
- Confounding between blocks
- Special types of blocks

Causes of Blocking

Physical constraints

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 Time constraints
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 Subject constraints
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 Multiple dation and/m hardware

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- Physical constraints
 - *n* probes, m > n signals to measure
- Time constraints
 - Only n experiments per day, with other activity between
- Subject constraints
 - Need to install new hardware between runs
 - Multiple disks on each machine

Example of Blocking

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Need to run 2 benchmarks under 2 network loads
 Can only do 2 runs per day
 Other conditions may vary from day to day
 Which pair to do first?

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- Which pair to do first?

Confounding Between Blocks

Example of the problem:

Three different ways to divide runs:

Day 1:000100100011Day 2:101101111001

- > Each choice confounds something with the day effect:
 - Factor a (level takes a day to change)
 - Factor b (same)
 - Interaction (equal levels one day, unequal other)

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Special Types of Blocks

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- One factor varies within another
 E.o., 5 computers, each with one NIC from each of 3
- E.g. 5 computers, each with one NIC from manufacturers
 ANOVA is different here: watch out!

- Split plot
 - Group by one factor, vary others randomly
 - Useful when expensive to change that factor
- Nested or hierarchical
 - One factor varies within another
 - E.g., 5 computers, each with one NIC from each of 3 manufacturers
- ANOVA is different here: watch out!

Informal Methods

formal Methods

- Often only want best performance
 Can simply pick combination that does best
- Better choice: sort by performance
 Identify which factors are common to top entries
 Eliminate any that aren't consistent

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Record-Keeping Principles

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 Be able to reproduce any experiment
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- Never throw away data
- Be able to reproduce any experiment
- Parameterize your software
 - Don't create experiments by editing source
 - Leads to irreproducibility
- Use version control!

The Need for Randomization



- Uncontrollable parameters may vary during experimentation
- Plotting error vs. experiment number detects systematic variation
 - But doesn't control it
- Randomization controls the problem
 - Turns it into error parameter

Example of External Trends

- Consider measuring disk performance:
 - Benchmark creates 1000 small files, 10 large ones, writes them, then deletes them
 - File size is varied as experimental parameter
 - One run takes several hours
 - Other people use system daily



Slide contains animations.

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- Consider measuring disk performance:
 - Benchmark creates 1000 small files, 10 large ones, writes them, then deletes them
 - File size is varied as experimental parameter
 - One run takes several hours
 - Other people use system daily
- Disk fragmentation may increase over time, changing results



Slide contains animations.

(Pseudo-)Random Number Generation



- Not all PRNGs are equal
- Most common is linear congruential
 - E.g., rand, random, drand48
 - Don't use low bits (modulo) to adjust range
 - Best to use floating result in [0, 1) and multiply by range
- Prefer longer periods
 - E.g., Mersenne Twist (see Google)

Random Seeding

- Risky to use "random" seeds (e.g., /dev/random)
 - Irreproducible (uncheckable) results
 - Risk of getting into middle of sequence from different experiment
 - Produces correlation
 - Especially with linear congruential
- Better to use different parts of long sequence
 - Note that many PRNGs only take a 32-bit seed
 - \Rightarrow Only 2³² different sequences
 - If period is 2³², then you're always diving into the middle of a sequence
- In any case, remember your seeds



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In any case, remember your seed

Complete Randomization

- Plan experiment first
 - Levels of each parameter
 - Number of replications
- List experiments by levels and replication number
- Choose experiments from list randomly
 - Use selection without replacement
 - Equivalently, shuffle list and use shuffled order



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Constrained Randomization



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- Complete randomization sometimes impossible
 - One experiment might destroy others
 - Lengthy setup times
 - Need to send computer back to manufacturer
- Must divide experiments into blocks
 - Randomize within each block
- Block effect confounded with true effect