

study design vs. inference



A SOURCE OF CONFUSION

why do statistics?



- Make a compelling argument about the world.
- Connect what's going on in the world (data) with a deductive framework (inference).
- If we control the process of data collection then we can have a better sense of how our deductive framework should behave.
- This leads to clearer reasoning about what the underlying structure of the world might be.
- This lecture: What if we don't control the way the data were actually generated? What if we didn't intervene?

study design vs. inference



- 90% of statistics classes are about inference
- Why?
 - It's useful, getting you those confidence intervals and p-values.
 - The Math is pretty cool.
 - It feels hard.
 - Because many of us don't really know much about the real world...

design



RANDOMIZATION AND SAMPLING

where does the data come from?

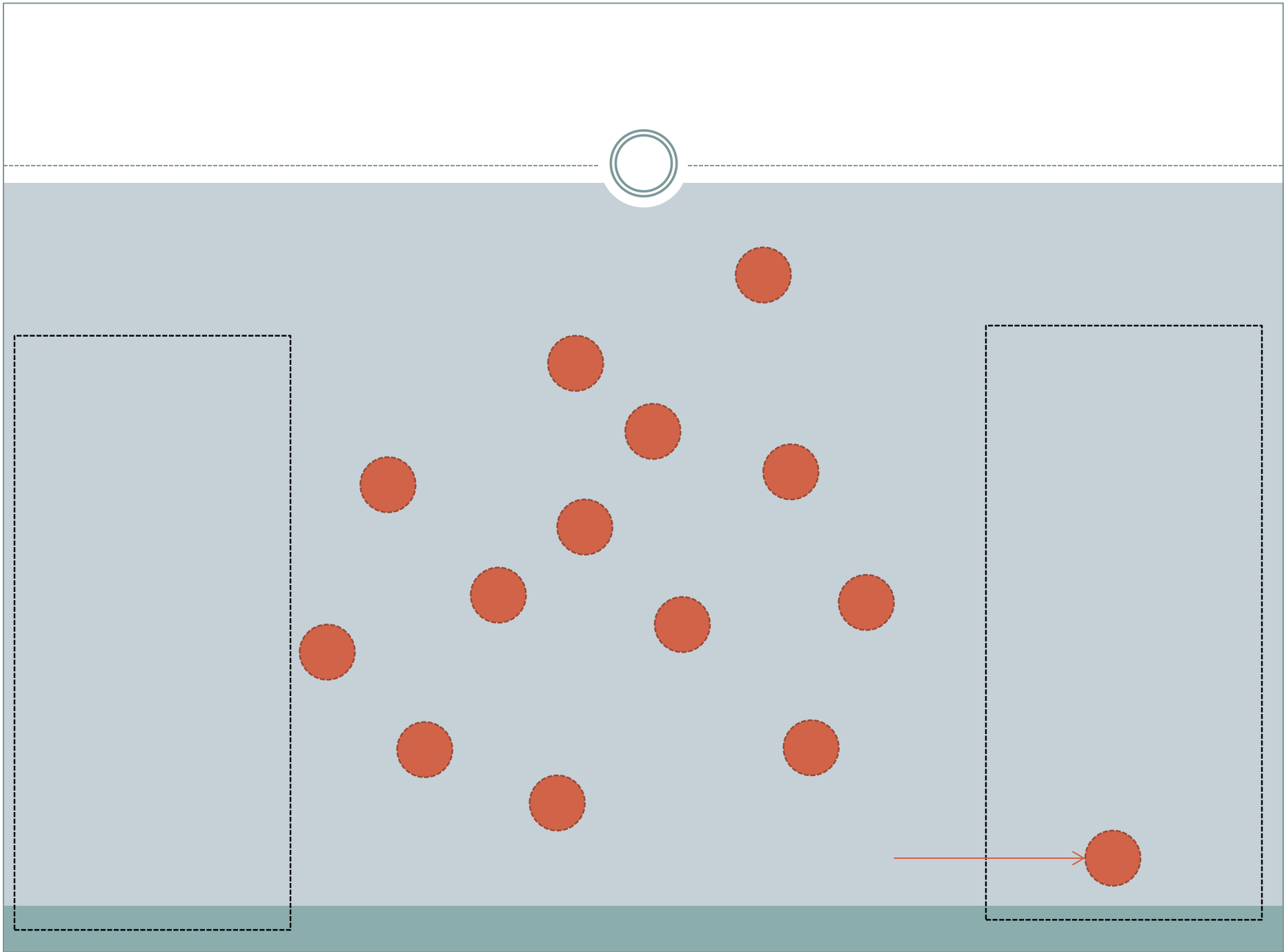


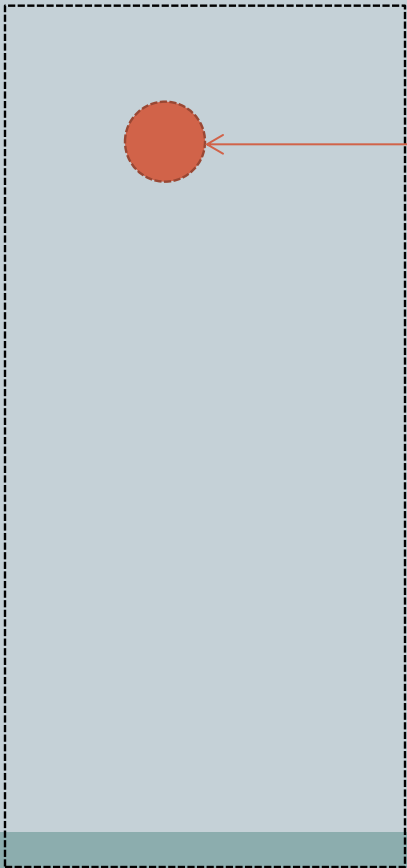
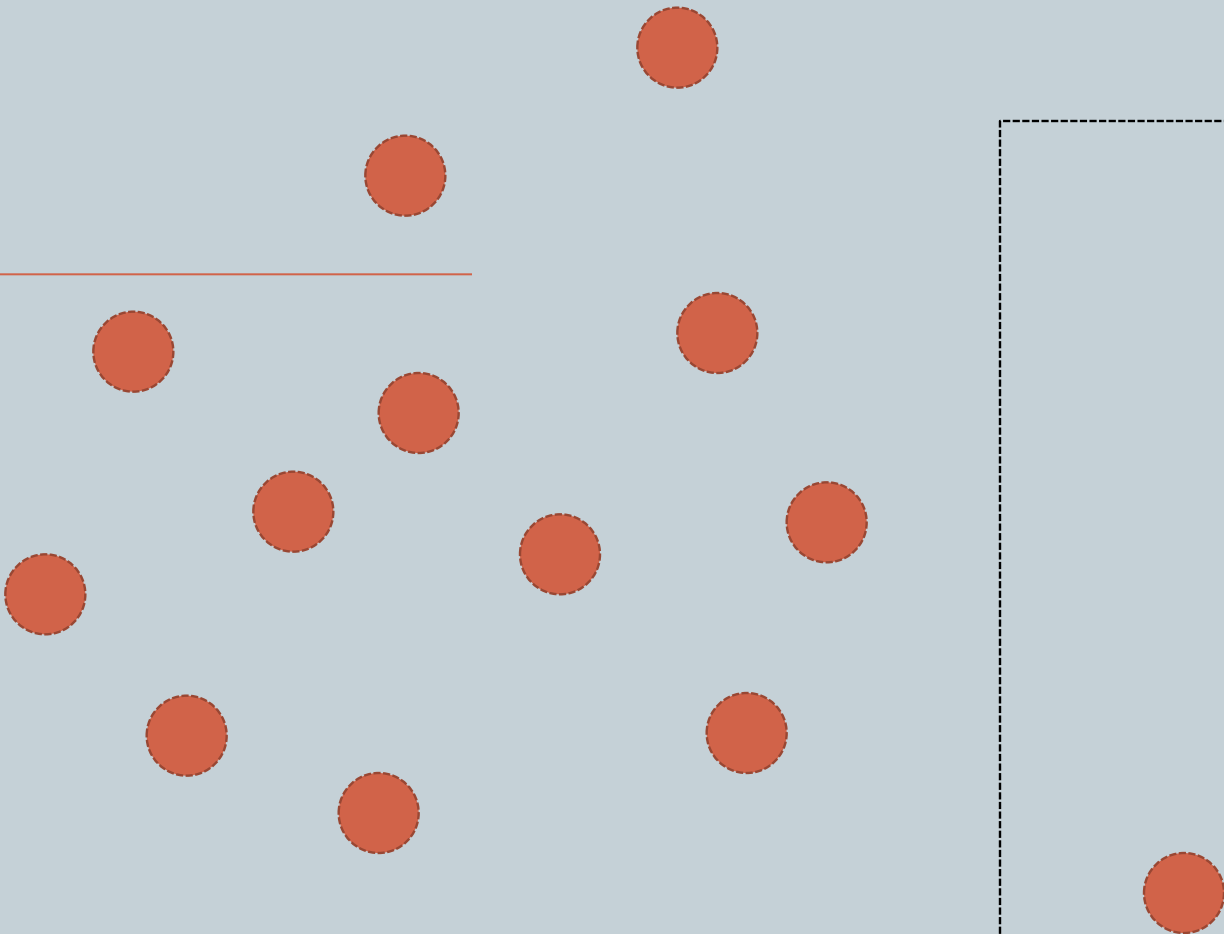
- We design trials.
 - Assign groups that are similar at baseline
 - Examine counterfactuals
- We also design surveys.
 - Representative groups
 - Understand population from subsets of those populations
- Both use elements of control and randomness

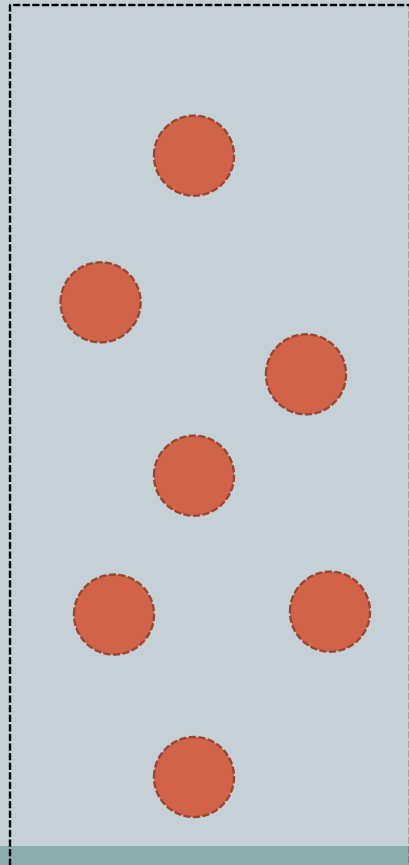
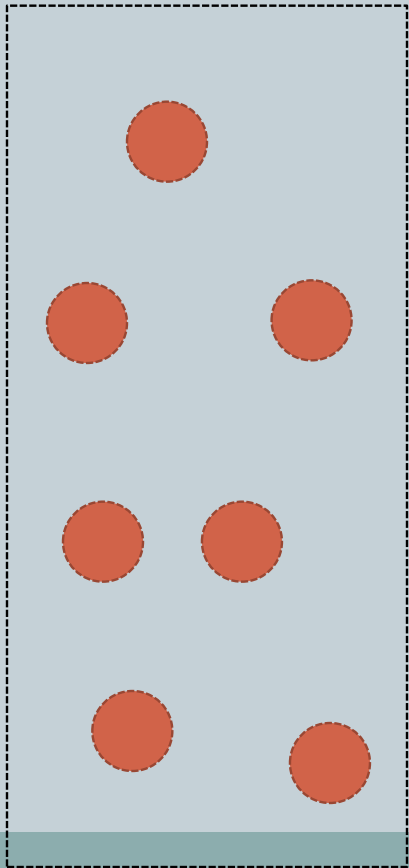
an example

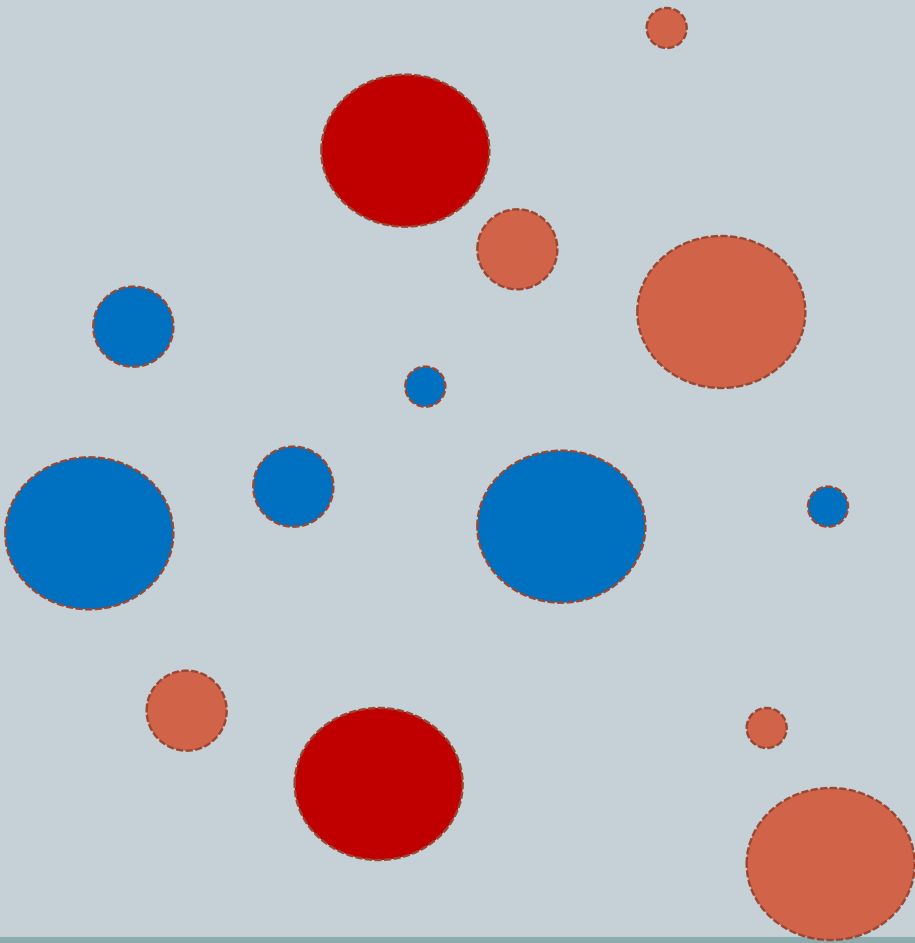
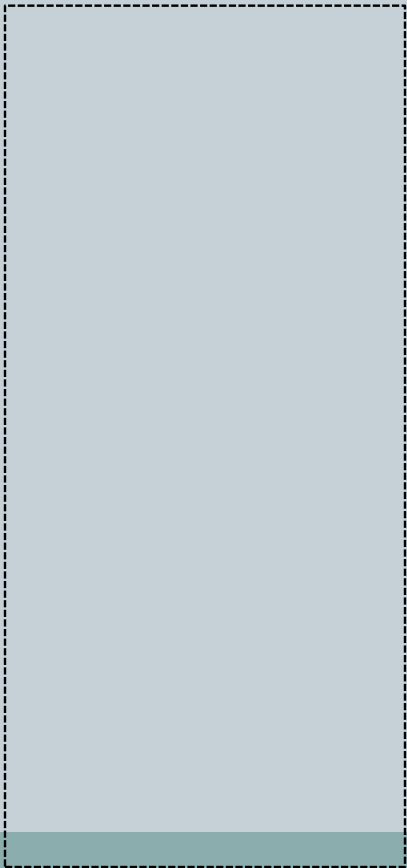


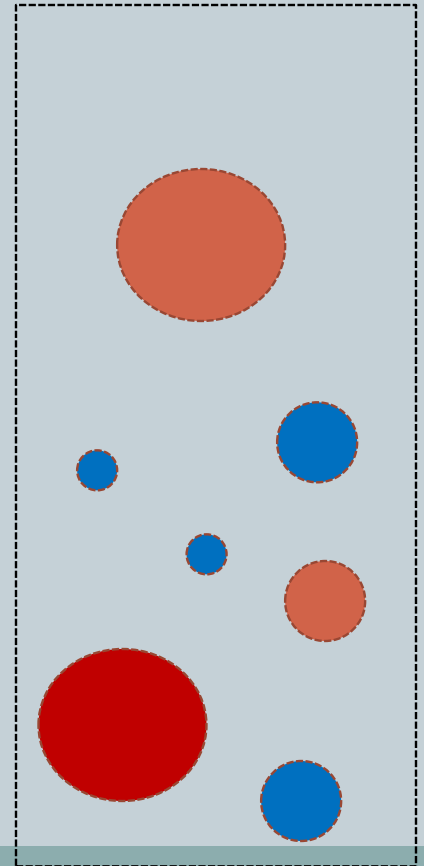
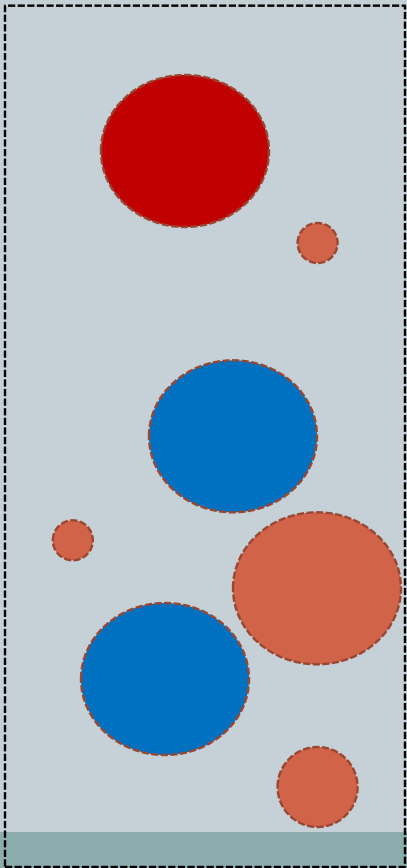
- Want to study a pill.
- Design the study
 - Uniform randomization
 - Matched pairs randomization
 - Crossover design
 - Cluster-randomized
- Inference
 - t-test
 - Matched-pairs t-test
 - Repeated measures model
 - Generalized linear mixed model
 - But... maybe all of those could be GLMM.

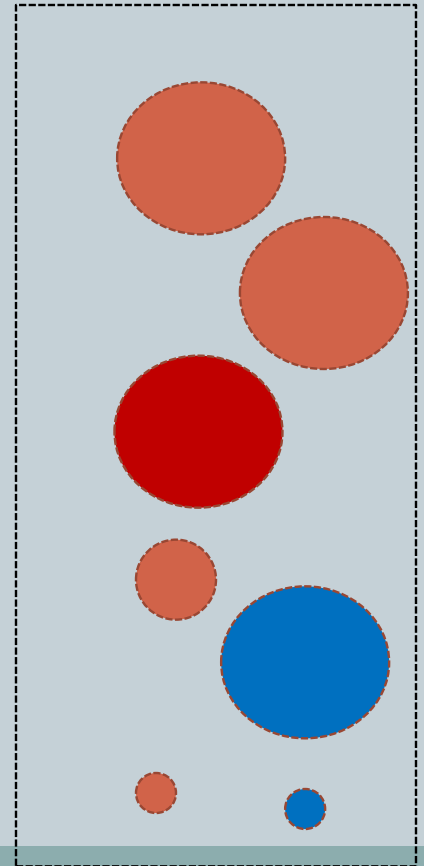
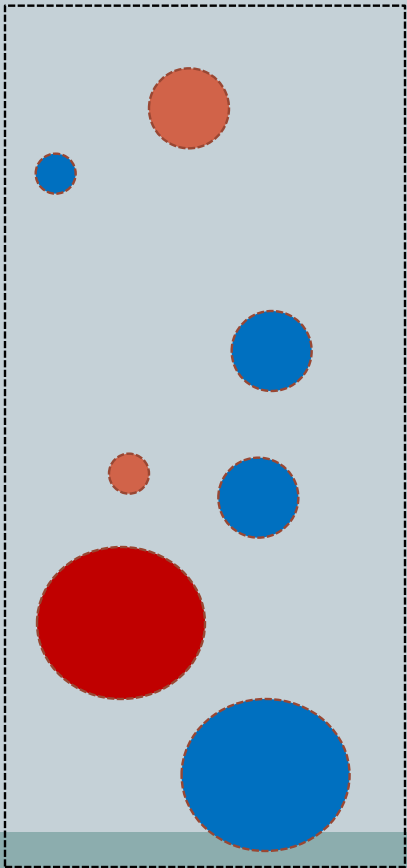


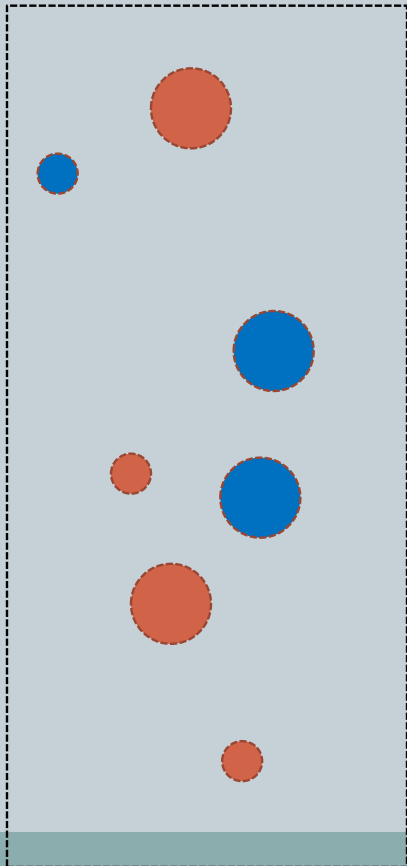




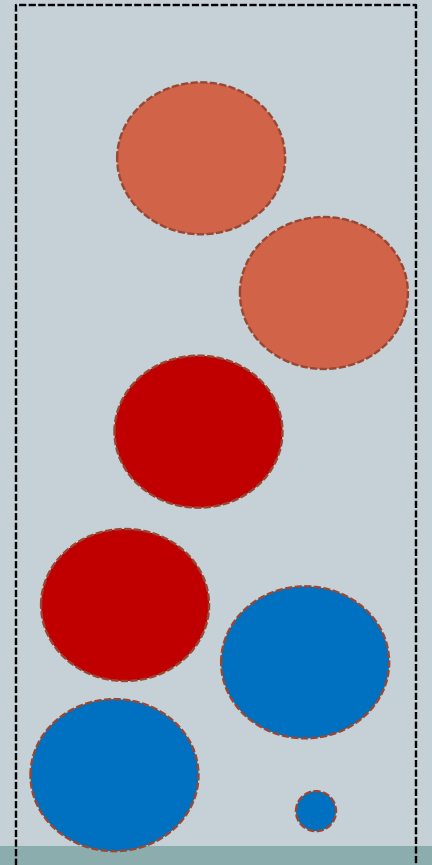


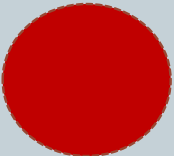
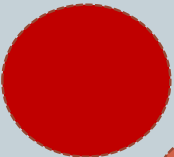
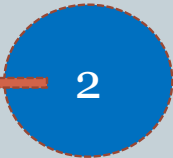
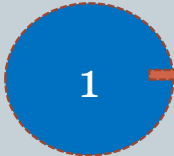
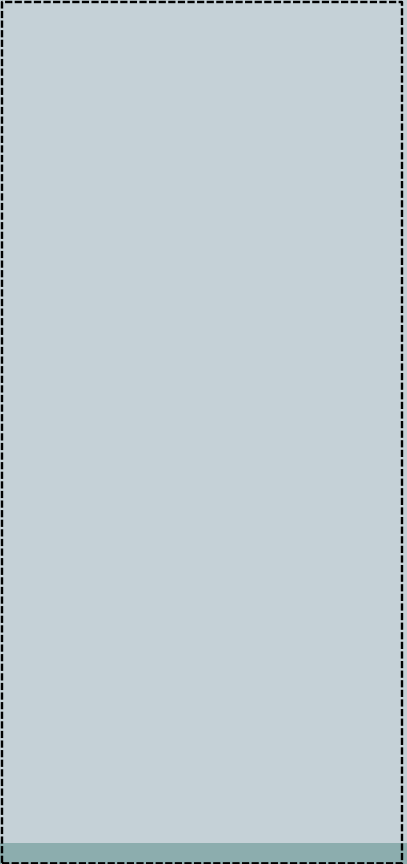
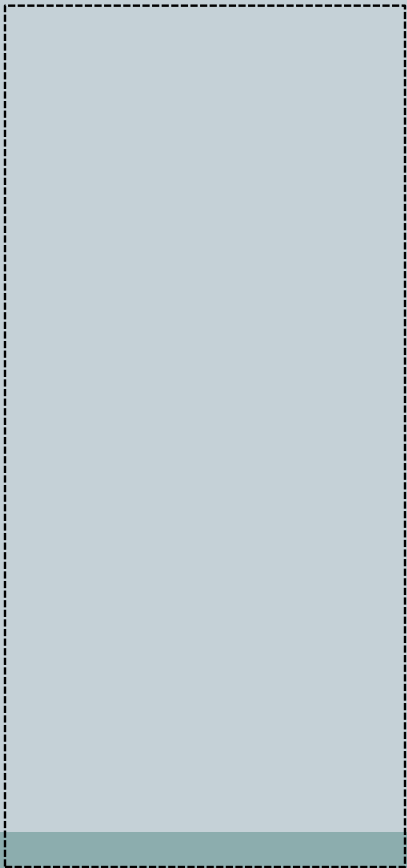


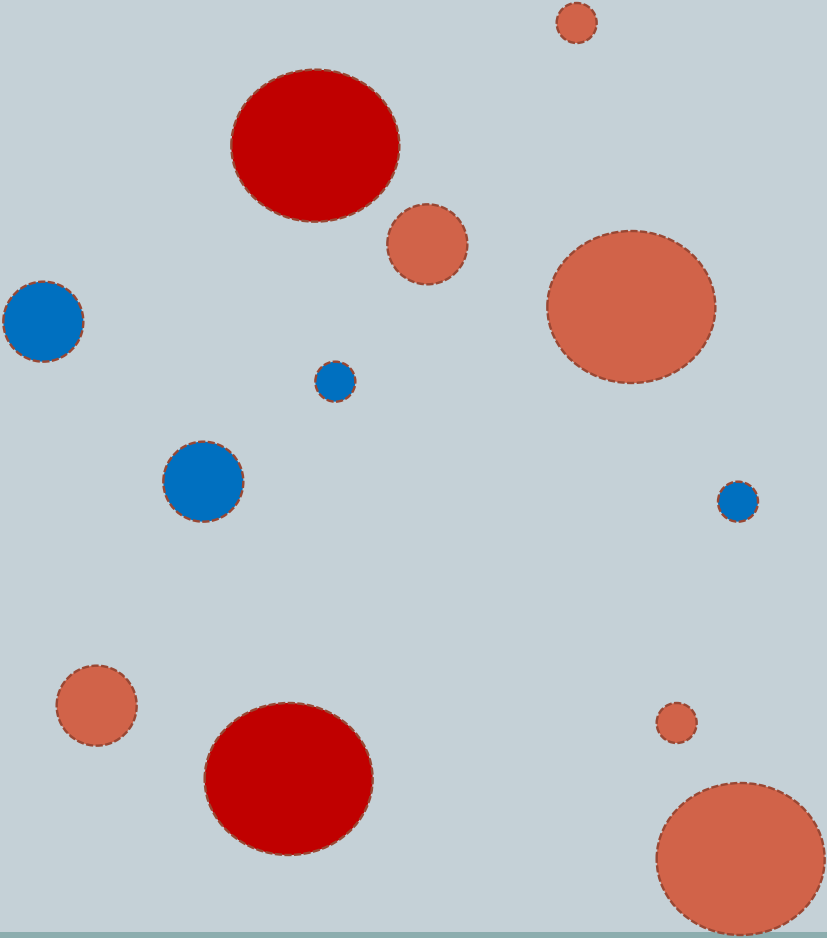
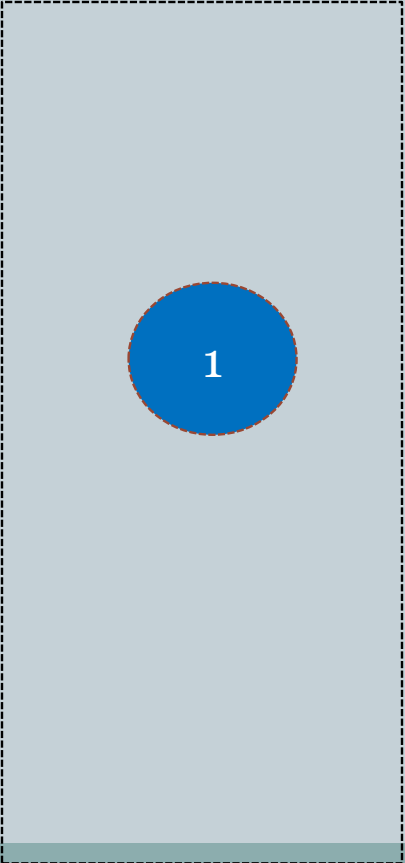
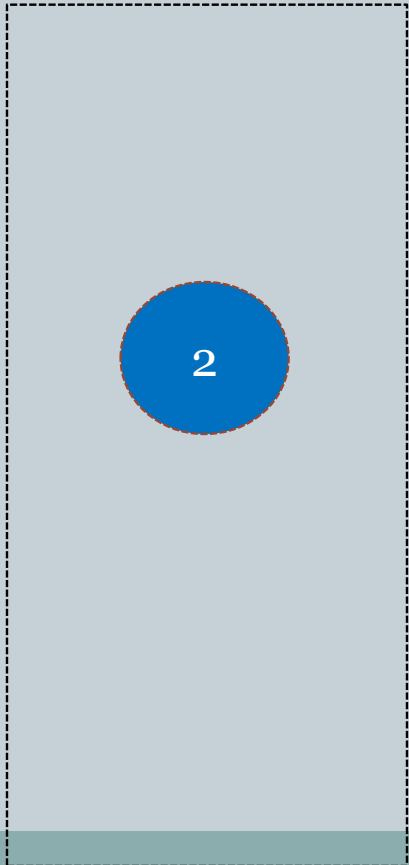


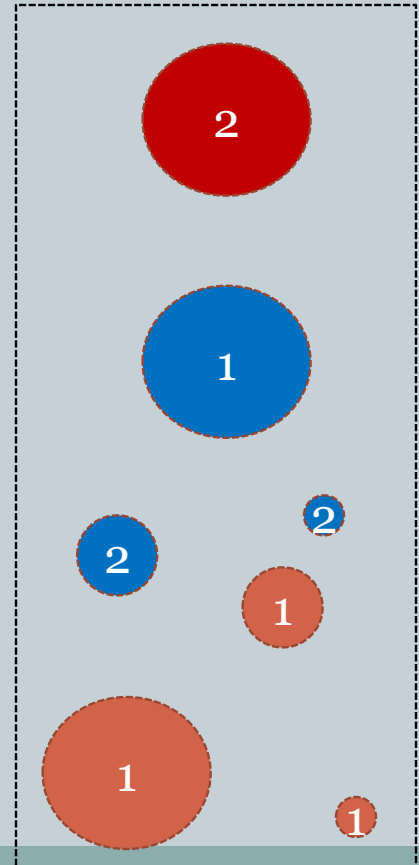
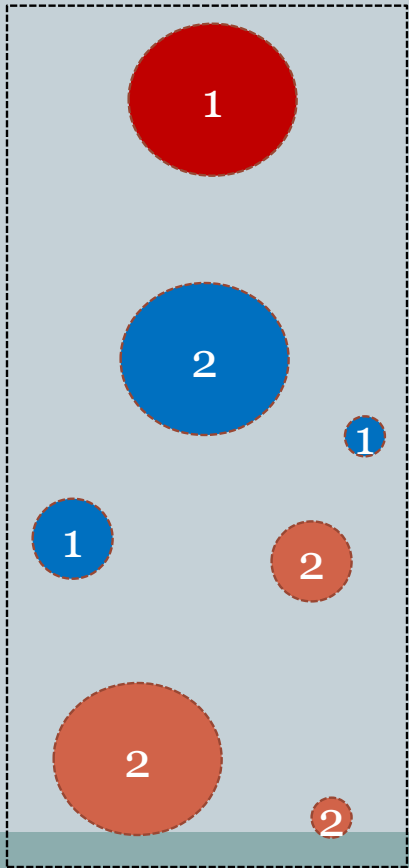


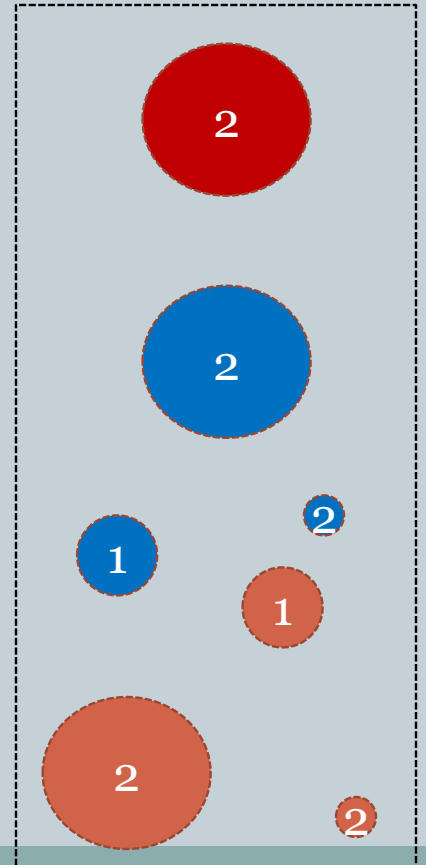
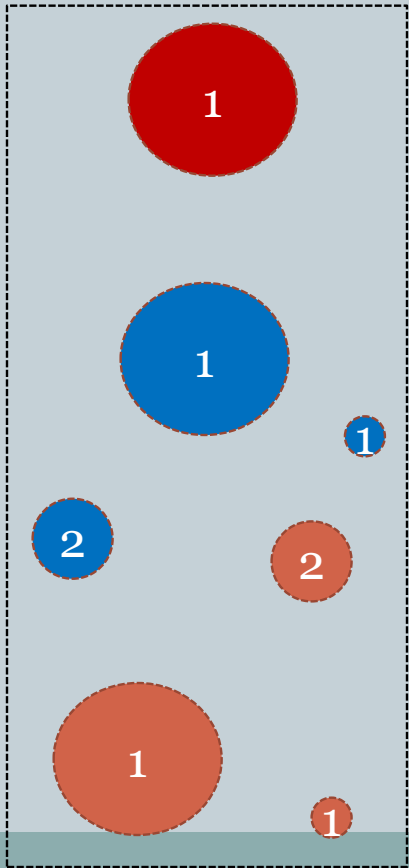
We're off to a **bad** start.











the design limits what could have happened



- Uniform randomization allows for quite different possibilities. $({}_nC_r)$
- Matched-pairs randomization limits the size, and range, of possible assignments. $(2^{\binom{n}{2}})$
- In some sense, we're losing something when we go to matched-pairs...
- ... but what are we losing? The “crazy” options that we know are going to lead us astray.

an example

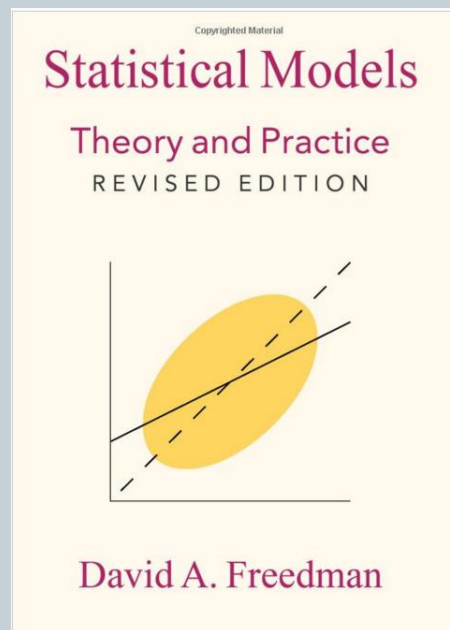


- Want to study an election.
- Design the study
 - Simple random sample
 - Proportional sampling
 - Snowball sampling
- Inference
 - t-test
 - Inverse probability weighting
 - Generalized linear mixed model
 - But... maybe all of those could be GLMM.

different beliefs about where data come from



- RCT and survey
- Structural equation modeling
 - $y_i = \beta_0 + \beta_1 x_{1,i} + \dots + \beta_p x_{p,i} + \varepsilon_i$
- If you want to be disabused of SEM spend some time reading



inference



picking inference



- Inference requires assumptions
- Linear regression:
 - Linearity and additivity
 - Independent errors
 - Homoskedasticity
 - Normality of errors
- Permutation test:
 - No interference
- “Fancier” methods tend to have more assumptions... and thus leave you open to more lines of attack.
- These attacks can be obviated by careful preparation during the design phase.

picking inference



- Use the simplest method that gets the job done.
- If you want to accomplish more, collect more data or do additional analyses. (“If have to use something more complicated than a t-test then someone messed up...”)

prospective study design



- A lot of the foundations have been worked out:
 - Experimental design
 - Survey sampling
- But, obviously, there are a lot of cool developments still going on:
 - Experimental design: adaptive trials, point-of-care randomization,
 - Sampling: active learning, explore exploit learning

observational (and retrospective) design



- This seems weird
- Usually a data set is in front of you, so you just analyze it
- It takes some thought to see this

Let's do an example.

observational study design



NEONATAL INTENSIVE CARE UNITS

Application: Regionalization



- Hospitals vary in their ability to care for premature infants.
- The American Academy of Pediatrics recognizes levels: 1, 2, 3A, 3B, 3C, 3D and Regional Centers.
- ***Regionalization of care*** refers to a policy that suggests or requires that high-risk mothers deliver at hospitals with greater levels of capabilities.

Outcome



Outcome



The data



- Every baby delivered in a 10+ year period
 - California
 - Pennsylvania
 - Missouri
- Mothers' information
 - ICD9 codes
 - ✦ Delivery
 - ✦ Post-delivery complications
 - ✦ Some pre-delivery
 - Some SES information
 - Zip code of residence
- Birth/death certificates
- Census information
 - PA and MO have zip code level
 - CA will have block group

Pre-delivery
Severity?

	Variable Type	High NICU	Low NICU	sd	Δ /sd
Mortality	Outcome	2.26%	1.25%	13.33%	0.08
Difference in Travel Time	Instrument	4.57	19.00	17.18	-0.84
% attending high level NICU	Treatment	100.0%	0.0%	49.7%	2.01
Birth weight	Preemie covariates	2,454.07	2,693.24	739.27	-0.32
Gestational age		34.61	35.69	2.80	-0.39
GI	% of preemies with type of congenital disorders	0.9%	0.6%	8.7%	0.04
GU		0.9%	0.8%	9.0%	0.01
CNS		0.9%	0.4%	8.3%	0.05
Pulmonary		0.8%	0.7%	8.8%	0.01
Cardio		1.4%	0.7%	10.5%	0.06
Skeletal		0.7%	0.9%	9.0%	-0.02
Skin		0.0%	0.0%	0.0%	0.00
Chromosomes		0.4%	0.3%	6.3%	0.02
Other_Anomaly		0.8%	0.1%	7.0%	0.09
Gestational_DiabetesM		Mother covariates	4.9%	4.3%	21.0%
Mother's education	3.76		3.58	1.19	0.16
Insurance - Fee for service	24.0%		24.5%	42.8%	-0.01
Insurance - HMO	32.3%		27.8%	46.0%	0.10
Insurance - Government	23.5%		24.2%	42.6%	-0.02
Insurance - Other	16.8%		21.4%	39.1%	-0.12
Uninsured	2.2%		1.6%	13.7%	0.04
Prenatal care	2.51		2.37	1.30	0.11
Single birth (y/n)	79.0%		86.1%	38.3%	-0.18
Parity	2.08		2.09	1.31	-0.01
Mother's age	28.41	27.71	6.25	0.11	
Median income	Census level covariates	41,484.25	40,258.92	14,587.24	0.08
Median home value		97,663.00	95,083.15	48,762.43	0.05
% completed high school		79.9%	80.0%	9.7%	-0.01
% completed college		22.2%	19.4%	13.1%	0.21
% renting		31.4%	27.9%	12.8%	0.28
% below poverty line		13.4%	11.8%	9.9%	0.16

Summary of Problem



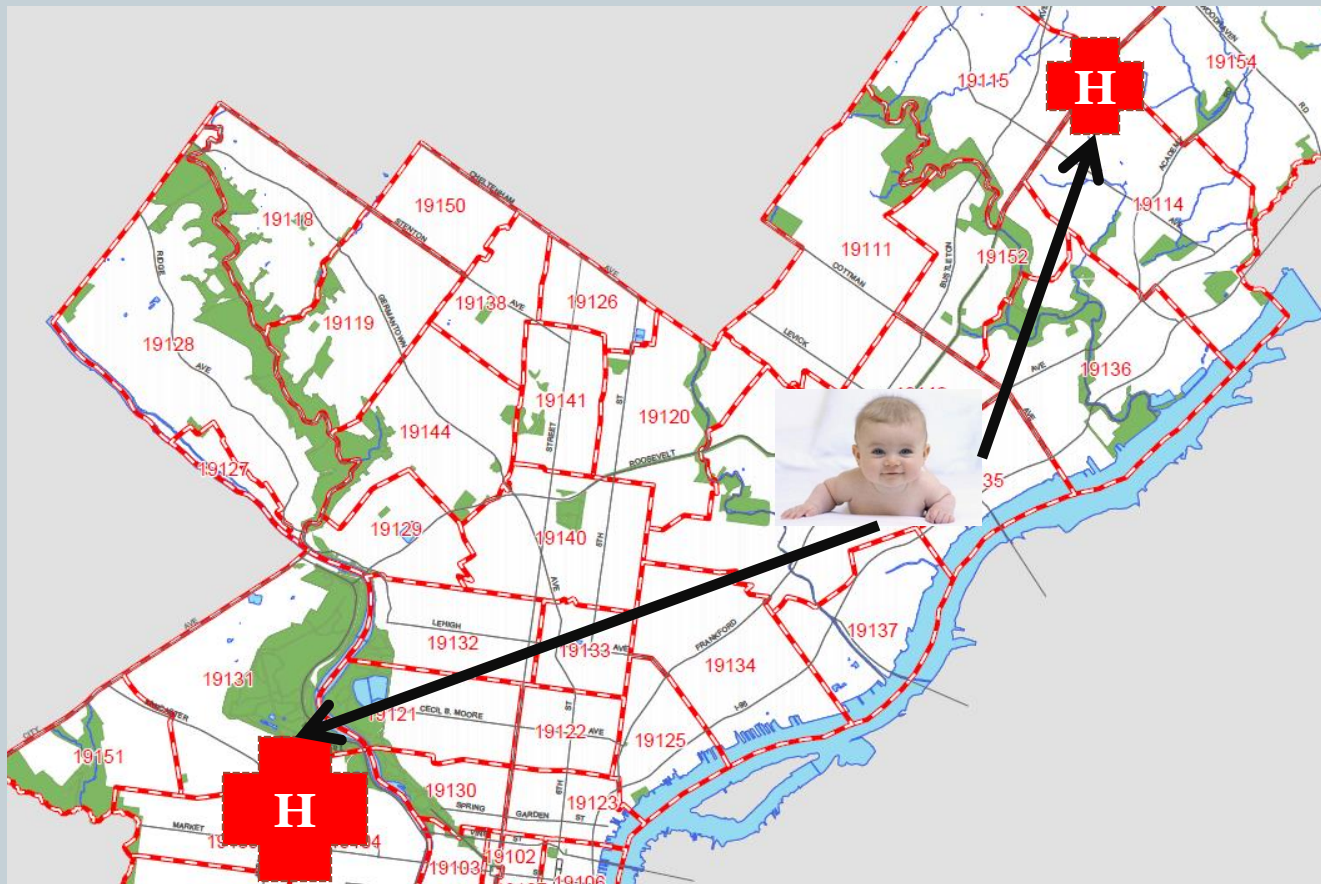
- Want to quantify effect of level of NICU on rate of death
- Observational data
- Sorting bias
- Some sorting variables are **unobserved**

real world randomness



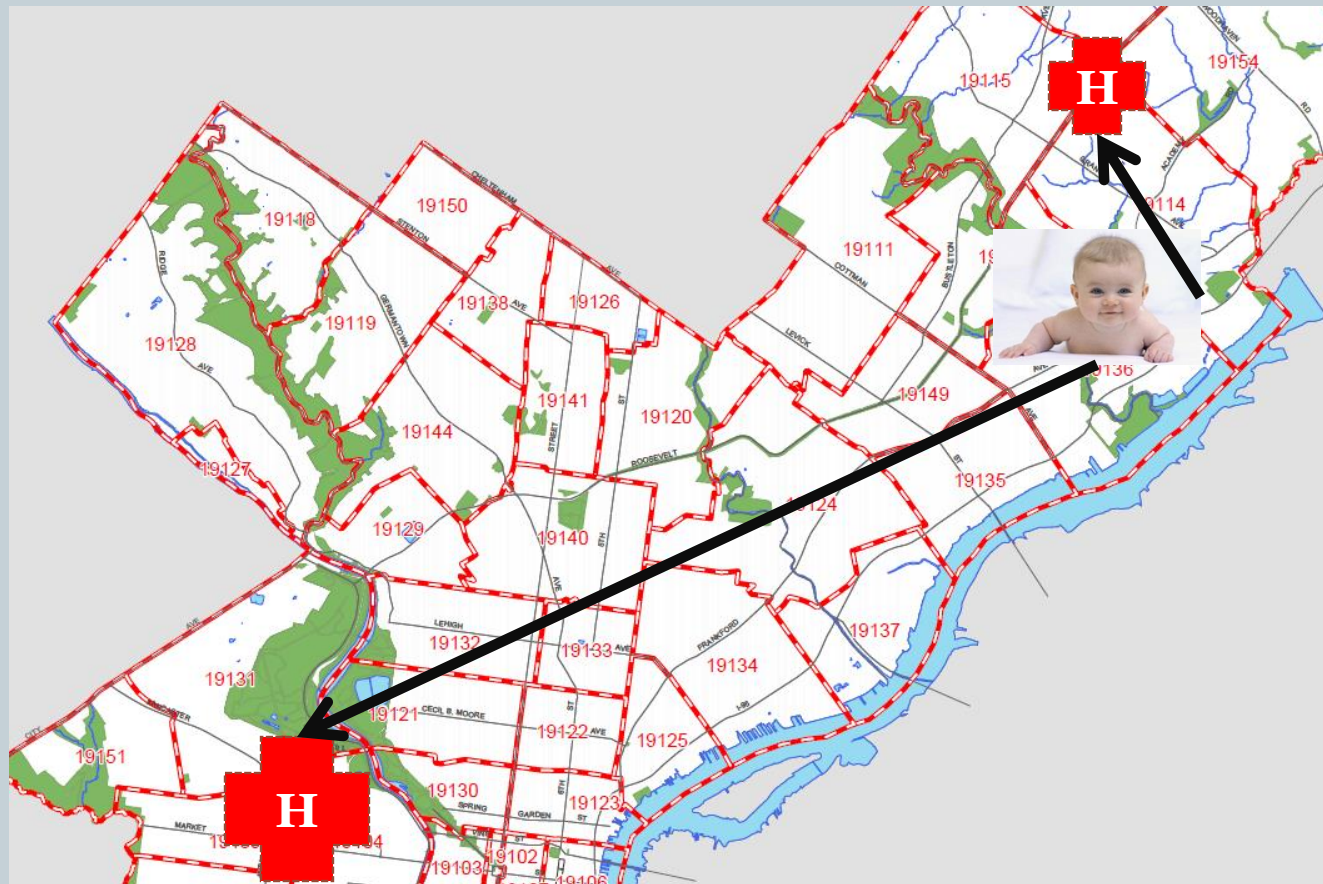
(M A Y B E)

Instrument: Excess Travel Time



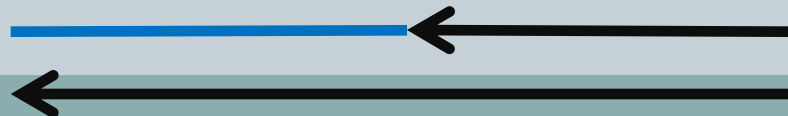
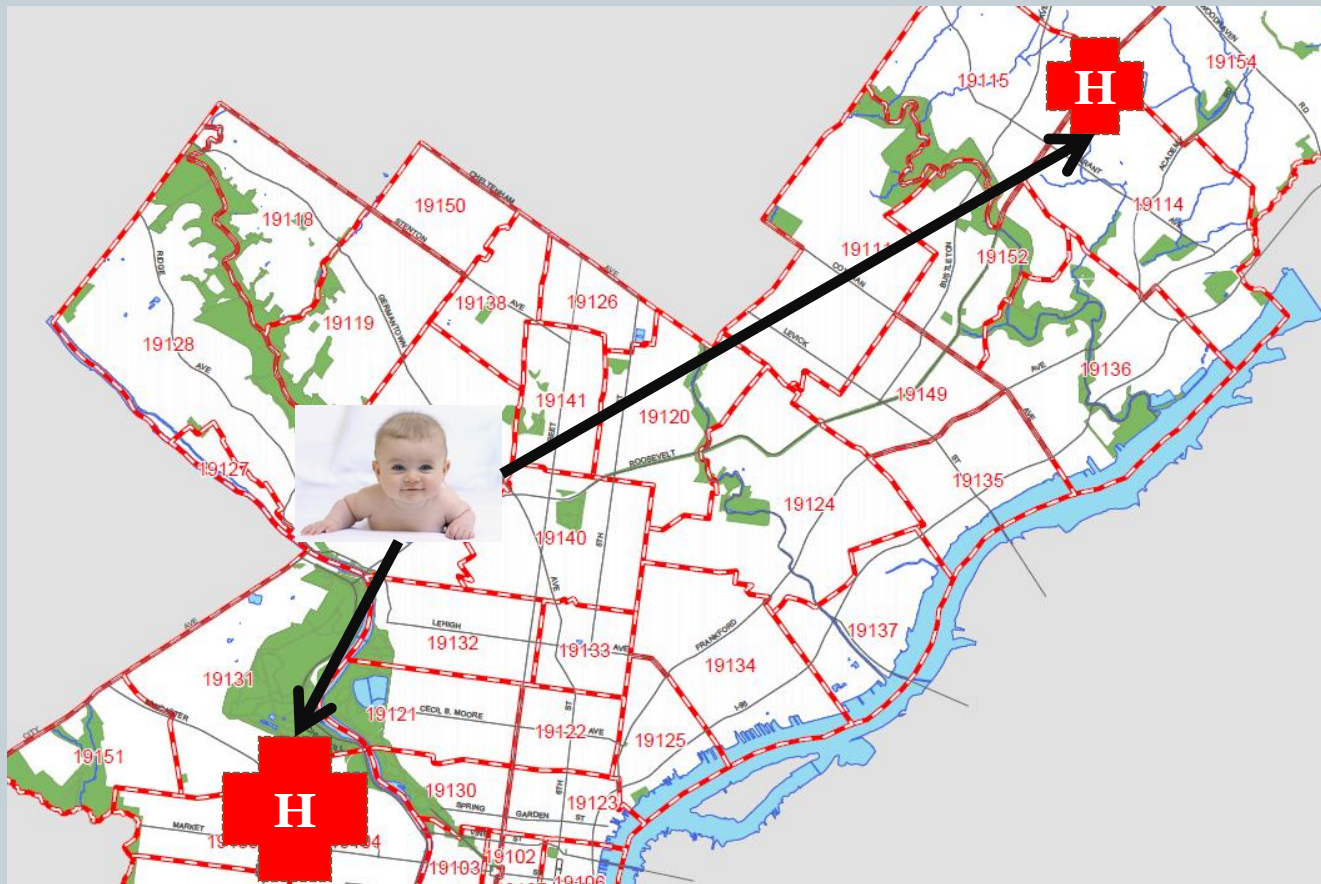
Excess Travel Time

Instrument: Excess Travel Time



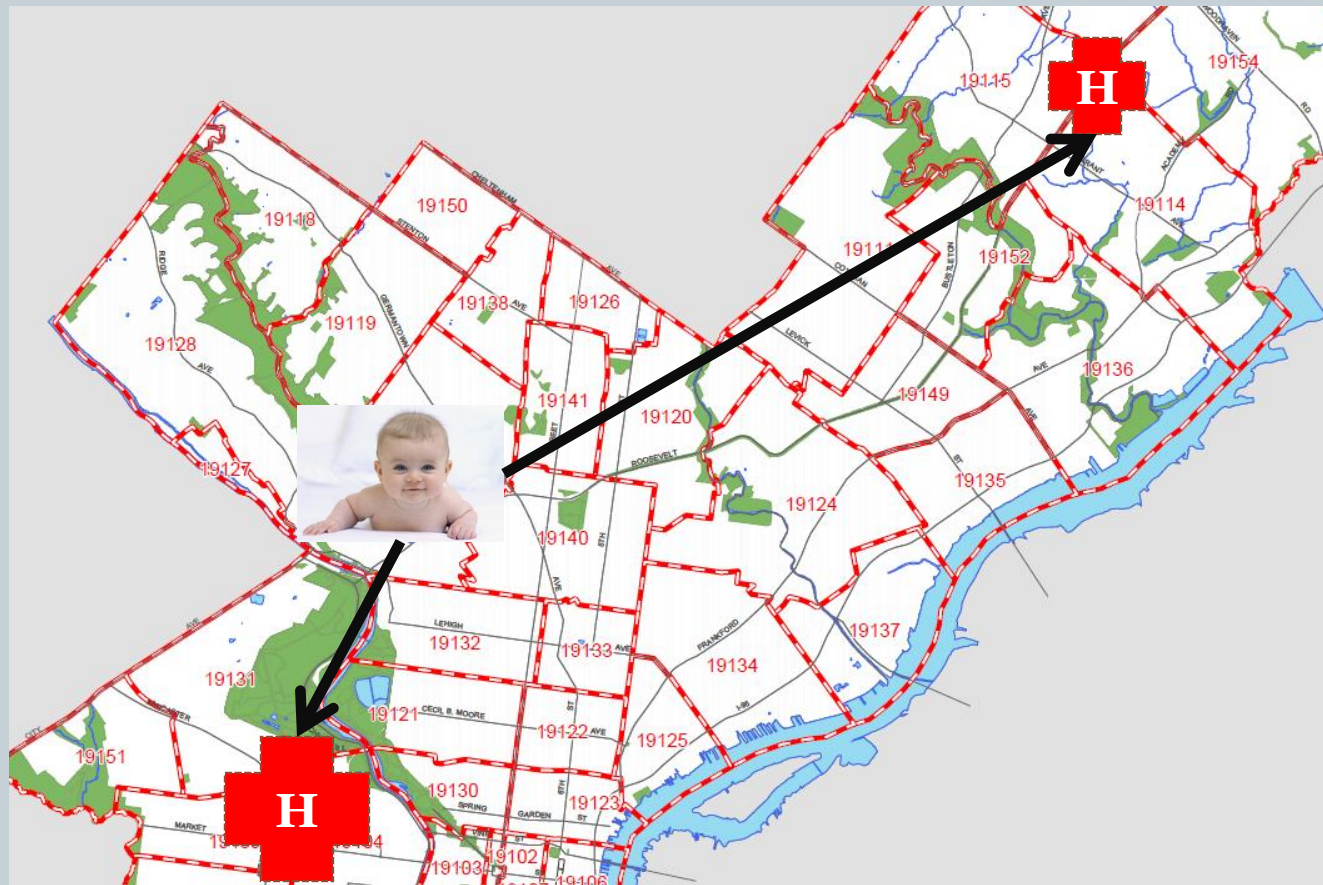
Excess Travel Time

Instrument: Excess Travel Time



Excess Travel Time

Instrument: Excess Travel Time



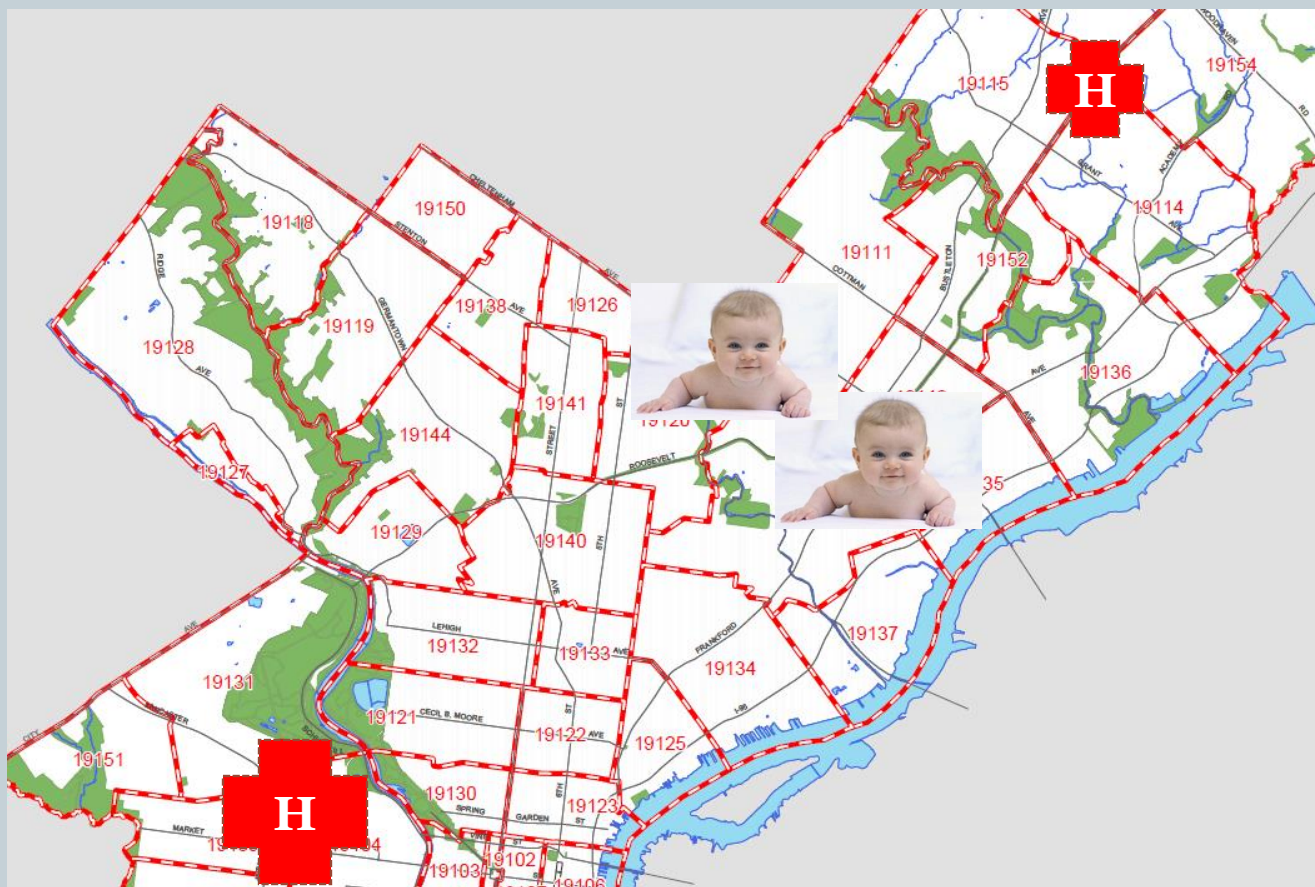
McClellan, McNeil & Newhouse; "Does more intensive treatment of acute myocardial infarction reduce mortality?" *JAMA*. 272(11): 859-66, September 1994

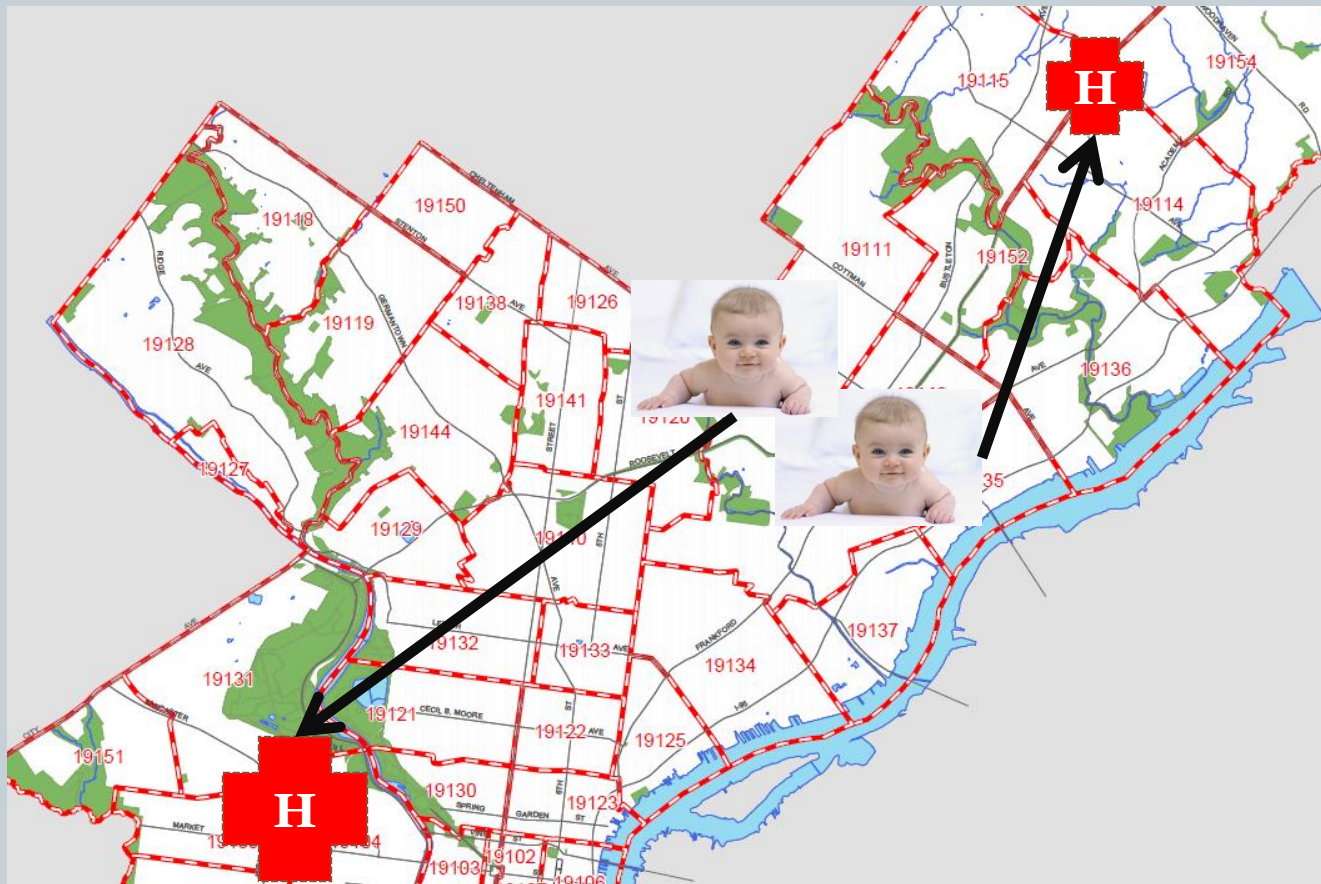
revised design

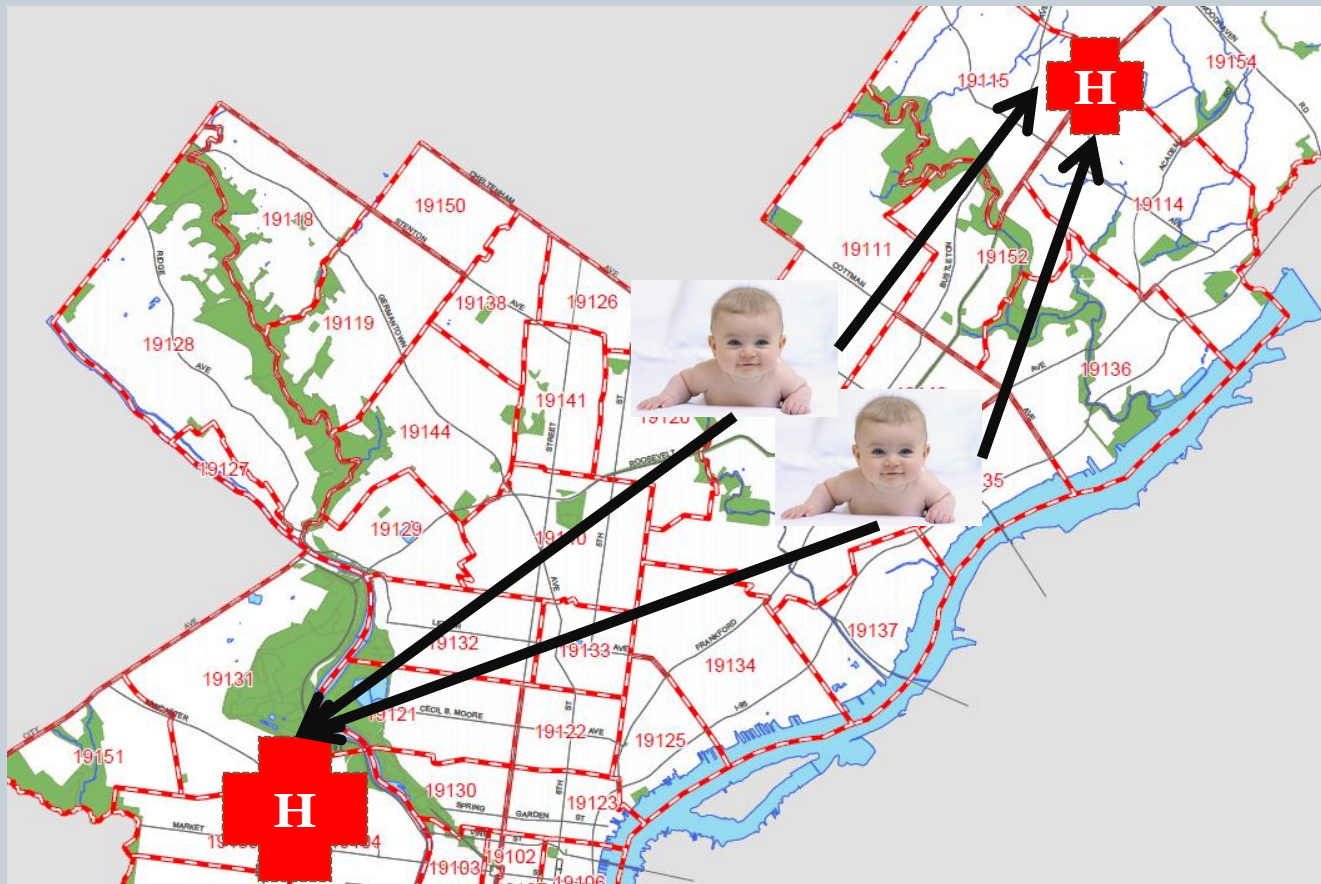


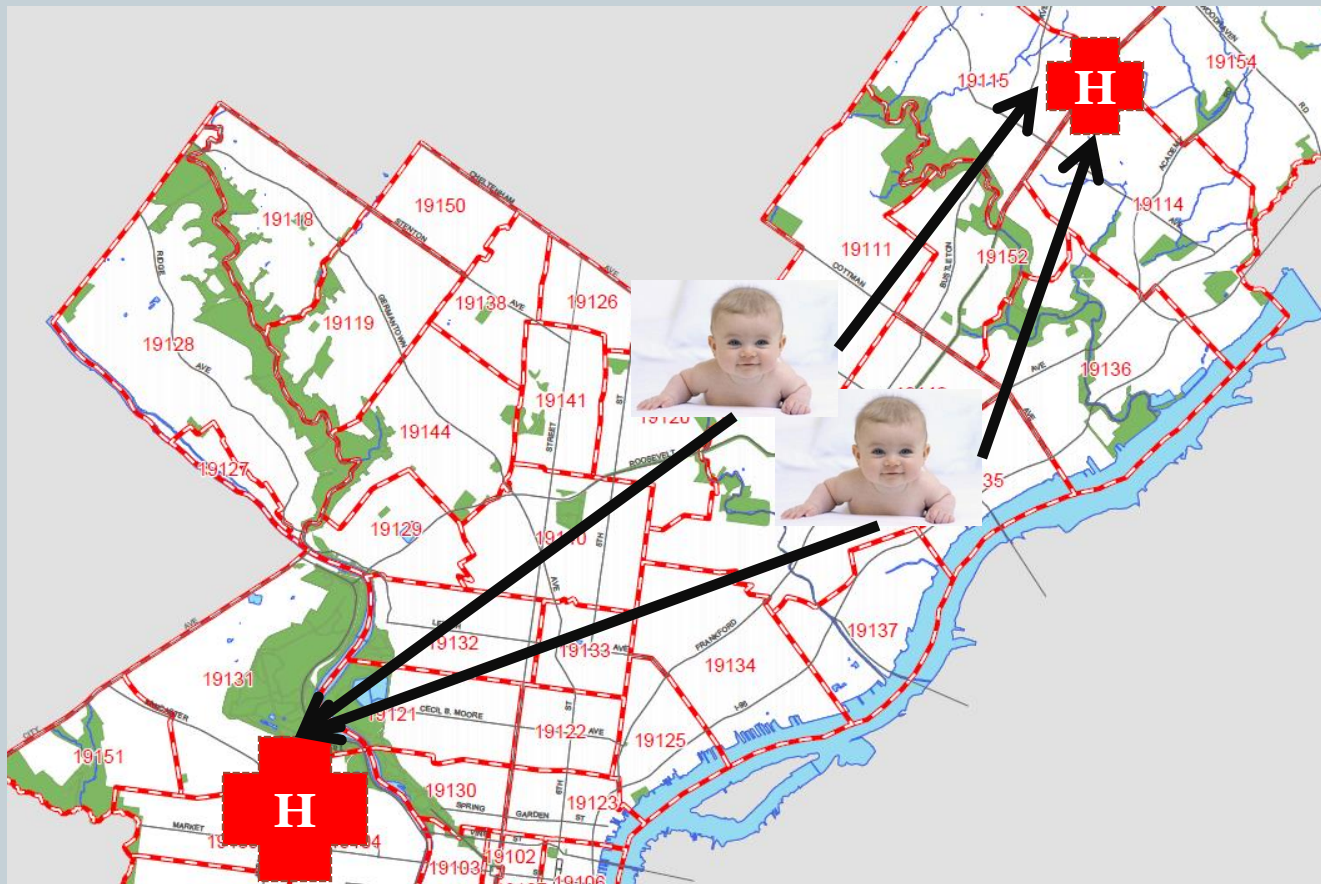
NEAR-FAR MATCHING



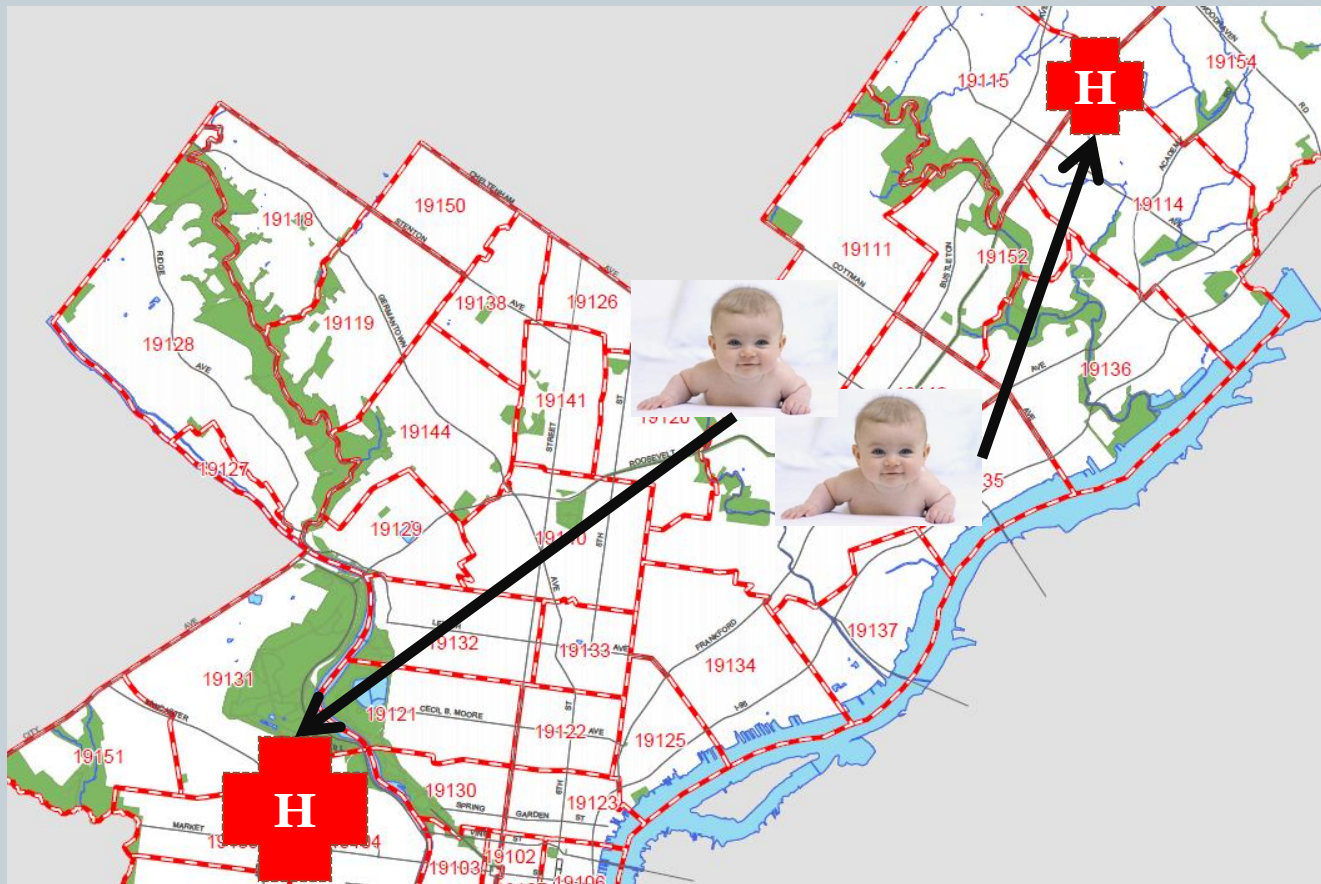




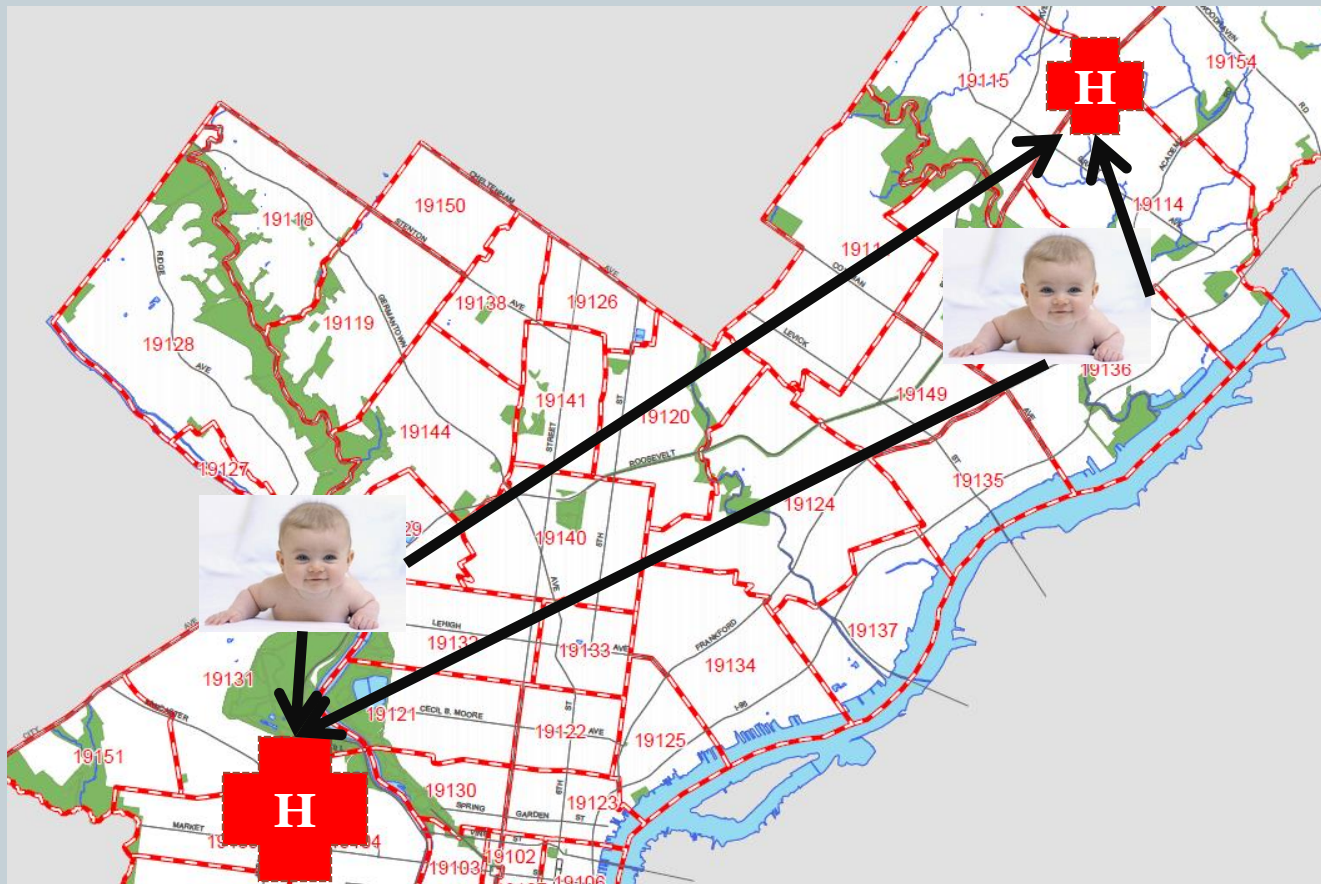




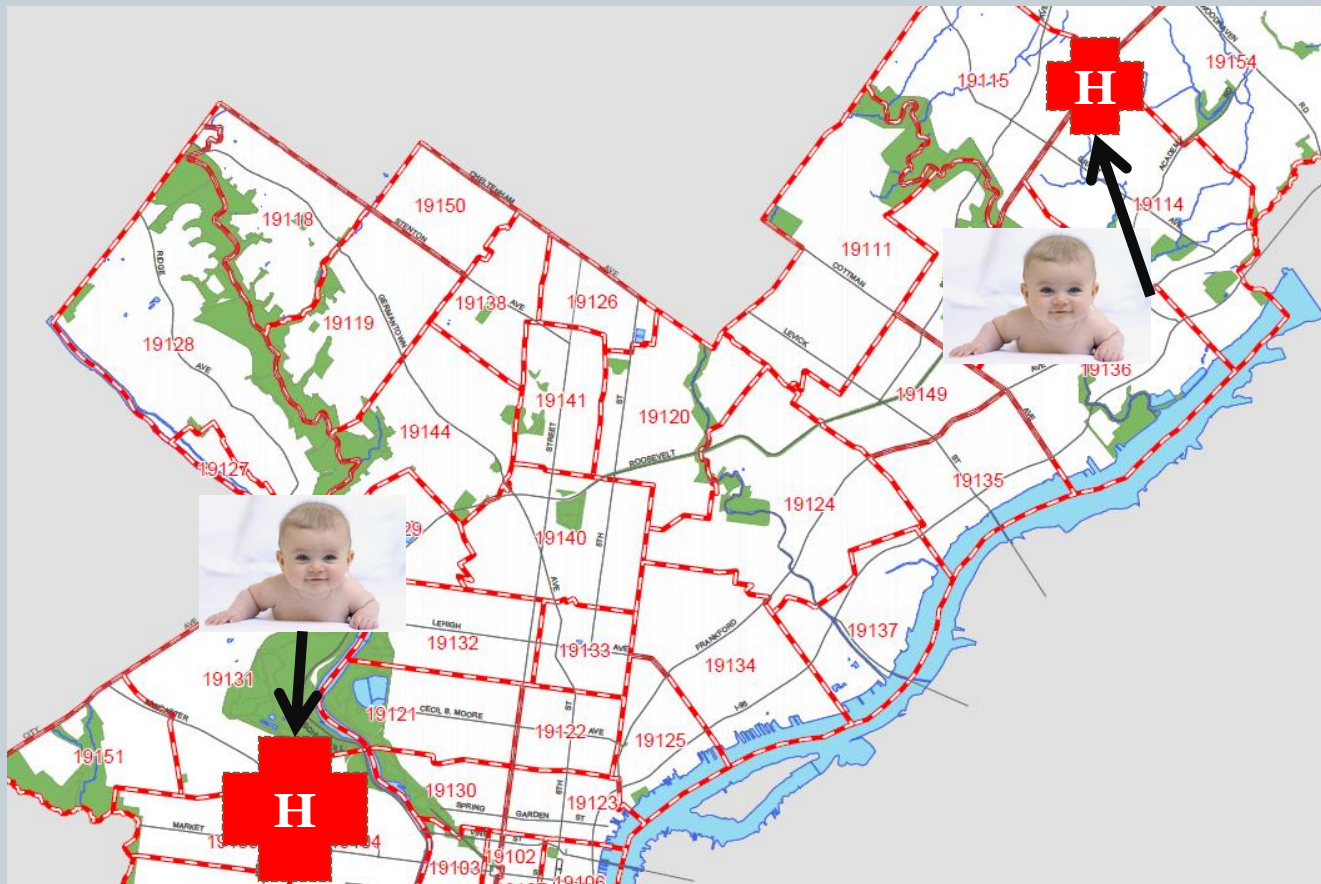
Sorting is potentially biased!



Sorting is potentially biased!



Sorting largely due to the randomness!



Sorting largely due to the randomness!

linking to inference



the fundamental form



- For RCTs, the fundamental form of inference is based on permutation tests (a.k.a. randomization tests)
- For survey, the fundamental form of inference is bootstrap (debatable)
- Everything else are necessary concessions to the particularities of a given problem
- Connect the structure of the data to your form of inference

takeaways



takeaways



- **Design comes in two flavors**
(actually, three... but the third one is not very healthy)
- **In prospective studies**
 - design is an obvious consideration
 - and one that **MUST** be passed through in order to obtain data
- **In retrospective studies,**
 - design is a less obvious consideration
 - but one that **MUST** be passed through... unfortunately without much attention paid