Randomization Inference

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Outline

- 1 Experiments
 - About experiments
 - The framework

- 2 Testing for no effect
 - About these tests
 - Test statistics

These notes are based upon Rosenbaum (2002).

- Experiments do not require units to be homogeneous
- 2 Experiments do not require units to be a random sample from a population of units
- Treatment only needs to be allocated randomly among experimental units
- Probability ("chance") enters the analysis only through random treatment assignment, which the researcher controls

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Neymann believed that units can be chosen randomly from the population to generate external validity.

Let there be N units divided into S strata prior to treatment, giving n_s units in strata s. Define the following

$$Z_{si} = \mathbb{I}\{\text{unit } i \text{ in strata } s \text{ received treatment}\}$$
 $Z_s = (Z_{s1}, \dots Z_{sn_s})'$
 $Z = (Z_{1}, \dots, Z_{S})'$
 $m_s = \sum_{i=1}^{n_s} Z_{si}$

Note that Z is a random variable with a distribution determined by the researcher. The only limitation on this distribution is that $0 < \Pr(Z_{si} = 1) < 1 \,\forall s, i$. Let Ω be the set of all possible realizations of Z.

Examples:

■ Fixed margins

$$m_s = k_s$$

$$K \equiv \#\{\Omega\} = \prod_{s=1}^{S} \binom{n_s}{k_s}$$

■ Binomial randomization

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This sort of test can be performed with very limited assumptions and may provide a good starting point before generating point estimates.

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Hence, when doing inference, we will take r as given (i.e., , condition on it) and compare it to other possible realizations of treatment.

- **1** The null hypothesis is assumed, thereby fixing the value of r.
- 2 Treatment assignment z is selected at random from Ω .
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We have the following procedure for test statistic t(z, r) for fixed set of responses and treatment assignment z:

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This yields a significance level of

$$\Pr(t(Z,r)) = \sum_{z \in \Omega} \mathbb{I}\{t(z,r) \ge T\} \dot{\Pr}(Z=z)$$

Test statistics

In The Lady Tasting Tea, the test statistic that we use is the number of cups correctly guessed:

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When we are calculating the significance of this result against the null, we ask, "if treatment were assigned differently, but his responses do not change, what's the probability that his slate of guesses would have performed at least as well as it did in this case?"

For example, "if we had ordered the cups differently, what's the probability that his guesses would have gotten at least as many cups correct as he did here?"