1. Hypotheses & Metrics - What to Test: pages, funnel, business model, backend functionality, algorithm, new product/feature

- **Hypotheses**: H_0 = control & treatment equal, H_a = different
- Metrics: Primary = biz objective, e.g. order conversion, Guardrail = critical biz metrics, e.g. page load time, Secondary = why primary changed
- Tests: 1. robust to extraneous vars (run A/A test) 2. proportions → use Z-test, e.g. CTR 3. avg/user → use T-test, e.g. avg rev/usr, avg posts/usr
- 4. multiple indep vars → use multivariate test (allows testing interaction between indep vars, but requires larger n)
- 2. Design A/B Test Population: unit (usr id, events, cookies), segment to sample from to reduce dev hrs (geography, device type, browser)
- Cohort: if funnel exclude inprogress usrs from test, if test insignificant test by segment (e.g. behavior {visited 10+ times}, outcome {bought})
- Sample Size: determines test length, set ahead, don't stop early, π analysis (α =.05, 1- β =.8, min detectable effect (MDE) that justifies feat cost)
- Lehr's Formula: approx samp size per group for 2-samp T-test /w equal σ^2 , $n \approx \frac{16\sigma^2}{\delta^2}$ where δ is MDE = μ_0 μ_1 , approx σ^2 by s^2 using past data
- Sampling Strategy: probability sampling (random, stratified, clustered), when to run (season, weekday/weekend)
- 3. Analyze Results Change Aversion: smaller initial effect Novelty: larger initial effect Sol: compare new usrs to veteran usrs in treatment group
- Network Eff: social networks & 2-side markets, treatment interferes /w control violates independ, sol: split groups by location, time, netwrk clusts
- Simpson's: trend when groups of data, no trend when groups aggregated → uncontrolled confounder, sol: segment data to see if trend persists
- Multiple Testing: rerunning test compounds FPR, sol: use ANOVA for multi-group tests, use Bonferroni Correction (divide α by num of tests)
- $\underline{ t 4. \ Make \ Decision}$ Independence: Check usrs were randomized, compare distributions across groups (demographics, device, browser), use χ^2 -test
- Significance Test: check 1. if statistically significant (primary & guardrail), and 2. if practically significant (a. lift = \frac{\text{treatment rate} \control \text{rate}}{\text{control rate}}}
- b. check effect size > MDE, is full confidence interval for p_1-p_2 > MDE, if only partially rerun with larger n to increase π)
- Conflicting Results: positive primary & negative guardrail → quantify both effects, determine net effect
- Launch: ramp up feature release to ensure change aversion or novelty effect aren't permanent