



# Clinical Trial Design

## Statistical Approaches and Considerations

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# Disclosures

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# Outline

- Numerous statistical and design approaches exist to increase trial efficiency
- Many have been tried in cardiology, oncology and other areas
- We review a selection here including
  - Event driven trials
  - Composite outcome
  - Adaptive designs
  - Enrichment designs
  - Borrowing of controls
  - Opportunities in era of big data



# Event driven trials

- In time-to-event outcome studies power depends on the number of events observed
- Efficiency is increased by stopping when minimum necessary number of events is reached
- All follow-up is included in the analysis
- Blinded interim monitoring of event count allows increasing or decreasing planned study duration or increase in enrolled sample



# Composite Outcomes

- Number of primary outcome events can be increased by combining different outcome types
- For example, major adverse cardiovascular events (MACE) consist of cardiovascular death, myocardial infarction and stroke
- Time to first event or hierarchical composite used
- They should be of similar severity
- Analysis of individual components always conducted as sensitivity





# Example: TAVI in aortic stenosis

- The Placement of Aortic Trans-catheter Valves (PARTNER) trial used Finkelstein-Schoenfeld hierarchical composite
- All-cause mortality first in hierarchy, hospitalization for heart failure second
- Multiple pairwise comparisons performed for all patient pairs, first with respect to time to death and then with respect to time to repeat hospitalization
- Method more powerful if some outcomes continuous



# Adaptive Designs

- Key study features (sample size, study duration, number of treatment arms) can be adapted based on information obtained at interim
- Blinded interim looks focus on combined parameters (event rate, overall mean, variance) and usually do not trigger alpha penalty
- Unblinded interim looks incorporate effect size observed at interim and usually trigger alpha penalty
- Decision often based on conditional power



# Example: drop-the-losers design

- In stage A,  $k$  experimental and one control treatment administered
- Data unblinded and analyzed and only the best treatment or any treatment exceeding pre-specified threshold proceed to stage B along with control
- Final summary statistic based on sample size-weighted combination of effect versus placebo from stages A and B



# Biomarkers

- FDA's Qualification process for drug development tools:
  - A *prognostic* biomarker is a baseline patient or disease characteristic that categorizes patients by degree of risk for disease occurrence or progression. A prognostic biomarker informs about the natural history of the disorder in that particular patient in the absence of a therapeutic intervention
  - A *predictive* biomarker is a baseline characteristic that categorizes patients by their likelihood for response to a particular treatment. A predictive biomarker is used to identify whether a given patient is likely to respond to a treatment intervention in a particular way (favorable or not)



# Example: Predictive biomarkers

Effect of l-Methylfolate 15 mg/d vs Placebo on Pooled Mean Change From Baseline for HDRS-28 Stratified by Baseline Level of Plasma Marker

| Marker        | N  | Pooled Mean Change vs. Placebo | 95% CI       | P-value |
|---------------|----|--------------------------------|--------------|---------|
| SAM/SAH ratio |    |                                |              |         |
| ≥ 2.71 mg/L   | 36 | 0.07                           | -3.33, 3.48  | 0.966   |
| < 2.71 mg/L   | 37 | -4.57                          | -7.73, -1.41 | 0.005   |
| 4-HNE         |    |                                |              |         |
| ≥ 3.28 µg/mL  | 37 | -4.55                          | -7.61, -1.50 | 0.003   |
| < 3.28 µg/mL  | 36 | -0.11                          | -3.67, 3.46  | 0.953   |



# Adaptive Enrichment Designs

- $G_0$  – all eligible subjects
- $G_1$  – subset of all eligible subjects who possess non-wild-type alleles on either genotype A and B
- $G_2$  – subset of eligible subjects who possess non-wild-type alleles on both genotype A and B
- Conditional power calculated at interim based on effect size in each set
- Opportunity to enrich by recruiting only to subset  $G_1$  or  $G_2$



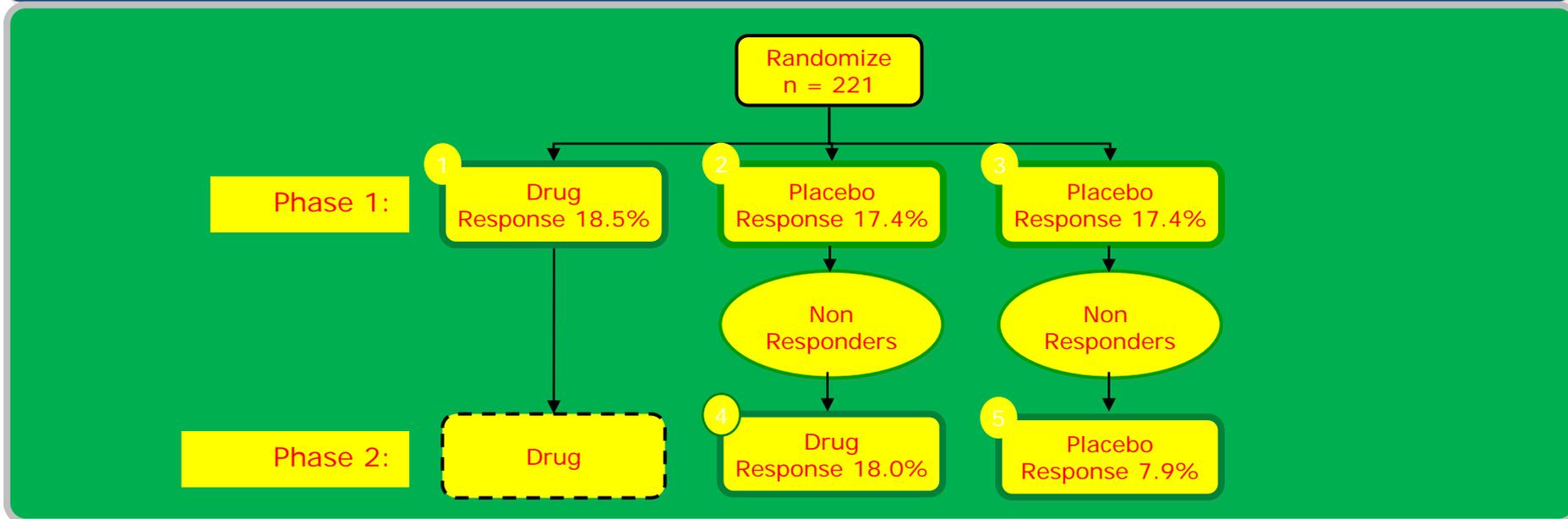
# Sequential Parallel Comparison Design

- Developed to reduce impact of placebo response
- In stage A, individuals randomized to treatment versus placebo (usually more to placebo)
- Stage A placebo non-responders re-randomized to treatment versus placebo
- Final summary statistic based on weighted combination of effects from stages A and B
- Particular choice of weights enables interpretation as effect after accounting for placebo response

# ADAPT-A: Aripiprazole Augmentation of SSRIs

Source: Fava et al, Psychotherapy and Psychosomatics 2012

|              | Response rate |         |            |
|--------------|---------------|---------|------------|
|              | Drug          | Placebo | Difference |
| SPCD Phase 1 | 18.5%         | 17.4%   | 1.1%       |
| SPCD Phase 2 | 18.0%         | 7.9%    | 10.1%      |





# Borrowing Controls

- All or some controls are borrowed from “historical” data
- Numerous options:
  - Pooling: adds historical controls to randomized controls
  - Performance criterion: uses historical data to define performance criterion for current, treated-only trial to beat
  - Test then pool: test if controls sufficiently similar for pooling
  - Power priors: historical control discounted when added to randomized controls
  - Hierarchical modeling: variation between current vs. historical data is modeled in Bayesian fashion



# Opportunities in era of big data

- We might be able to run very large simple trials for fraction of cost
  - EHR-enabled trials
    - pcornt ADAPTABLE – trial of aspirin doing in secondary prevention with 20,000 patients with EHR as primary source of data capture
  - Registry-based trials
    - SAFE-PCI in women – comparing access site (radial vs. femoral) in 1800 PCI or angiography with possible PCI women
    - Several ongoing Registry-RCTs in Sweden



# ADAPTABLE Study Design

## Patients with known ASCVD + $\geq 1$ “Enrichment Factor”

Identified through EHR screening and electronic patient contact by CDRNs/PPRNs  
(PPRN patients would need to connect through a CDRN to participate)

Patients contacted electronically with trial information and e-consent via web portal  
Treatment assignment will be provided directly to patient

ASA 81 mg QD

ASA 325 mg QD

Randomized Electronic Follow-Up: 3 vs 6 months  
Supplemented with EHR/CDM Data Queries

**Duration:** Enrollment over 24 months;  
maximum follow up of 30 months

**Primary Endpoint:** Composite of all-cause mortality, hospitalization for MI, or hospitalization for stroke

**Primary Safety Endpoint:** Hospitalization for major bleeding

### \*Enrichment Factors

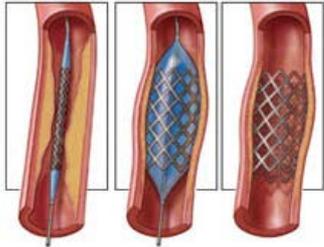
- Age > 65 years
- Creatinine > 1.5 mg/dL
- Diabetes mellitus (type 1 or 2)
- Known 3-vessel CAD
- Current CVD or PAD
- Known EF<50% by echo, cath, nuclear study
- Current smoker



# Methods - SAFE-PCI for Women Workflow



*Randomization*



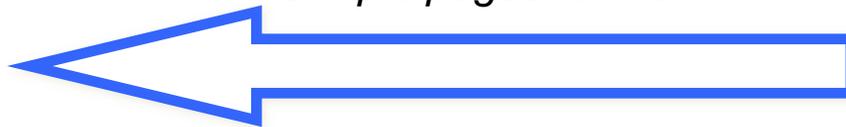
*Demographics  
Medical History  
Procedural data*



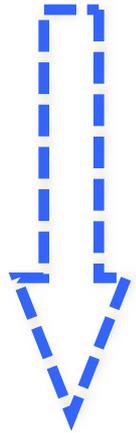
*Autopopulate trial  
database with registry  
"big" data*



*Unique pages for trial*



**Analytic  
Database**





# Conclusions

- Numerous study designs and statistical approaches intend to improve trial efficiency
- Different approaches at different stages of adoption and different potential for application in neuroscience studies
- In general, regulators more open to innovative approaches in smaller studies and/or earlier stages of development (II vs. III)
- Careful consideration and appropriate statistical expertise in the planning stage needed to identify approaches most likely to succeed