

# JENNIFER THOMPSON

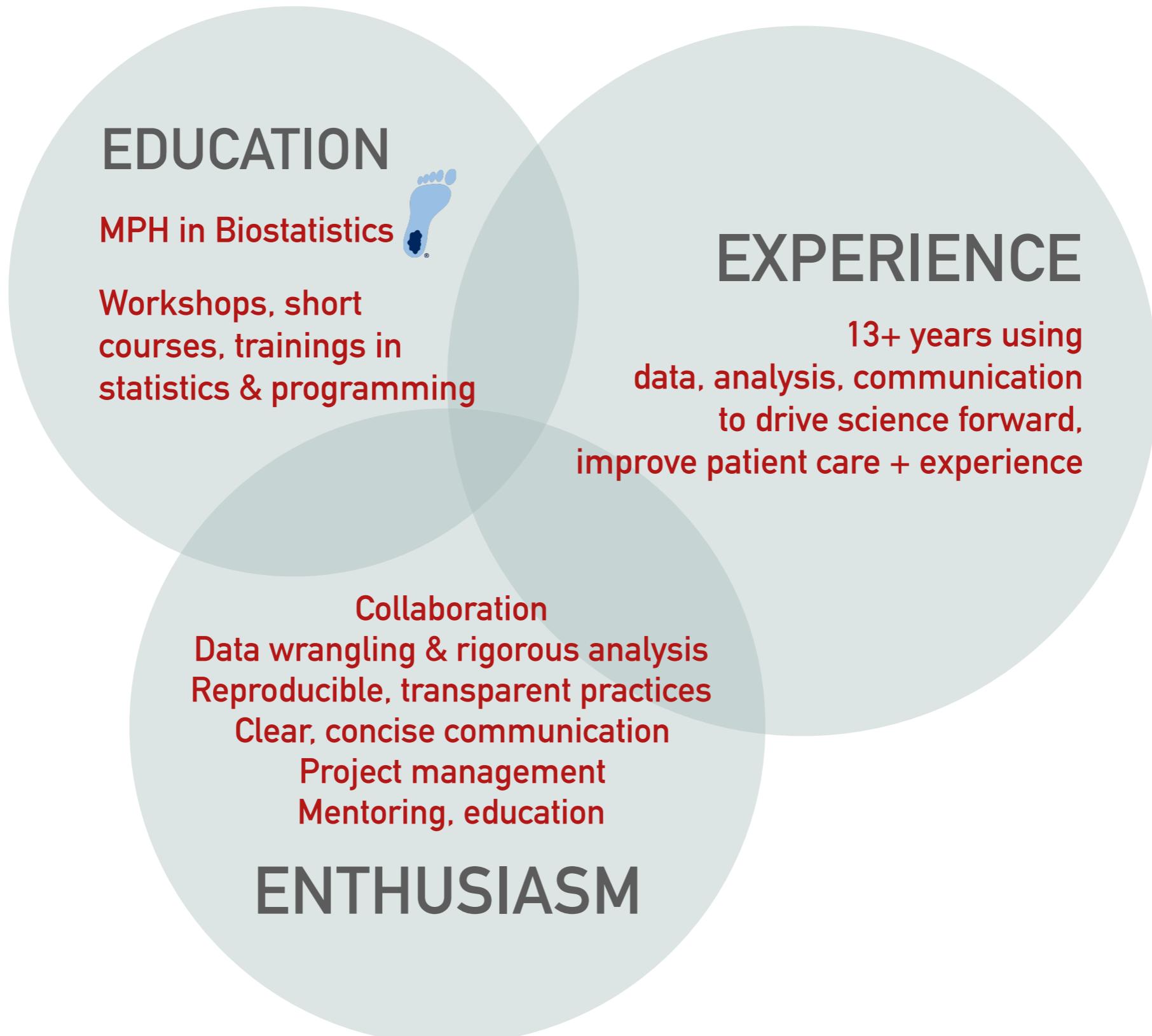
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*American Hospital Association  
October 5, 2018*



# (PROFESSIONAL) ME: A QUICK INTRODUCTION

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# **MY APPROACH TO RESEARCH**

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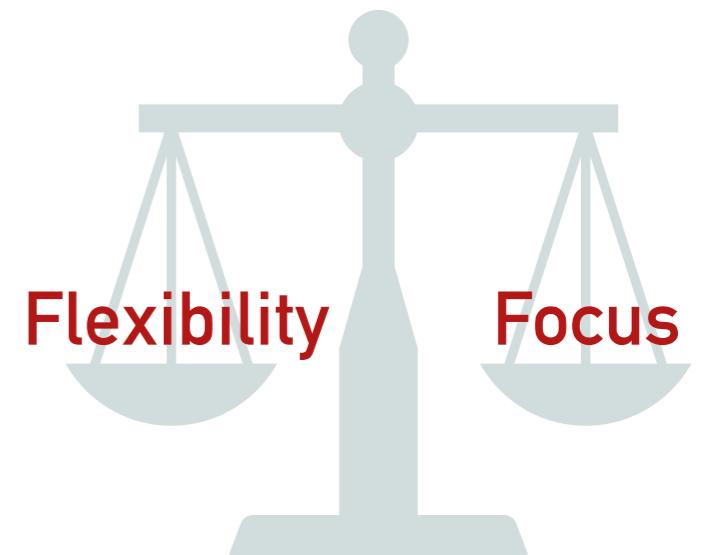
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## 1. Define the relevant question(s)

- Which answers are **actionable** - would change practice, increase understanding?
- Which questions do we have **resources** to reliably answer?

## 2. Form a plan

- Convert questions to **measurable** outcomes
- Develop study **design**, statistical **analysis plan**: as complex as needed, but no more
- Include ongoing maintenance (eg, web app), dissemination of results
- Use plan to manage ongoing project scope



# MY APPROACH TO RESEARCH

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## 3. Assess and connect **resources**

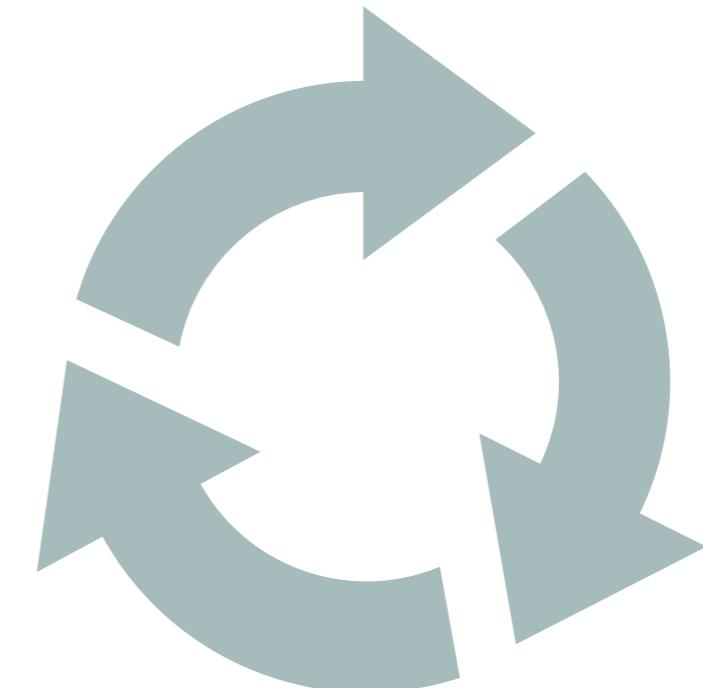
- Data
- Domain experts, other team members
- Analytic/presentation tools

## 4. Wrangle & explore **data**

- Are **assumptions** accurate?
- How should data be **transformed** to answer our questions?

## 5. Execute the plan

- Sounds deceptively simple!
- Communicate roadblocks to team, stakeholders
- Adapt (and document) as needed



# MY APPROACH TO RESEARCH

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## 6. Alongside entire team, interpret & **communicate** impact, results, methods

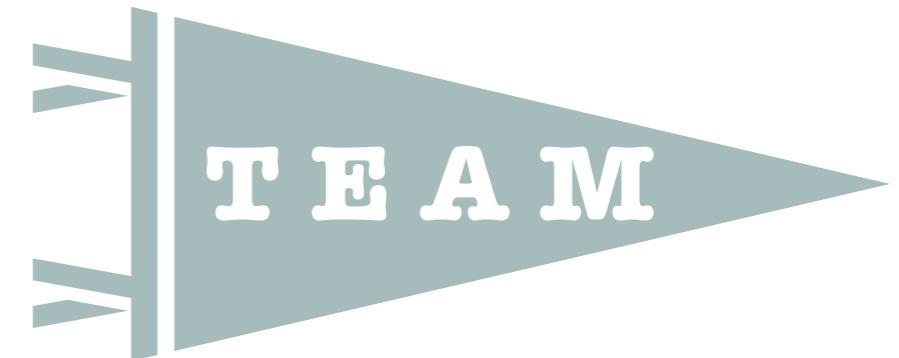
- Need clear message without loss of nuance
- Each level of detail important for different audience
- Emphasize conciseness, visuals in product(s) appropriate for goals, audience  
(manuscript/white paper, web site/app, presentation...)



# THROUGHOUT THE PROCESS

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- Priorities reflect our goals & mission
- Data is complete, accurate
- Research team \*understands\* raw data
- Thorough, appropriate analysis
- Primary message & methods communicated appropriately for specific audience(s), whether by me or colleagues
  - Can my clinical teammates explain our results to other clinicians?
  - Can I explain real-world impact to other statisticians?
- Work done in reproducible, rigorous manner
  - Sensible organization and documentation
  - Literate programming/reproducible reporting
  - Version control
  - Aspirational: Code review



# **SCENARIO 1:**

## **NIH R01 CLINICAL TRIAL**

# SITUATION 1: PROSPECTIVE, NATIONWIDE DATA COLLECTION

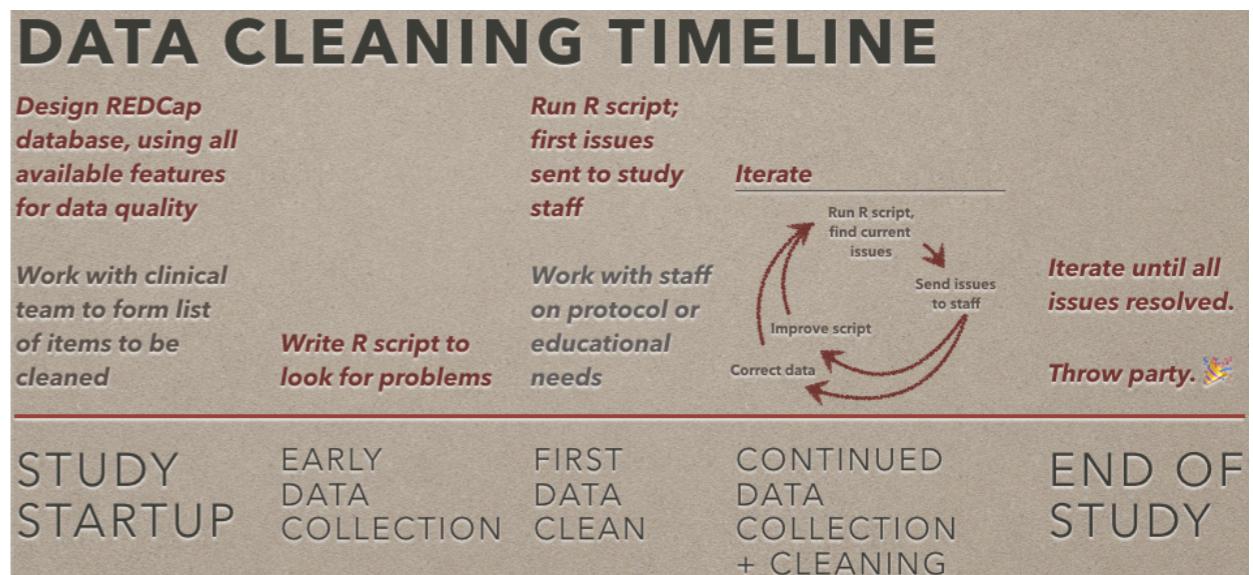
- ~15 sites, 6 years of enrollment -> staff turnover, fuzzy memories
- Complex, longitudinal study design -> detailed data collected by hand, with different data required in different study phases



## SOLUTION

## RESULTS

Have trained others on this process;  
tutorial available on GitHub

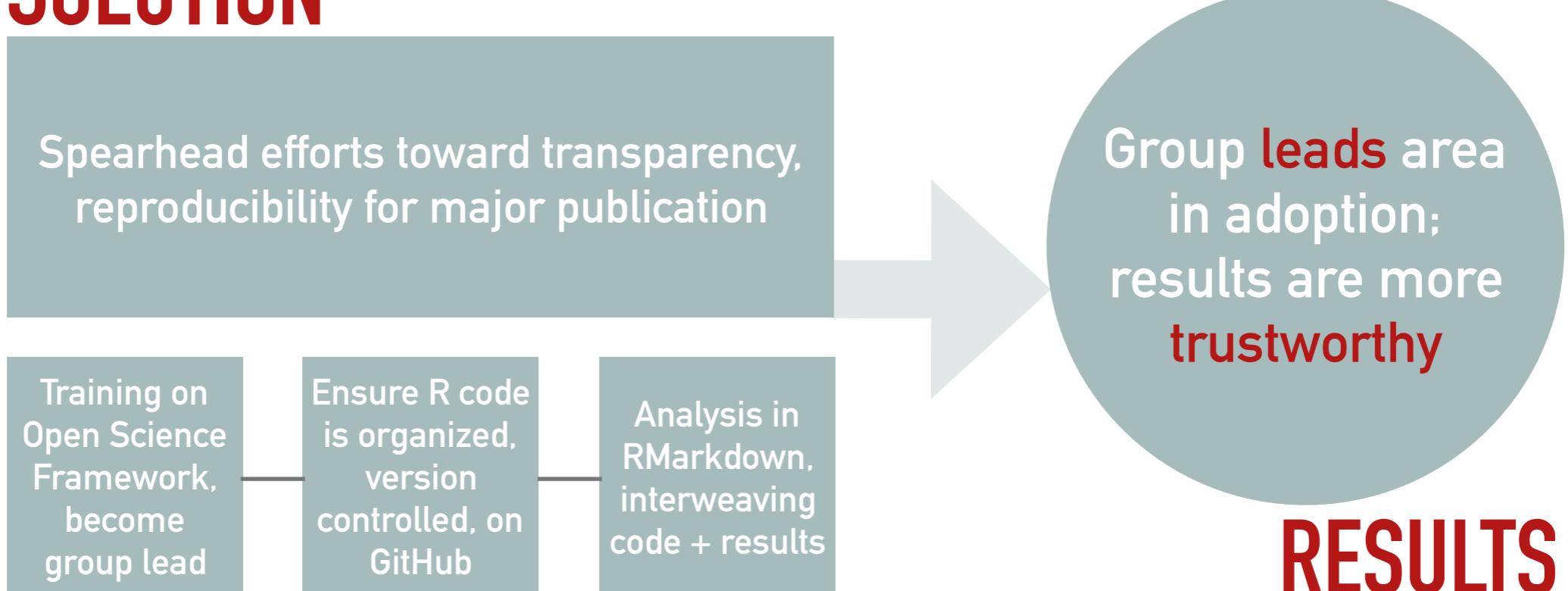


# SITUATION 2: REPRODUCIBLE, OPEN RESEARCH TAKING OFF

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- Reproducibility is of growing importance in scientific community; open science becoming more and more the norm 🤘
- Group doesn't have direct experience with tools and practices

## SOLUTION



## SITUATION 3: BALANCING COMPLEXITY AND CLARITY

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- Primary outcome is a) complex, combining three related metrics, and b) has a notoriously bimodal distribution
- Best choice for modeling has a clinically difficult interpretation, especially combined with outcome's inherent complexity



# SITUATION 4: COMPLEX, LONGITUDINAL STUDY DESIGN

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Study design (& therefore data collection) is unusual & complex:

- Four study phases; patients could be disqualified within 5 days or randomized at varying times after consent
- Outcomes are summaries of raw data



## SOLUTION

Rigorous data wrangling process

Formal tests for raw & transformed data

Detailed file structure + version control

Thorough commenting, documentation



Reliable analysis datasets

Easier to identify, fix bugs

Easier to collaborate

## RESULT

More **reliable**, transportable code + data

# **SCENARIO 2:**

## **NIH COHORT STUDY**

# SITUATION: NEED TO MONITOR ENROLLMENT, STUDY CONDUCT

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- Team, PI need regular updates on enrollment targets, protocol points, follow-up rates
- Need to serve variety of folks with broad range of backgrounds
- Previous: Detailed PDF; worked, but buried important metrics

## SOLUTION

Dashboards for quick info, warning system for pain points

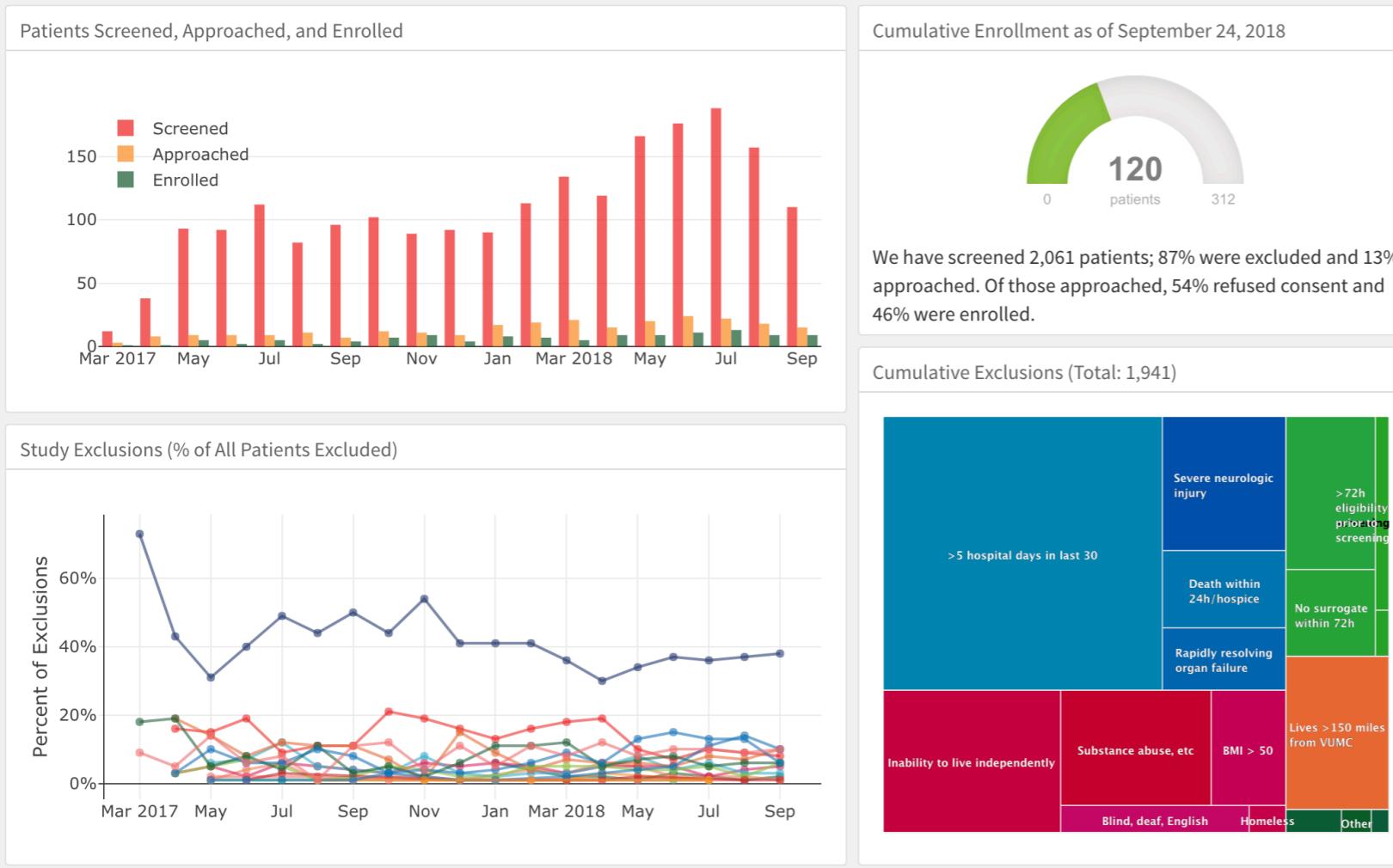


**Quickly** see successes, attention areas

Visual cues **incentivize** targets

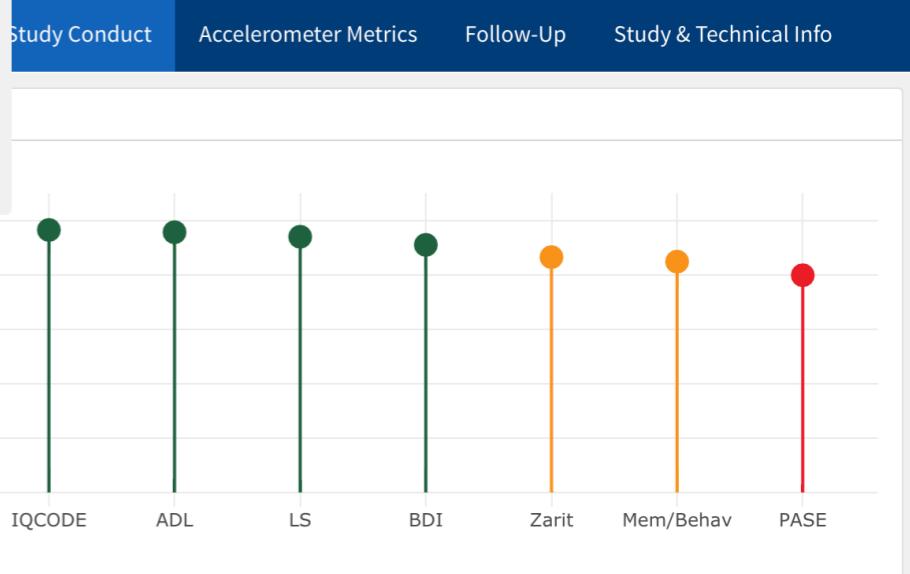
Starts **communication** that leads to solutions

## RESULTS



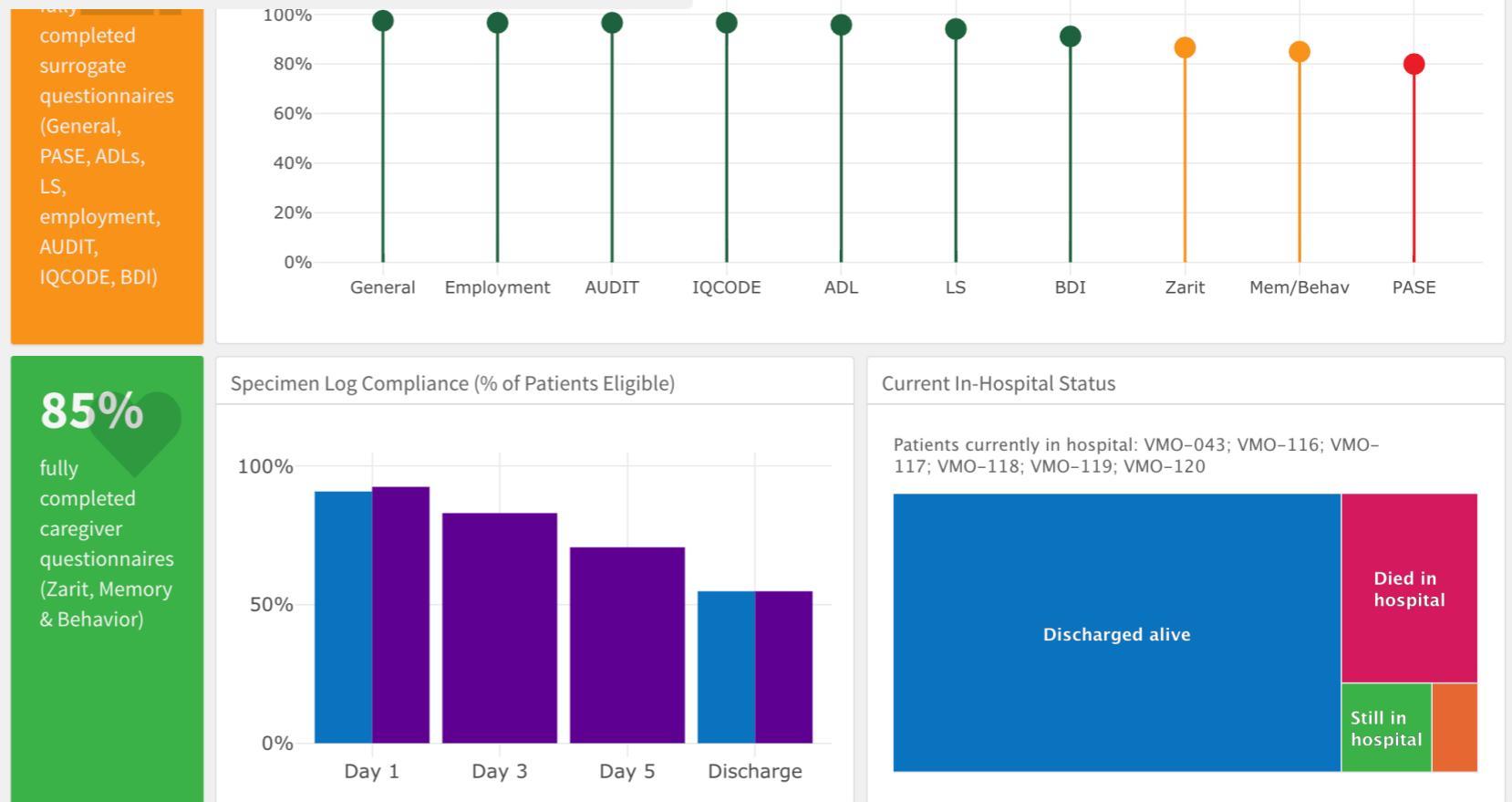
## Enrollment tab

- Quick big picture of exclusions, refusals, enrollment
- More detail available via interactive graphics
- Use study logo, color palette for consistent branding



## In-hospital tab

- Conditional coloring: quick sense of ability to collect surrogate info
- Lollipops show same for individual instruments
- Specimens collected in blue + purple tubes; using those colors gives quick visual cues
- Patient IDs staff is still following daily = sense of current data collection burden



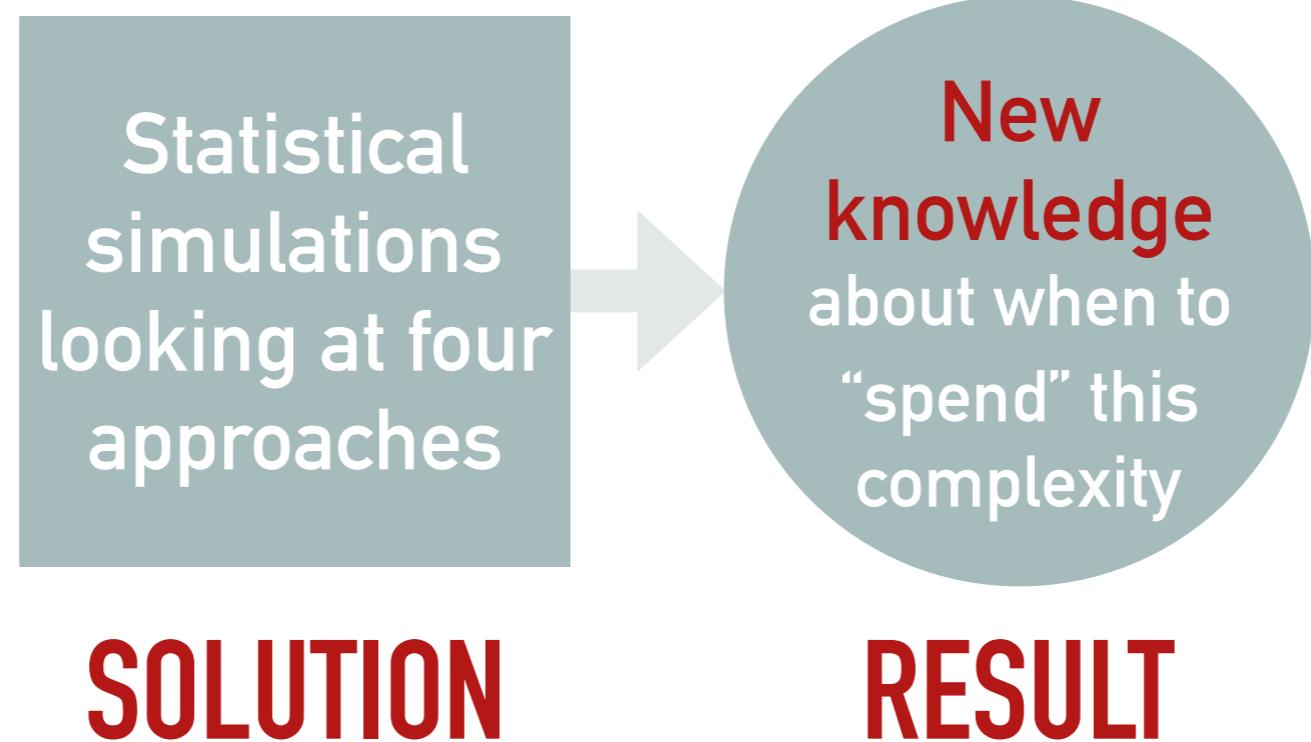
# **SCENARIO 3:**

## **EMR DATA**

# SITUATION: USING EMR VS PROSPECTIVELY COLLECTED DATA

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- Rich resource with inherent issues, specifically plenty of missingness
- Mental status not assessed as often in trauma unit (subject of this project) as in medical, surgical units; primary exposure relies on these assessments
- We don't know which imputation approach is least likely to bias our results; we assume complex = better, but want to make sure it's worth the resources



# **SCENARIO 4:**

# **MANAGING COMPETING**

# **PRIORITIES**

# SITUATION: ONE GROUP, >10 INVESTIGATORS

- As research group has grown, statistical resources have not kept pace
- Currently balancing needs of 10 faculty + multiple fellows
- Need communication, prioritization that works

WHAT BEST  
SERVES OUR  
MISSION?

## SOLUTIONS FOR OUR CIRCUMSTANCES

Weekly investigator meetings  
Slack channel for priority list

“Focus project” vs “ad hoc” weeks

Request form for new projects (aims, \$, IRB)

Draft shell of product before analysis begins

## RESULTS

Better communication  
Pls understand priorities, rationale

Progress toward goals  
Major projects move forward, smaller requests covered

Remove logistical barriers, set expectations before analysis begins

Increased efficiency  
Complete analysis doesn't wait for someone to take ownership

IN PROGRESS: ONBOARDING DOCS FOR NEW INVESTIGATORS

# **SCENARIO 5:**

## **CONTINUING EDUCATION**

# SITUATION: GRAD SCHOOL DOESN'T TEACH US EVERYTHING

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- Fields of data science, research, biostatistics constantly evolving; no one can know everything
- Difficult to balance “getting things done” with learning new skills; not all local colleagues have the same level of interest

## SOLUTIONS

Find/create communities that encourage growth

Create or take advantage of opportunities for training, connections

Freely share code, knowledge (GitHub, tutorials...)

## RESULTS

Awareness of new ideas, research, tools, collaborators

Larger toolbox for solving problems

Ability to communicate with folks from variety of backgrounds

Publicity for methodological, clinical work

# THANK YOU!

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*Talks, writing, CV, links:  
[jenthompson.me](http://jenthompson.me)*

