

# Changing the Default

## Moving to an Opt-Out Approach for Treating Tobacco Use Disorders

Kimber Richter

Departments of Population Health, Biostatistics (KUMC)

Children's Mercy Kansas City

Department of Health Services, Policy & Practice (Brown)



# Happened to you/a family member?

Health care provider, after intake, tells you

- Your blood pressure is well above normal...
  - She asks you if you are ready to change it within the next 30 days
- You have shortness of breath due to asthma...
  - He asks you if you are willing to address it at this time

# Guidelines

- 5 “A”s of tobacco dependence treatment
  - US – 3<sup>rd</sup> “A” – “Is the tobacco user willing to try to make a quit attempt at this time?”

# Strength of Evidence

<b>ASSESS = 1C</b>	1 = Strong Recommendation: Benefits appear to outweigh risks and burdens	<i>C = Low Quality Evidence</i> ---from observational studies, clinical experience, or flawed trials
<b>ASSIST = 1A</b> <u>Medications</u> <u>Supportive</u> <u>Counseling</u>	1 = Strong Recommendation: Benefits appear to outweigh risks and burdens	<i>A = High Quality Evidence</i> ---from multiple well designed trials

# “Unwilling” Smokers Benefit from Cessation-Oriented Care

- Smokers *not* ready to quit actually quit at the same rates as those who *are* ready to quit (Ellerbeck, 2009)
- Inter99 Study - smokers not planning on quitting will accept treatment and quit (Pisinger, 2005)
  - Only 11% planning to quit in next month
  - 27% enrolled in groups
  - 35% of enrollees quit
    - Only half of those who ultimately quit, initially said they were planning to quit
- Harm??? No data (smokers, providers, systems)

# Screen vs Proactively Treat?

## Screen!

- Guideline recommended
- ???



## Proactively Treat!

- Some will quit who say they're not ready
- Smokers, even those not planning to quit, more satisfied with providers who offer tobacco treatment
- If we don't we'll miss treating 80% of smokers

# Defaults Affect Behavior

- For any choice point, there's a default – what you get if you do nothing
- Making an option the default increases the chances that it will occur
- Organ donation—
  - Germany – no one is a donor, have to “opt in” – 12%
  - Austria – everyone is a donor, have to “opt out” – 99%
- HIV screening – screening rates increased when it was changed to opt-out



# Opt-In vs Opt-Out Tobacco Treatment in Hospital

*Changing the Default*

## Opt In

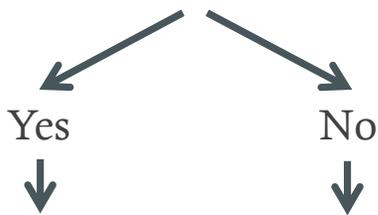
**Offer:** in-patient medication



**Offer:** Brief Advice to quit



**Ask:** Willing to try to quit?



**Offer:**

1. Treatment plan
2. Post discharge Meds
3. Post discharge Support

Motivational  
Intervention

## Opt Out

**Provide:** in-patient medication



**Provide:** Brief Advice to quit



**Provide:**

1. Treatment plan
2. Post discharge Medication
3. Post discharge Support

# Continuum of Intervention Directiveness/Intensity

OPT IN ----- OPT OUT

- Ask all tobacco users:  
Willing to try to quit?
- Yes: Offer meds/support
  - No: motivate

Provide meds/support to  
all unless they refuse

# Continuum of Intervention Directiveness/Intensity

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- Ask all tobacco users:  
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Offer meds/support to all

Provide meds/support to  
all unless they refuse

# Continuum of Intervention Directiveness/Intensity

OPT IN ----- OPT OUT

Ask all tobacco users:  
Willing to try to quit?  
• Yes: Offer meds/support  
• No: motivate

~~Offer me support to all~~

“We’ll call you to provide in-home follow-up post-discharge to check in on how you are doing”

# Choice Architecture

*Nudge, Thaler & Sunstein*

Components	OPT OUT	OPT IN	
INTRODUCTION:	<p><i>"Because quitting is the best thing you can do for your health, KUMed provides tobacco treatment for everyone who smokes."</i></p>	<p><i>"Quitting is the best thing you can do for your health. Are you planning on staying quit once you leave the hospital?"</i></p> <p style="text-align: center;">  </p>	
INPATIENT COUNSELING:	<p><i>"Let's create a brief treatment plan that outlines your thoughts, feelings, and plans to treat your tobacco use"</i></p>	<p><i>"If you'd like, we can create a brief treatment plan that outlines your thoughts, feelings, and plans about your tobacco use."</i></p>	<p>Brief motivation: <i>"I'd like to talk with you about the risks of continuing to smoke and the roadblocks in trying to quit."</i></p>
INPATIENT MEDICATION:	<p><i>"I'm going to work with your medical team to get you inpatient medication to prevent withdrawal"</i></p>	<p><i>"Would you like inpatient nicotine replacement to prevent withdrawal?"</i></p>	<p><i>"Would you like inpatient nicotine replacement to prevent withdrawal?"</i></p>
OUTPT COUNSELING:	<p><i>"We provide in-home counseling post-discharge to help you with your plan."</i></p>	<p><i>"Would you like in-home counseling post-discharge to help you with your plan?."</i></p>	
OUTPT PRESCRIPTION/ STARTER PACK:	<p><i>"We send everyone who is medically eligible home with a prescription and 2 weeks of free NRT."</i></p>	<p><i>"If you are medically eligible, would you like a prescription and 2 weeks of free NRT?"</i></p>	

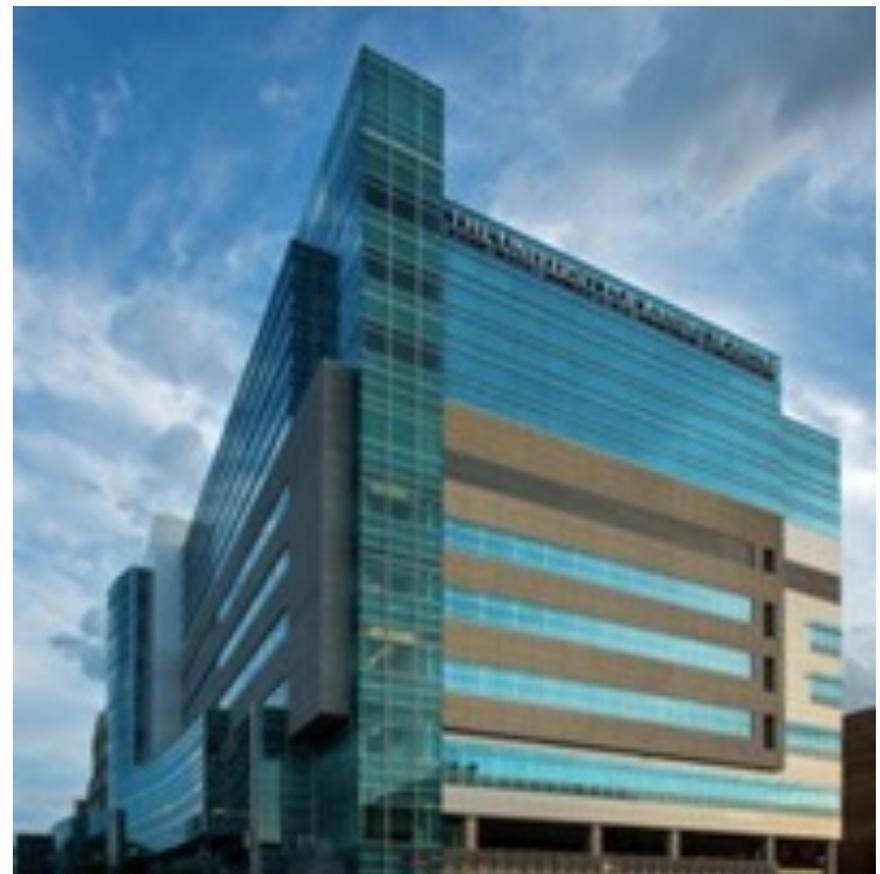
# Innovations Demanded by Research Question

- Delayed consent – population based study, want all smokers, not just willing study participants
- Adaptive trial – interim analyses every 13 weeks, reweight randomization to favor stronger arm
  - More ethical – get more power for 3+ arm studies
- Bayesian design

# THE UNIVERSITY OF KANSAS HOSPITAL

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- Randomized clinical trial
- 1,000 smokers
- Integrated into hospital service



# Changing The Default For Tobacco Treatment

**Aim 1: To determine the population impact of changing the default for tobacco cessation treatment**

- Hypothesis: More enrolled in **OPT OUT** will utilize counseling, medications, and be abstinent from smoking at 1 month post randomization compared to **OPT IN**

**Other Aims: To identify 6-month abstinence, treatment reach, patient response, costs**

# Adaptive Trials

- Definition (FDA): “...an adaptive design is defined as a clinical trial design that allows for prospectively planned modifications to one or more aspects of the design based on accumulating data from subjects in the trial.”
  - Planned
  - Clearly-defined
  - Valid (e.g. 5% Type I error rate, etc.)

<https://www.fda.gov/media/78495/download>

# CTD Design

- Initially randomize participants equally to the two arms until 400 participants randomized.
- After that, do an interim analysis that changes the allocation to weigh more towards the better performing arm (using 1-month endpoint).
- Interim analyses occur every 13 weeks until success or 1000 max participants enrolled.
  - Success occurs if the posterior probability of one arm being better than the other is bigger than .9925 for both 1-month & 6-month endpoints.
  - The type I error of this design is 5% and the power is more than 80%.
    - The details of these calculations uses simulation and we will not go into these details today

# CTD Main Outcomes

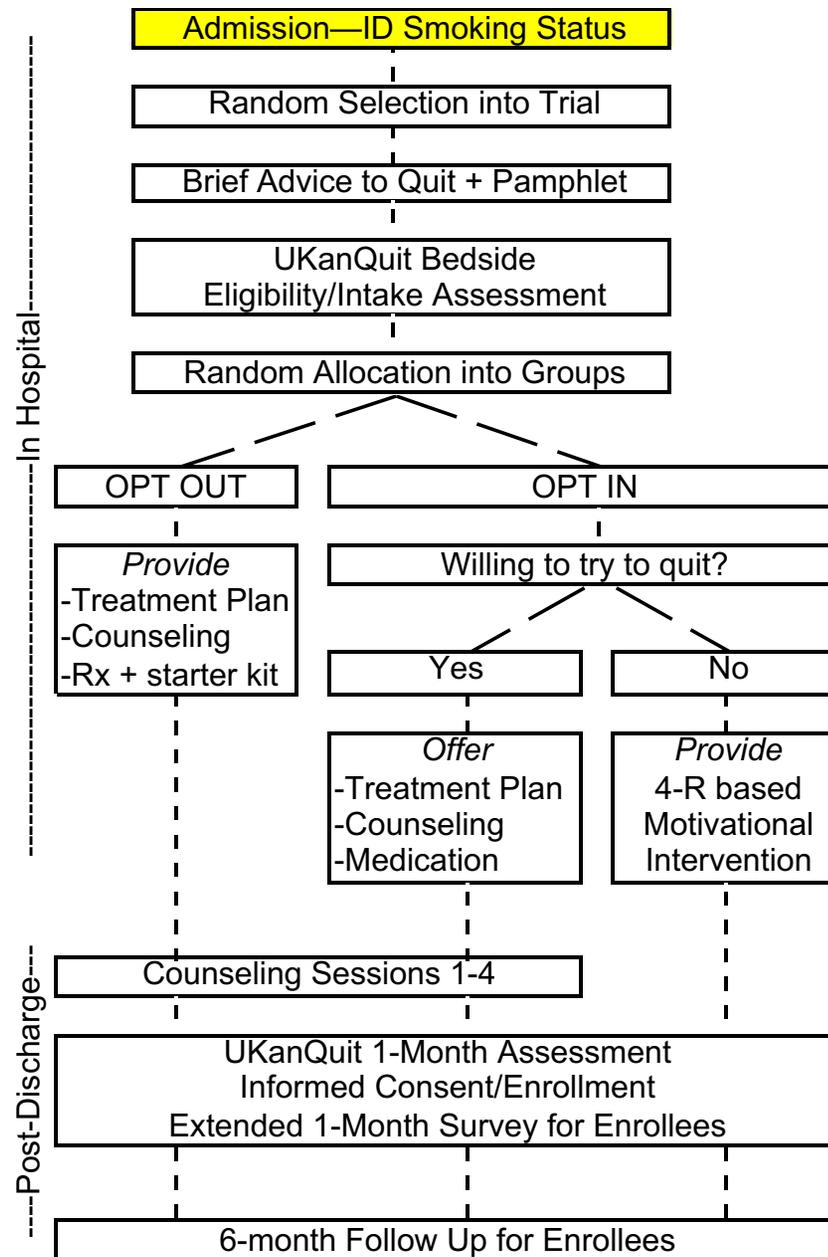
- Primary endpoint: rate of 7-day biochemically verified cigarette abstinence at 1 month after randomization in OPT IN arm versus OPT OUT arm.
- Co-Endpoint: biochemically verified abstinence at 6 months

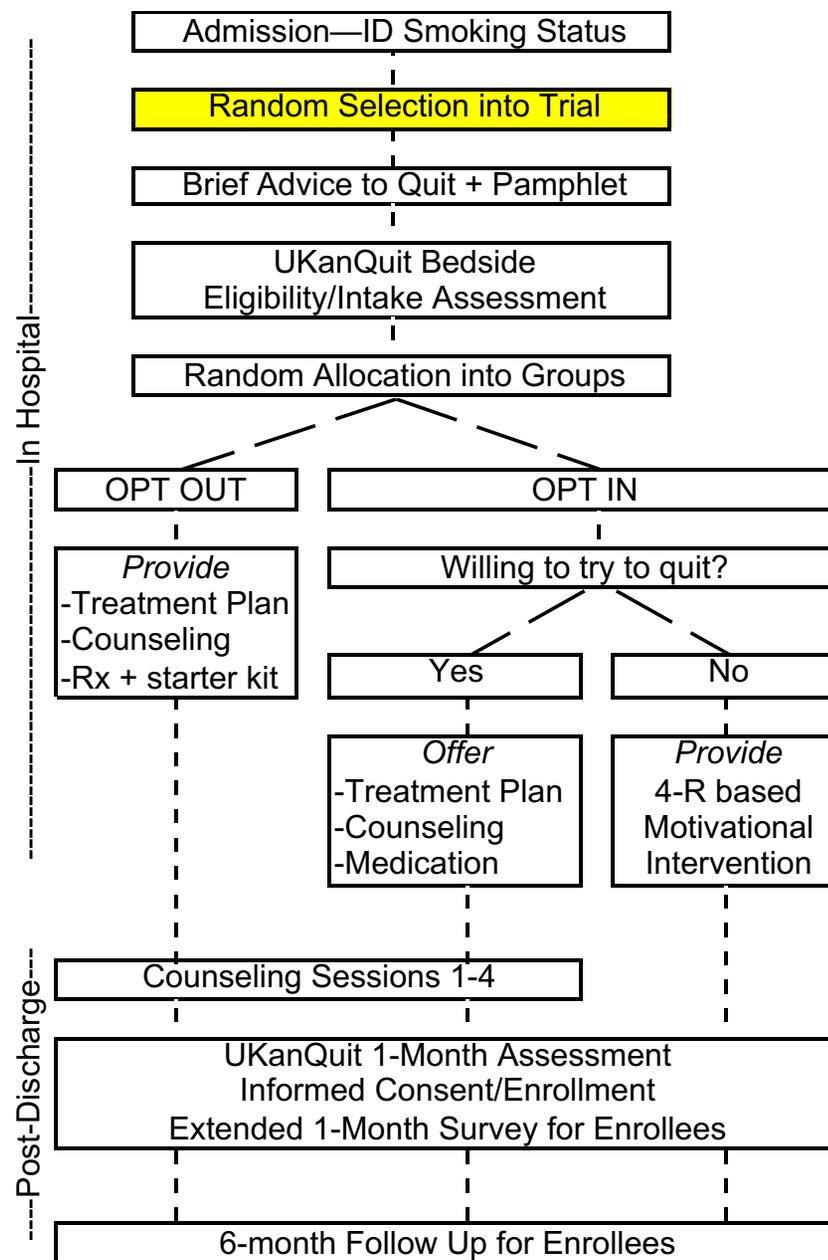
# What Will be Missing from the Results of this Trial?

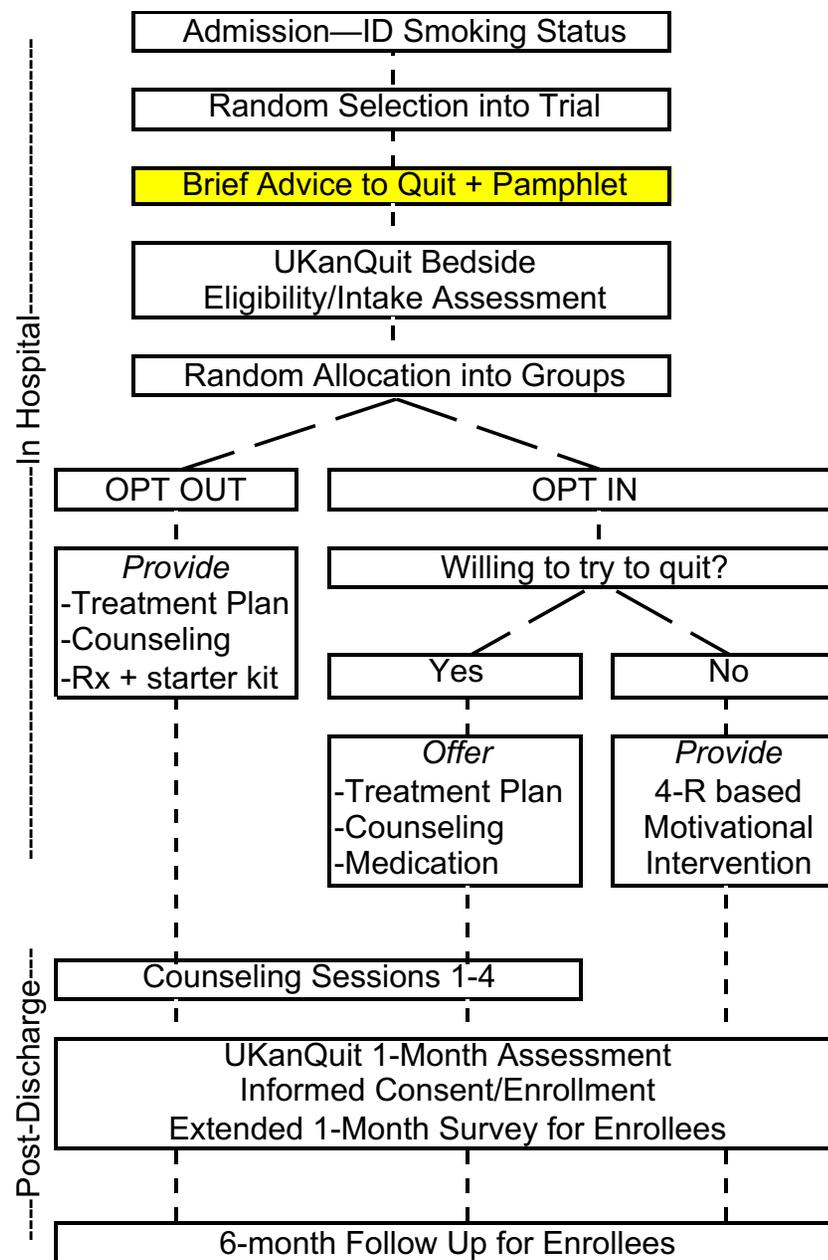
- p-value!!!
  - For example, some trial results can say the comparison between drugs A and B is statistically significant (e.g.  $p=.0137$ ).
  - What **does** this mean?
  - It means that “the probability of being more extreme than the test statistic summarizing the differences between drugs A and B, under the null hypothesis of drug A is the same as drug B, is 0.0137.”
  - Awkward! Can't we just calculate the probability drug A is best?

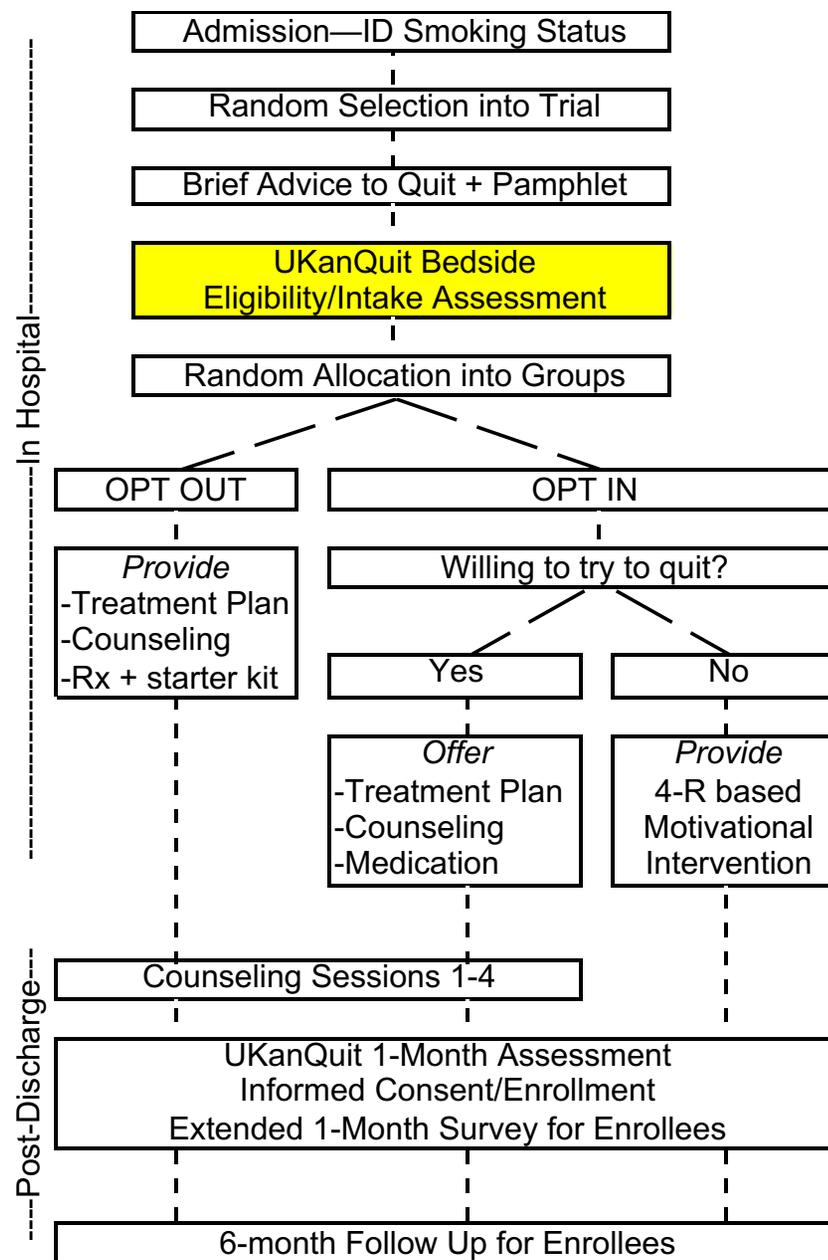
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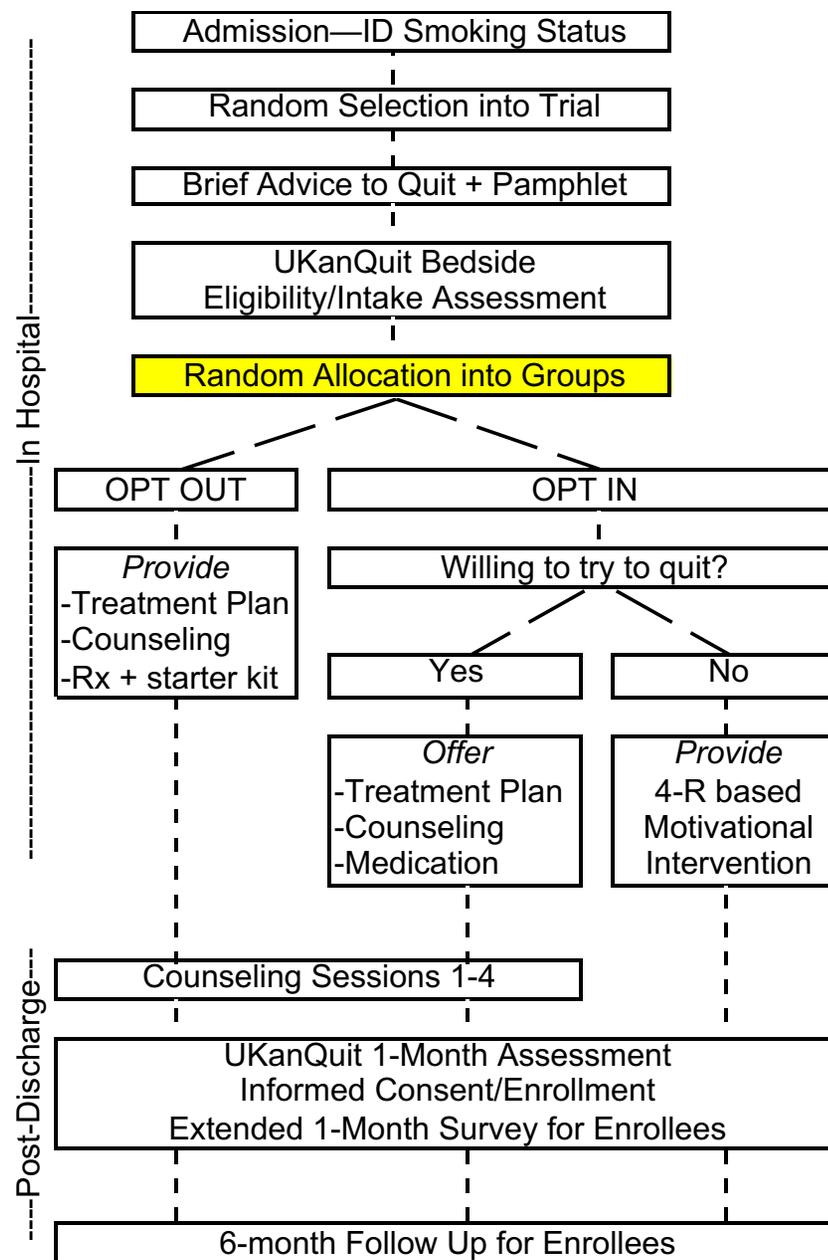
- Can't we just calculate the probability drug A is best?
- Bayesian posterior probabilities
  - **OPT OUT** has a .XX probability of being the best @ 1-month.
  - **OPT OUT** has a .XX probability of being the best @ 6-months.
  - Much clearer!!!

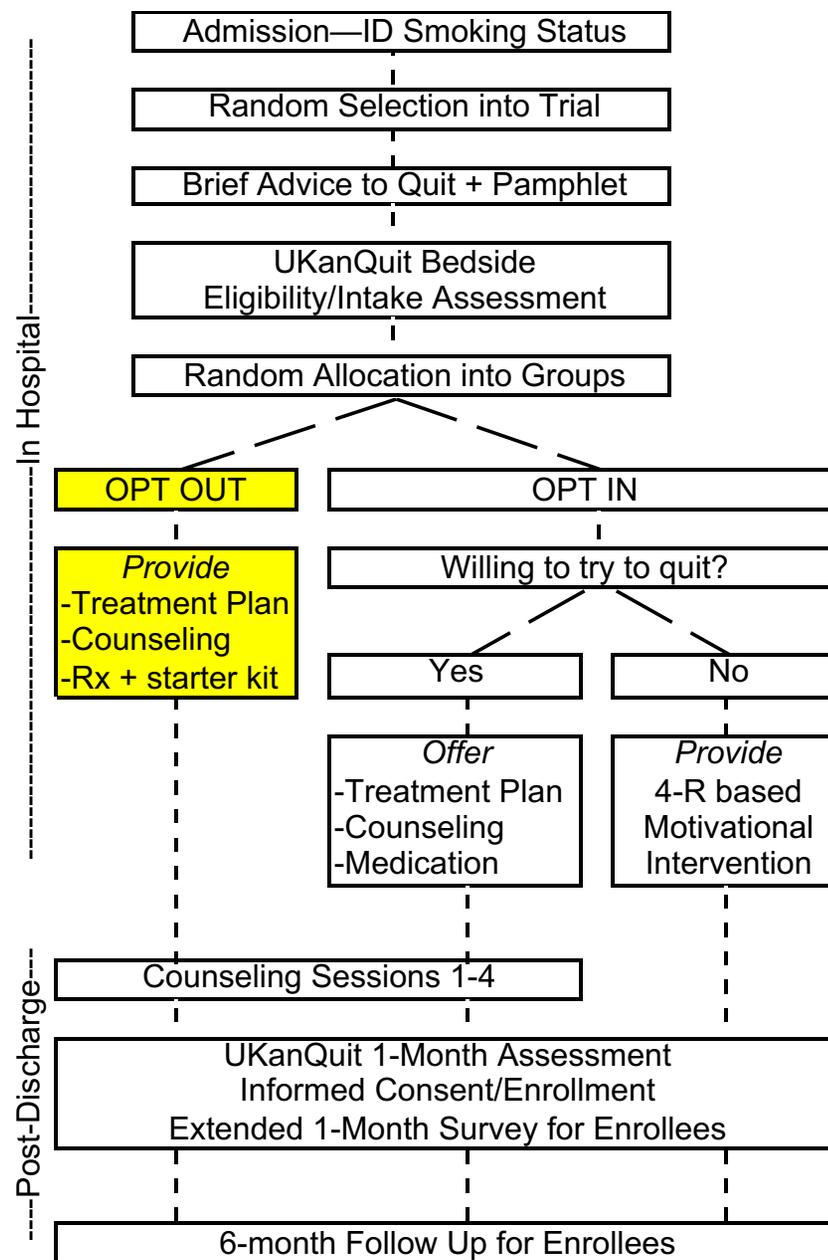


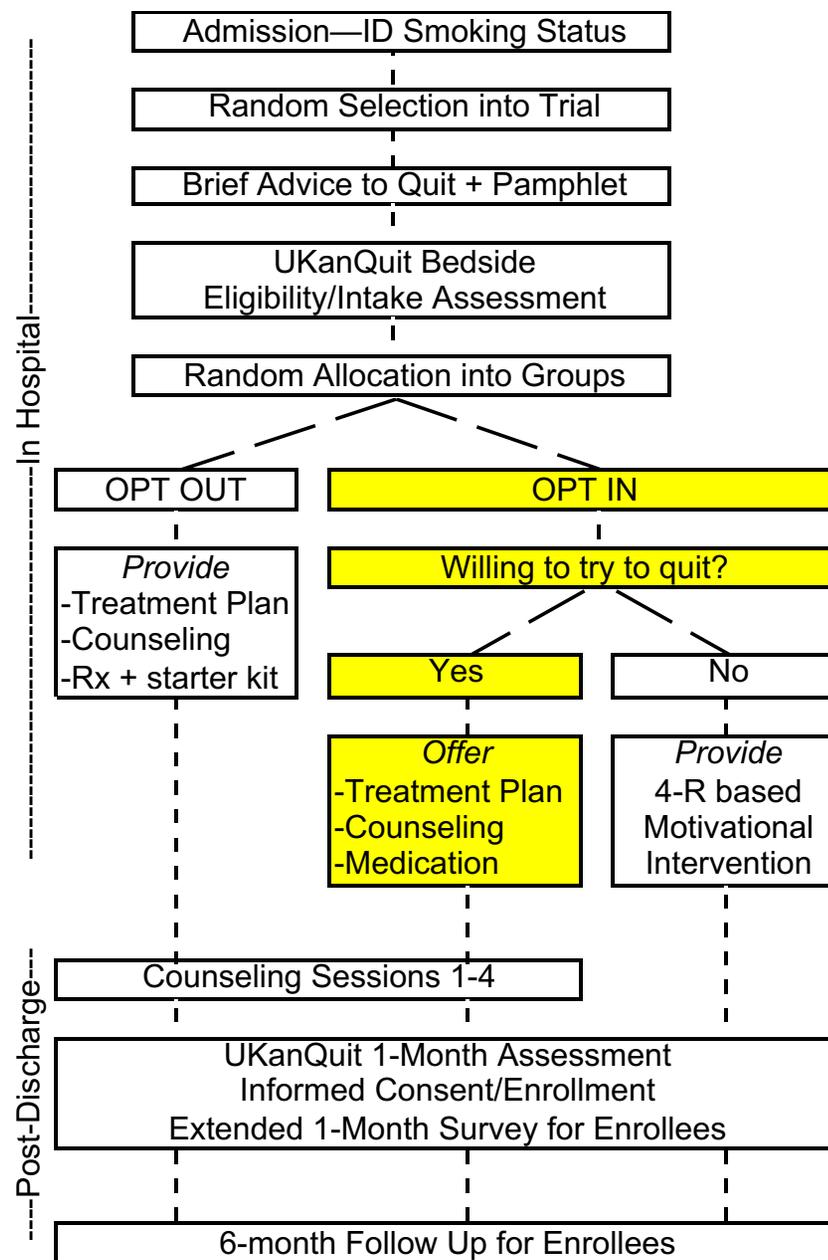


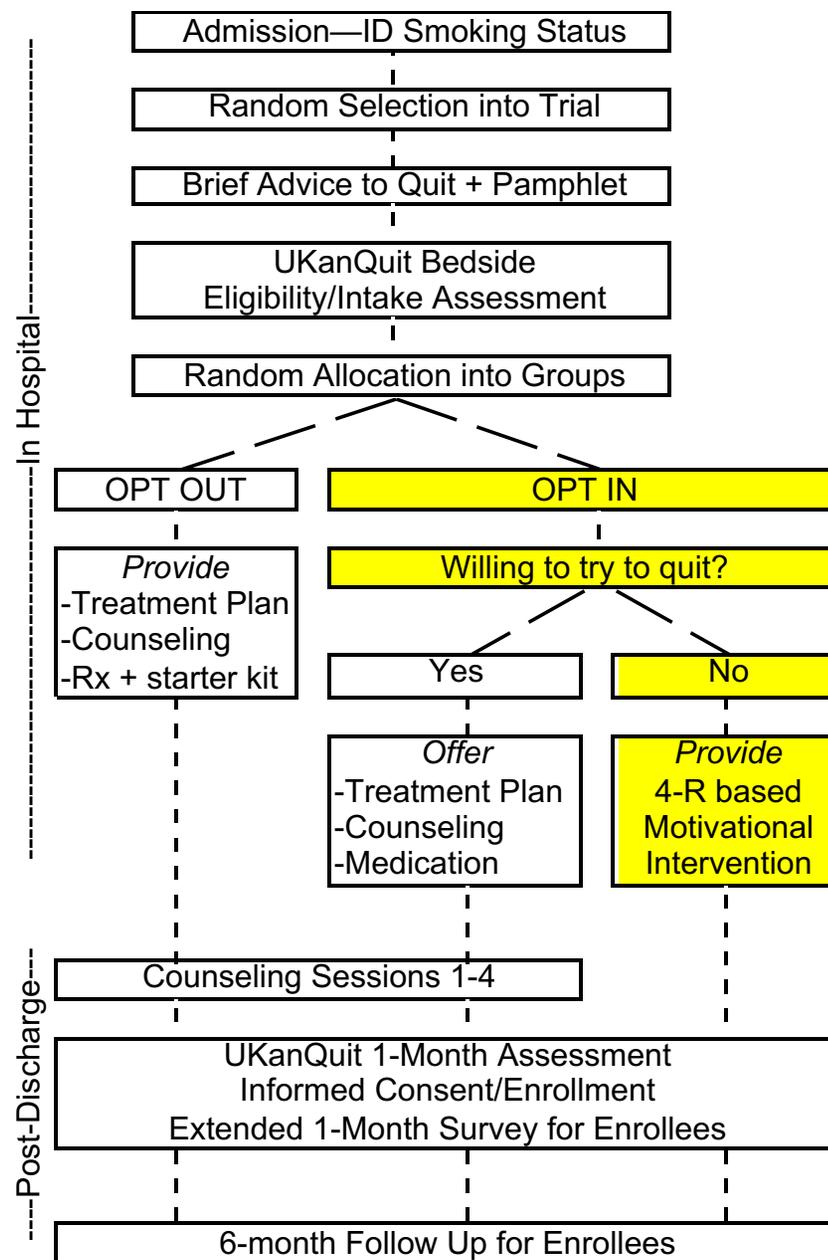


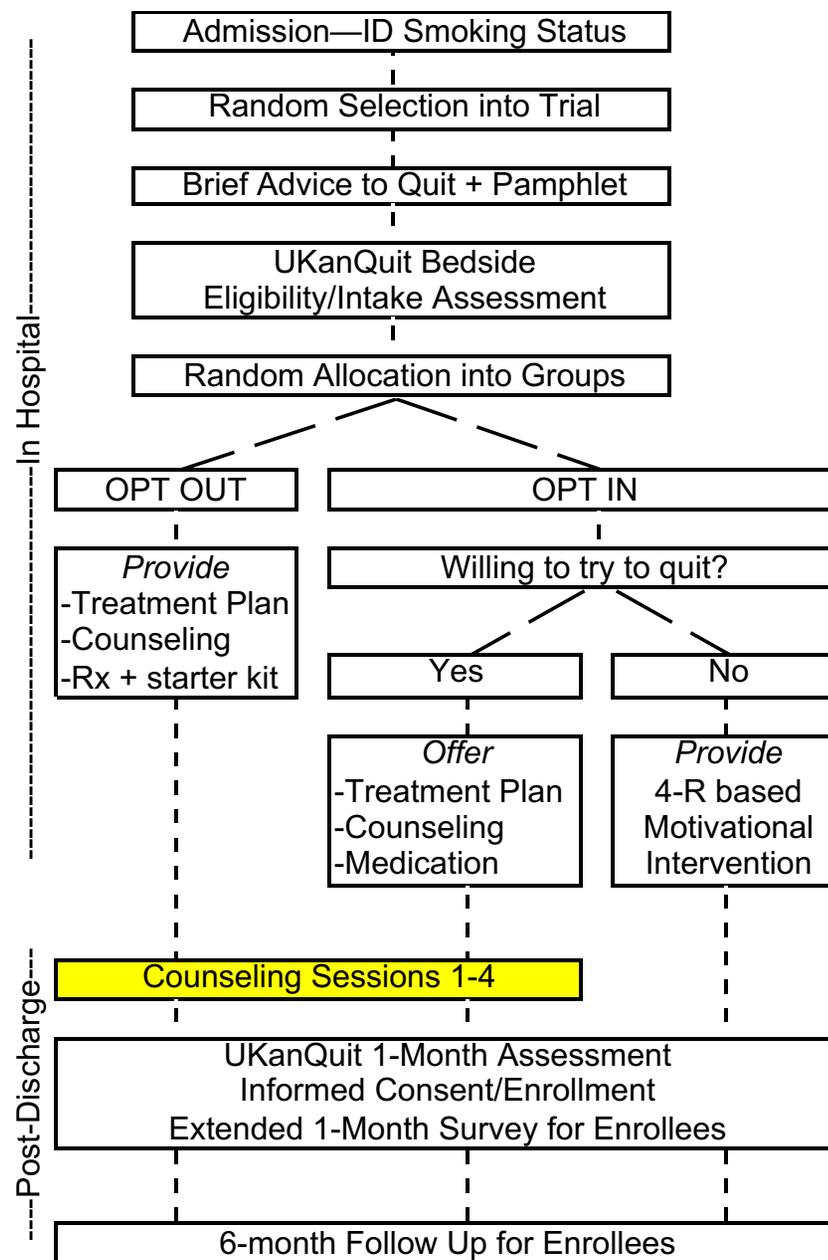


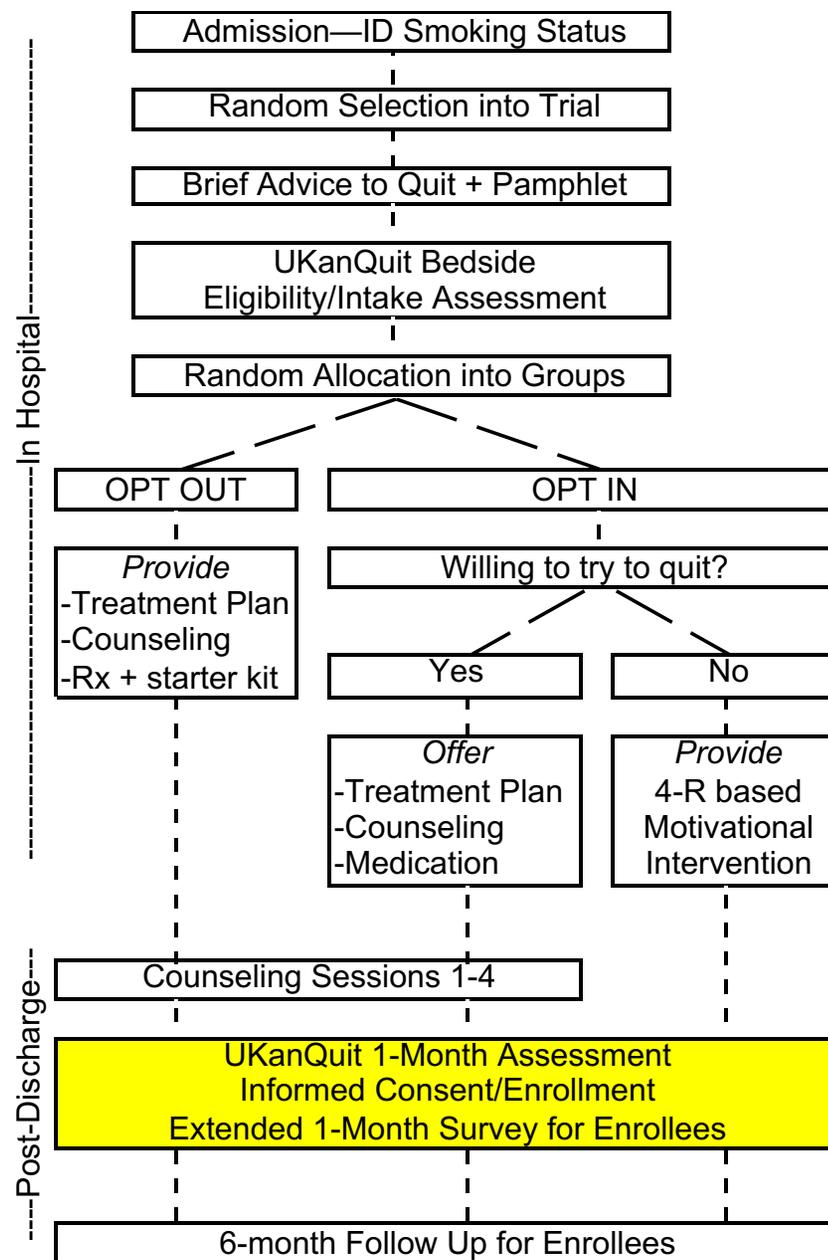


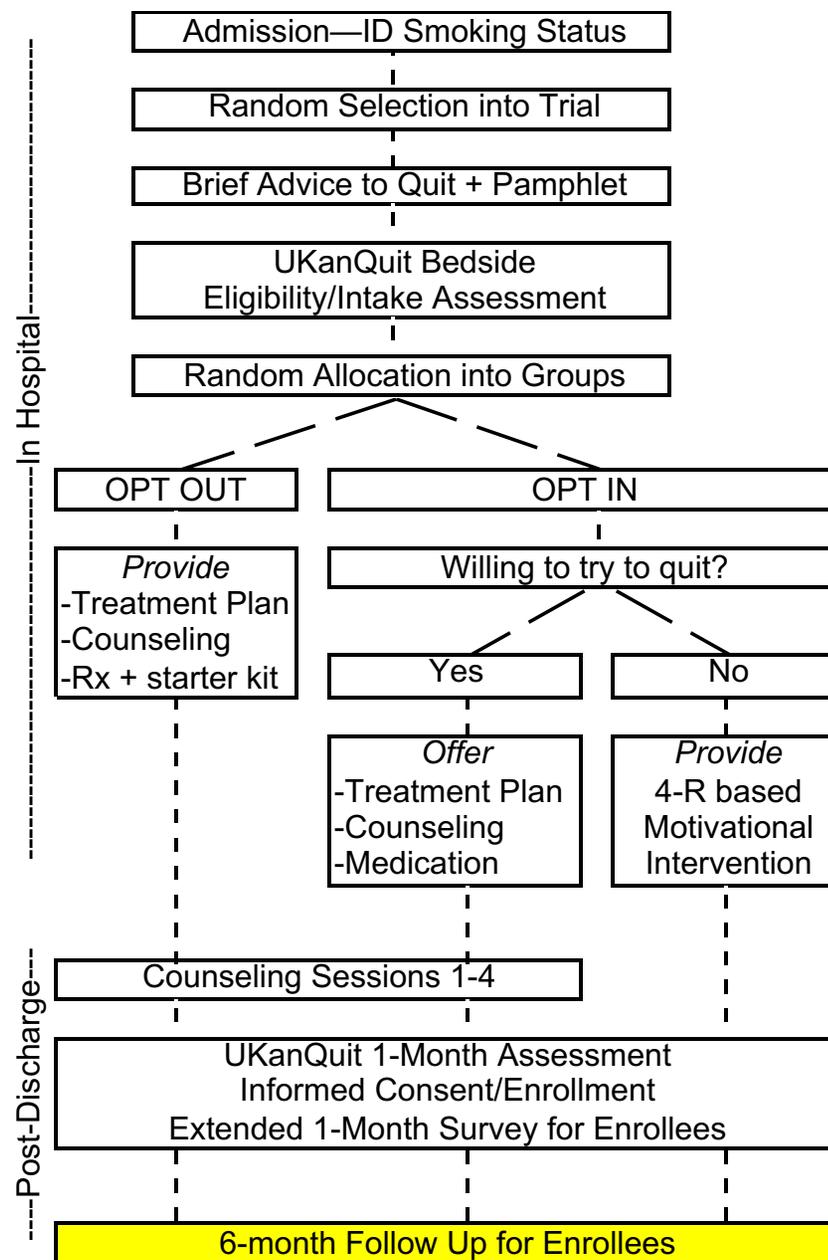












SCREENED 2,666

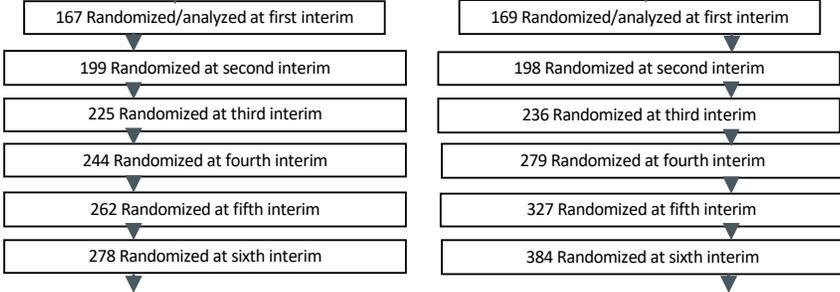


Patients screened  
(n = 2,666)

- Ineligible (n = 1,666)
- Medically ineligible for NRT (n = 550)
  - Speaks other language precluding enrollment\* (n = 47)
  - Physically/cognitively unable\* (n = 278)
  - No telephone (n = 139)
  - No secondary contact (n = 75)
  - Smoke <1 cpd (n = 417)
  - Using quit smoking medication prior to hospitalization (n = 121)
  - Received smoking cessation treatment or already participating in a clinical trial (n = 75)
  - Refused consult (n = 139)
  - Pregnant (n = 18)
  - Other (n = 112)
    - Incarcerated (n = 13)
    - Deceased (n = 3)
    - Screened more than twice (n = 95)
    - Household member enrolled in CTD (n=1)

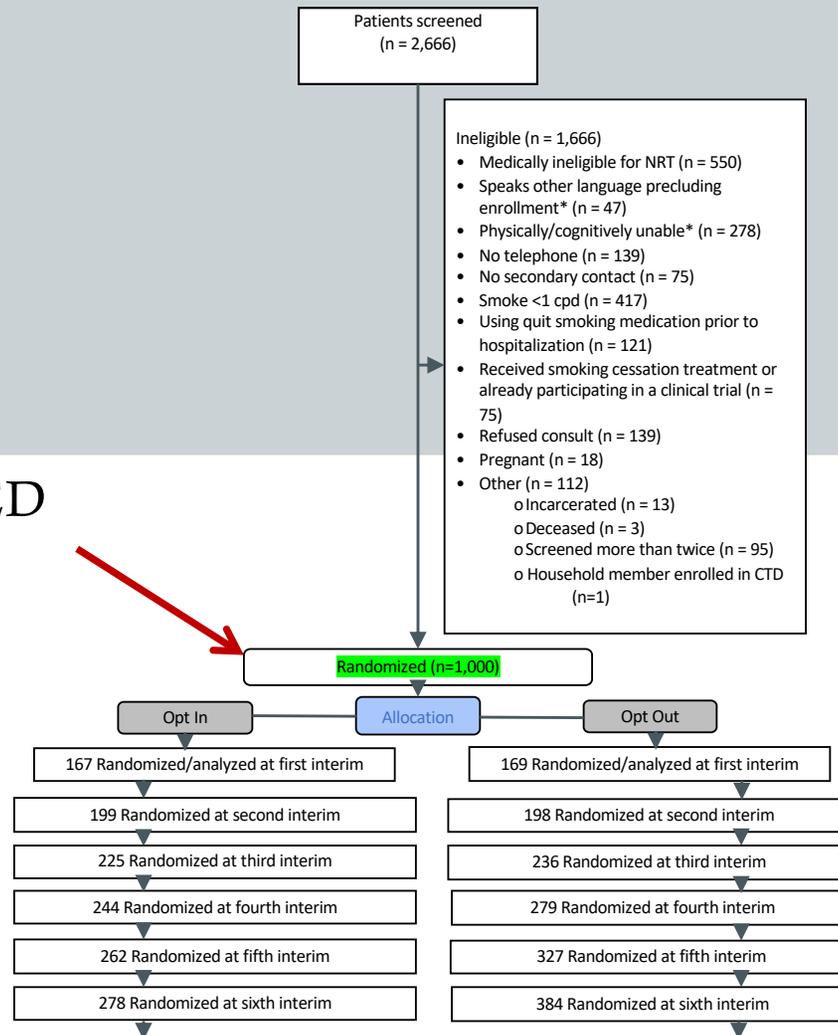
Randomized (n=1,000)

Opt In Allocation Opt Out



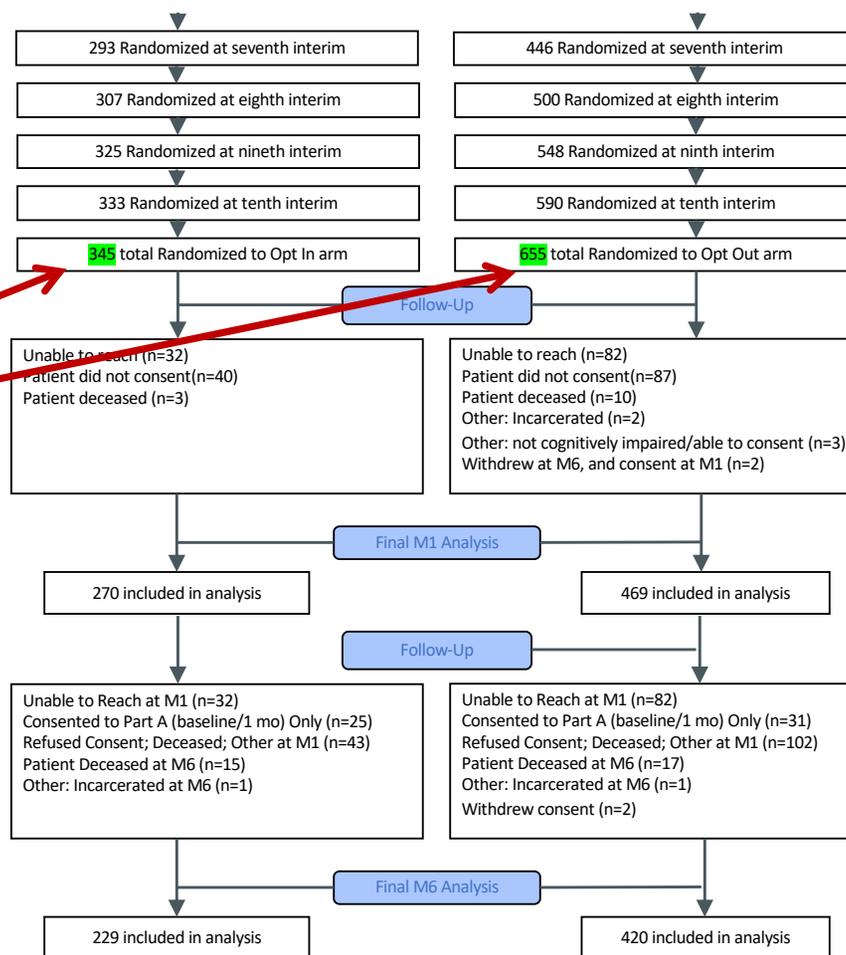
Protocol added 6-months co-primary outcome

RANDOMIZED  
1,000



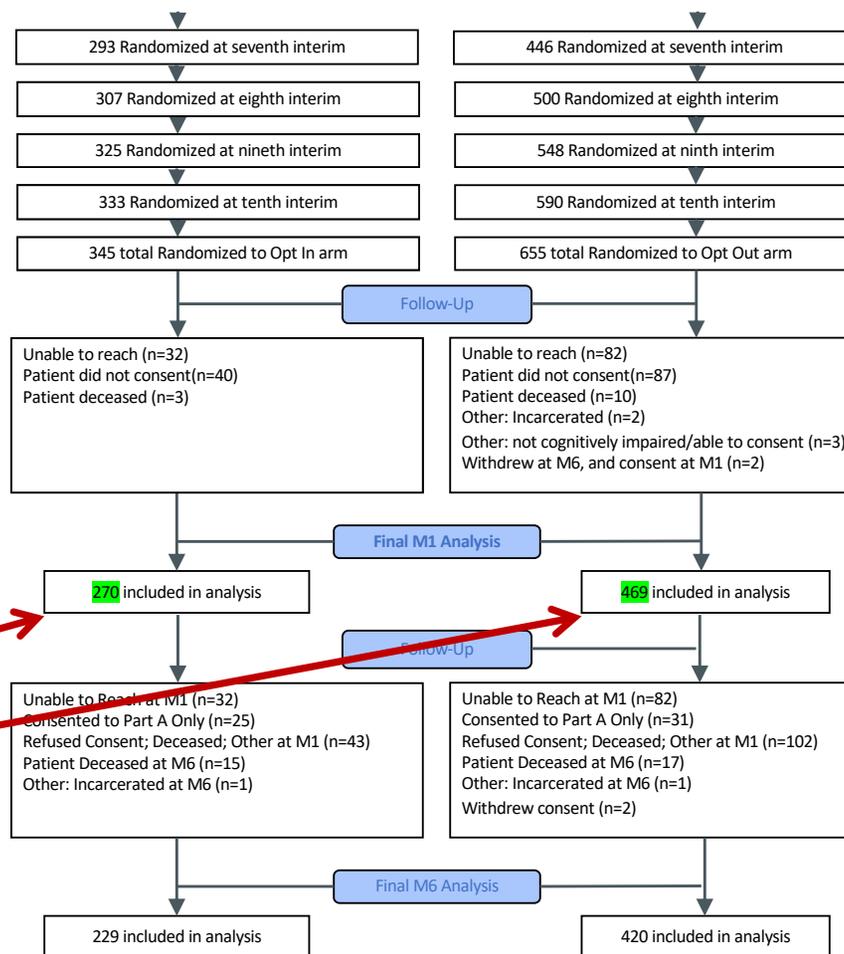
Protocol added 6-months co-primary outcome

345 OPT IN  
655 OPT OUT

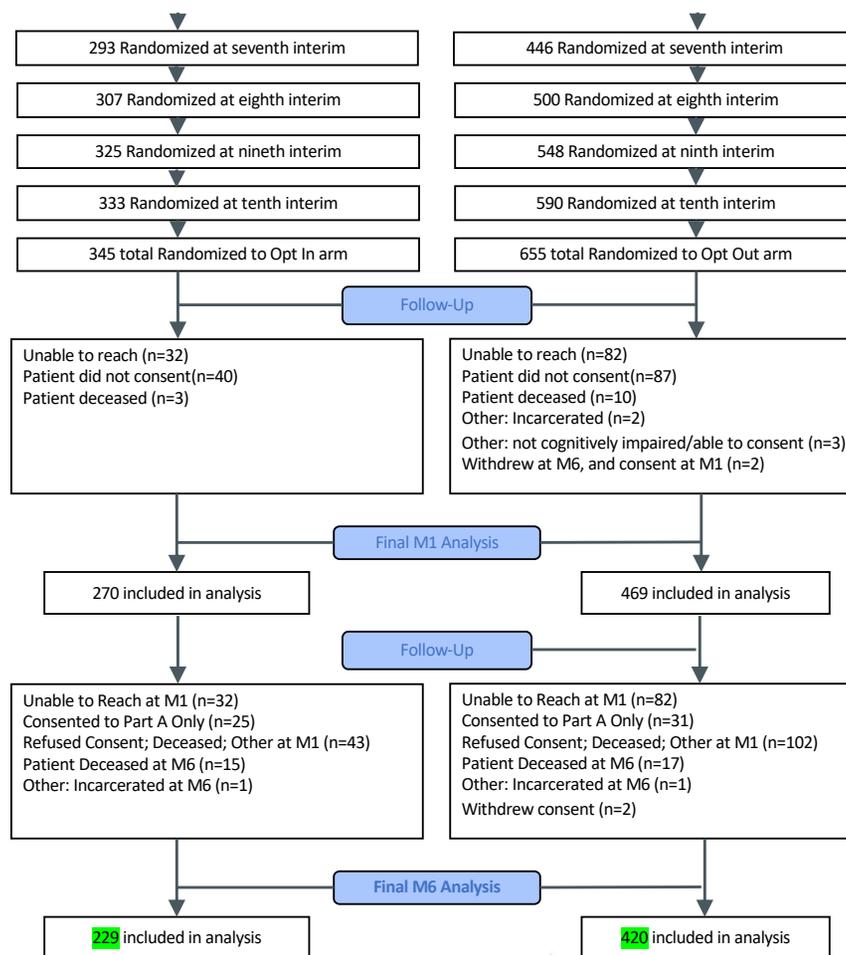


CONSENTED  
&  
ENROLLED

270 OPT IN  
469 OPT OUT



739 = Study sample, main outcomes  
74% of randomized included in trial



229 OPT IN

420 OPT OUT

649 = Study sample, 6 month outcomes

# Table 1 (N=739)

	Opt In (270)	Opt Out (469)	d <sup>^</sup>
<u>Demographics*</u>			
Age (mean)	51.7	51.2	0.03
Female	45.6	48.2	0.06
Non-Hispanic White	58.5	57.6	0.02
Medicaid Primary insurance	18.9	21.1	0.08
<u>Smoking Behavior</u>			
HSI (heaviness of smoking index, mean)	2.5	2.2	0.17
Willing to stay quit post-discharge	64.1	66.3	0.05
Used e-cigs, past 30 days	4.8	8.1	0.31

\*Percentages/n=739 unless otherwise noted

Cohen's d effect size: <0.2 negligible, 0.2-0.5 small, >0.5-0.8 medium, >0.8 =large

# 1 Month Main Outcomes (N=739)

OPT OUT improves 1-month quit rate compared to OPT IN

Abstinence Rates (95% Credible Interval)

Bayesian Posterior Probability  
Opt Out better than Opt In

Opt In

Opt Out

15.8 (11.8, 20.5)

21.5 (17.9, 25.4)

**.971**

# 6 Month Outcomes (N=649)

OPT OUT improves 6-month quit rate compared to OPT IN

Abstinence Rates (95% Credible Interval)

Bayesian Posterior Probability  
Opt Out better than Opt In

Opt In

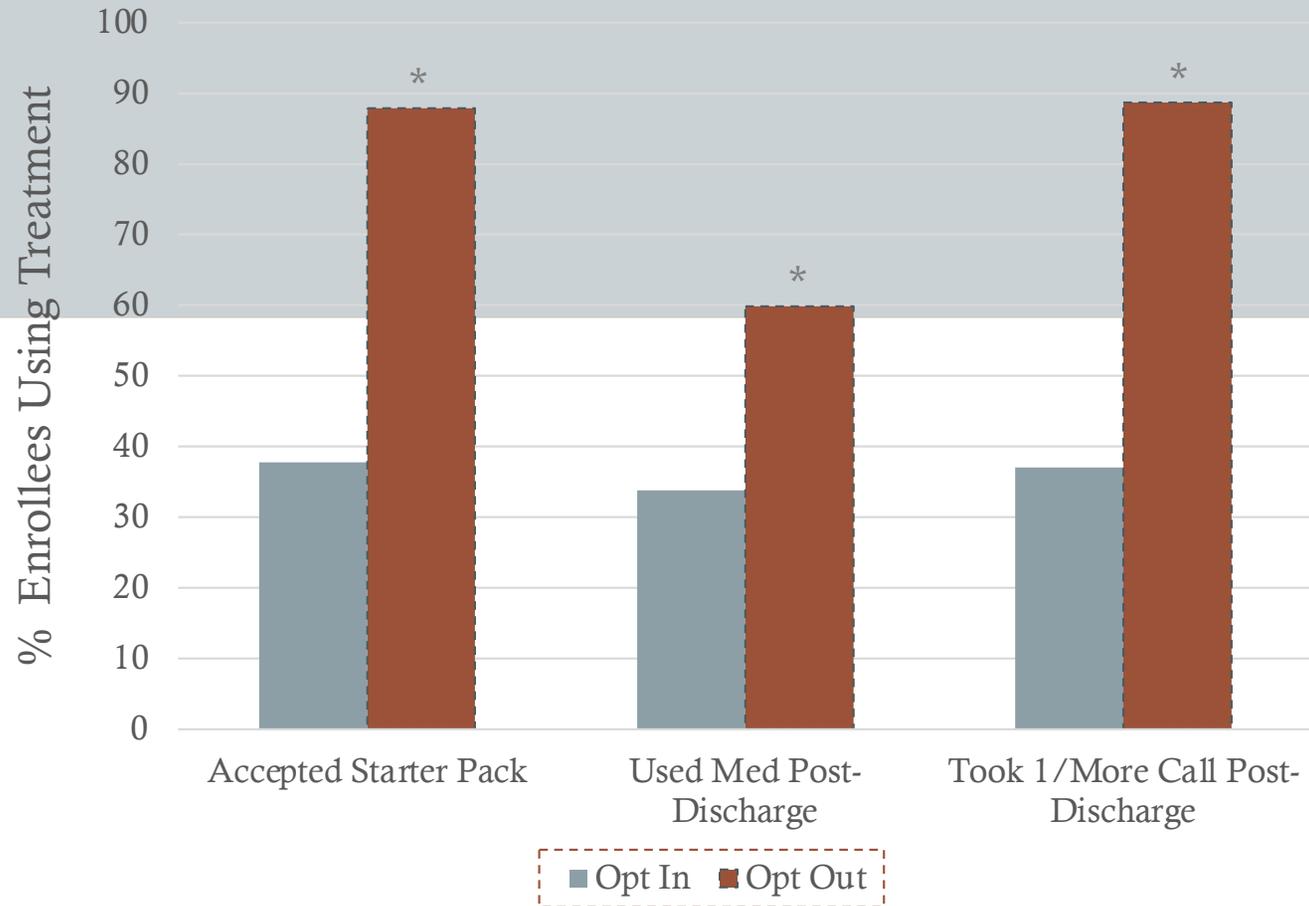
Opt Out

17.8 (13.2, 23.1)

18.5 (15.0, 22.4)

**.591**

# Medication & Counseling Use (N=739)

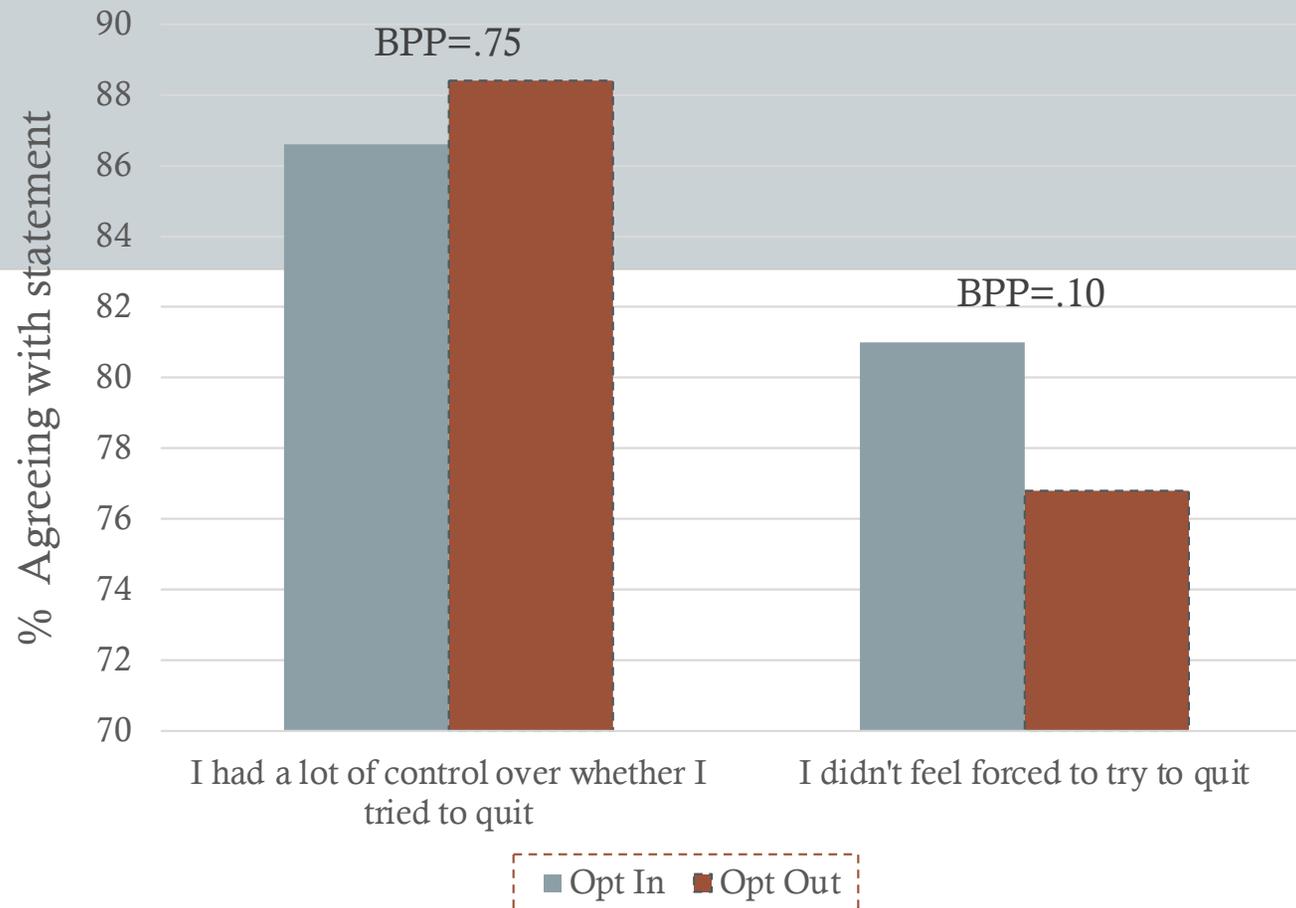


\*Bayesian Posterior Probability 1.00

# Perceived Coercion

- Perceived coercion items from the MacArthur Admission Experience Survey
- Did patients feel forced to quit in the opt-out arm, and if this affects treatment response
- At 1 month, we ask:
  - I had more influence than anyone else about whether I tried to quit
  - I had a lot of control over whether I tried to quit smoking
  - I chose to try to quit smoking
  - I felt forced to try to quit smoking
  - It was my idea to try to quit smoking

# Control, Feeling Forced to Quit (N=739)



# Cost Effectiveness

The incremental cost effectiveness ratio (ICER) was \$678.6, which represents the cost of getting one more person to quit in the opt out condition.

	<b>Counseling (mean \$ per/person)</b>	<b>Starter pack (mean \$ per/person)</b>	<b>Sum (mean \$ per/person)</b>	<b>Verified quit n (%)</b>
<b>Opt in (n=270)</b>	15.35	21.46	36.81	43 (15.9%)
<b>Opt out (n=469)</b>	25.19	49.62	74.81	101 (21.5%)
<b>Difference</b>	9.84	28.16	38	5.6%
<b>ICER</b>				<b>\$678.6</b>

# ITT Quit Rates (N=1,000)

OPT OUT improves 1-month but not 6-month quit rate compared to OPT IN

	Abstinence Rates (95% CI)		Bayesian Posterior Probability Opt Out better than Opt In
	Opt In	Opt Out	
Month 1	12.4 (9.2, 16.1)	15.4 (12.8, 18.3)	<b>.902</b>
Month 6	11.8 (8.7, 15.5)	11.9 (9.5, 14.5)	<b>.512</b>

# Ethics of Opt-Out Care

- Default treatment is coercive and/or paternalistic
- Which is more paternalistic?
  - Asking if they're ready and only offering meds/counseling if they say they are "ready"
  - Giving meds/counseling, letting patient decide if they want/not
    - What happens in most medical care
- Where there is strong evidence that support a given therapy, the default should be set to that therapy
- Defaults should be options that make the choosers better off, as judged by themselves
  - 70% of smokers *want* to quit, even if they're not ready/willing now

# Stand to Gain

If Opt-Out proves more effective:

- Free to deliver care to 3x-5x as many smokers (20%-100%)
- Simplify treatment algorithm – don't have to ask/judge if patient is ready or willing
- No excuses for not treating
  - 2836 European physicians – 2 top barriers to treating:
    - patients' lack of willpower and low interest in quitting
- Reduce tobacco use rates, illnesses, deaths, costs

# Population Impact

## Selective vs Universal Treatment

- In a population 100 people – which is better?
  - 50% quit rate among 20% of people?
  - 20% quit rate among 100% of people?
- Even if a lower percentage quits, if you spread effective treatment across a broader population, you can get greater numbers of quits

# The Single Biggest Barrier to Providing Treatment



# Conclusions

- Compared to OPT IN, OPT OUT
  - High probability of improving the quit rate at 1-month
  - Low probability of improving the quit rate at 6 months
- OPT OUT outperformed OPT IN:
  - medication utilization
  - counseling utilization
  - sense of control over quitting
- At a much lower cost than cancer treatments, and comparable cost to other cessation interventions

# Discussion

- OPT OUT approach did not result in better rates of long-term abstinence
  - Ditch this approach?
  - What would trials of other treatments, in other health areas (asthma?) conclude?
- Population-based trials, and trials of brief interventions, might benefit from delayed consent
- Adaptive trials can get more patients the effective treatment and yield results faster
- Bayesian analysis is new, simpler, and unknown to researchers/reviewers...

# The Team

Edward F. Ellerbeck

Babalola Faseru

Delwyn Catley

Byron Gajewski

Tresza Hutcheson

Taneisha Scheuermann

Theresa I. Shireman

Chuanwu Zhang

Laura Martin

Jinxian Hu

Laura Mussulman

Niaman Nazir

Elena Shergina

Andrea Elyachar

Marjorie Cooper

Lety Sarmiento

Craig Warlick

Vivek Patel

Genevieve Casey

Alison Summers Hageman

# Implementation Science & Equity

*Center on Biomedical Research Excellence [Score 2.0]*

Kimber Richter, PhD MPH

Christie Befort, PhD

- ❖ Bench – bedside time lag: 17-years
  - 30–40% of patients fail to receive care consistent with current evidence
  - 20-25% of care that is provided is not needed or is potentially harmful

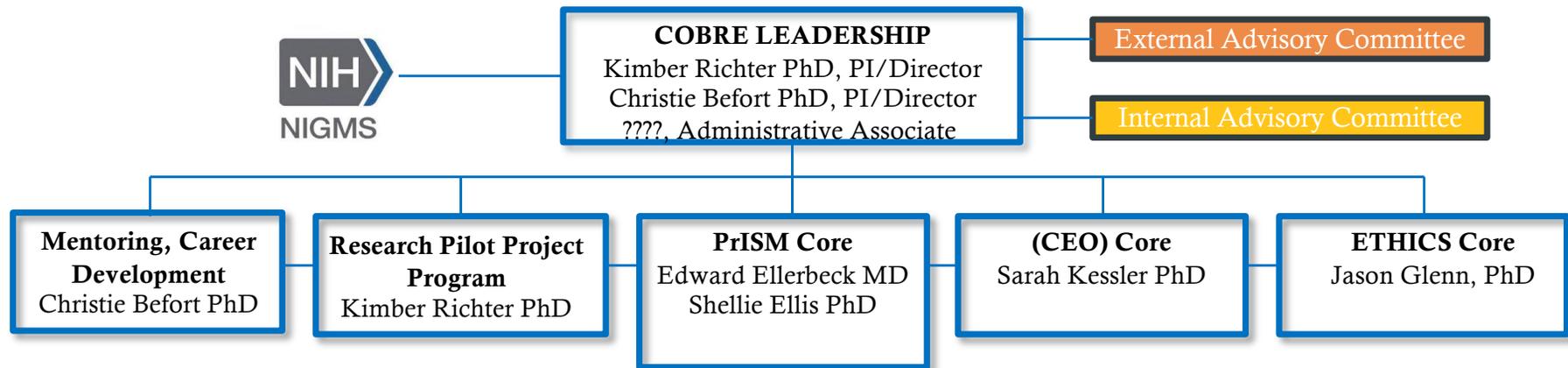


# ISE COBRE

## *Overall*

- ❖ Provide scientific mentoring and institutional support for implementation science and equity
- ❖ Provide core infrastructure in methodology, engagement and ethical issues in human subjects' research
- ❖ Select and train outstanding, multi-disciplinary senior and early-career faculty

# ISE COBRE Organizational Structure



# Key Core Functions

## *Administrative Core*

- Mentoring & Career Development Plans
- Evaluation

## *Community Engagement & Outreach*

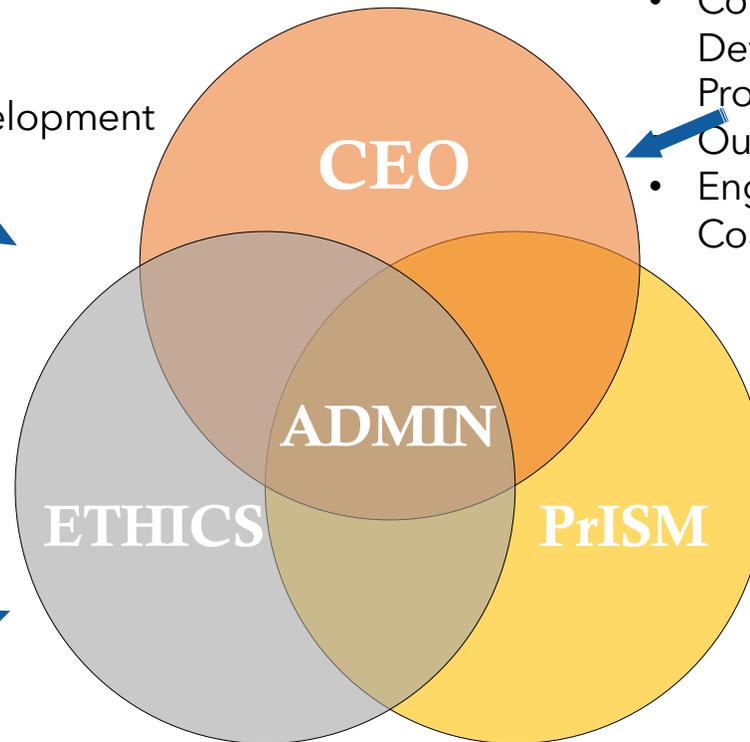
- Community Advisory Board Development for Research Projects
- Outreach Services
- Engagement & Outreach Coaching

## *Pragmatic Implementation Science Methods*

- Study Design
- Training Implementation
- Methods Support

## *ETHICS, Human Subject & Regulatory*

- Ethics Consultation
- Navigation
- Ethics/Compliance Training

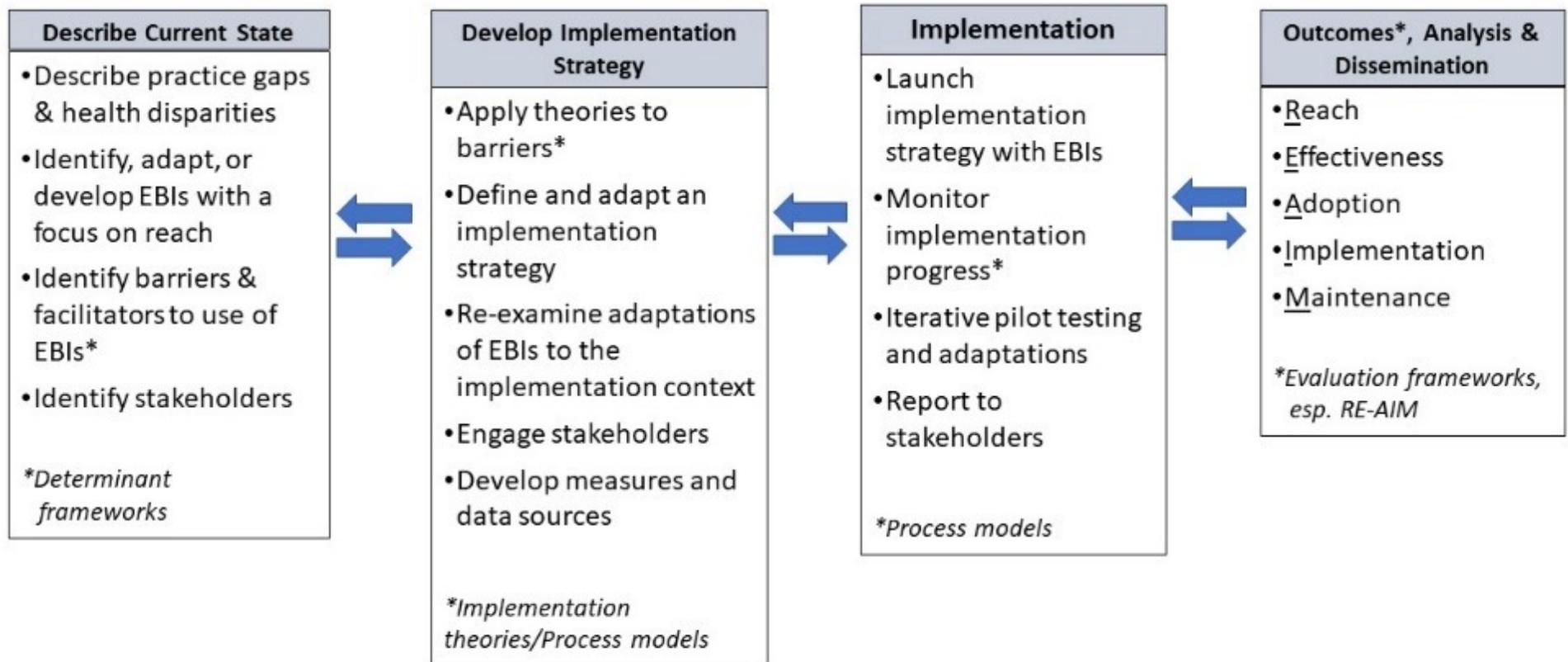


# ISE COBRE Pilot Projects 4 of 5

Project	Lead (Department)	Mentors
Nutricity: A mHealth nutrition intervention to improve diet quality among Latino children	Heather Gibbs, PhD RD LD (Dietetics & Nutrition, SHP)	Debra Sullivan, PhD RD Jamie Zoellner, PhD RD (UVA)
Implementing Advance Care Planning as a Healthy Aging Activity in Rural Primary Care	Heather Nelson-Brantley, PhD RN NEA-BC CCRN-K (SON)	Christie Befort, PhD Barb Polivak, PhD RN FAAN Terri Fried, MD (Yale)
Preliminary Studies on Implementation of Smoking Cessation Interventions for Low-Income Women	Taneisha Scheuermann, PhD (Population Health, SOM)	Kim Richter, PhD MPH Ross Brownson, PHD (Wash U)
Improving the Quality of Prenatal Care for Low-Income, Black Women	Sharla Smith, PHD MPH (Population Health, SOM)	Megha Ramaswamy, PhD MPH Kevin Ault, MD FACOG, FIDSA

# PrISM

Figure B. Stepped Approach to PrISM support of Implementation Research



# PrISM Core—Tools & Resources

- ❖ Equity focused frameworks
  - PRECIS-2 Tool
- ❖ Clinical informatics
  - HERON
  - Greater Plains Collaborative
- ❖ Biostatistics and data analysis
  - REDCap
  - Velos eResearch
- ❖ Qualitative methods, instrument development and mixed methods
  - ❖ Nvivo
  - ❖ Atlas Ti
  - ❖ Dedoose