

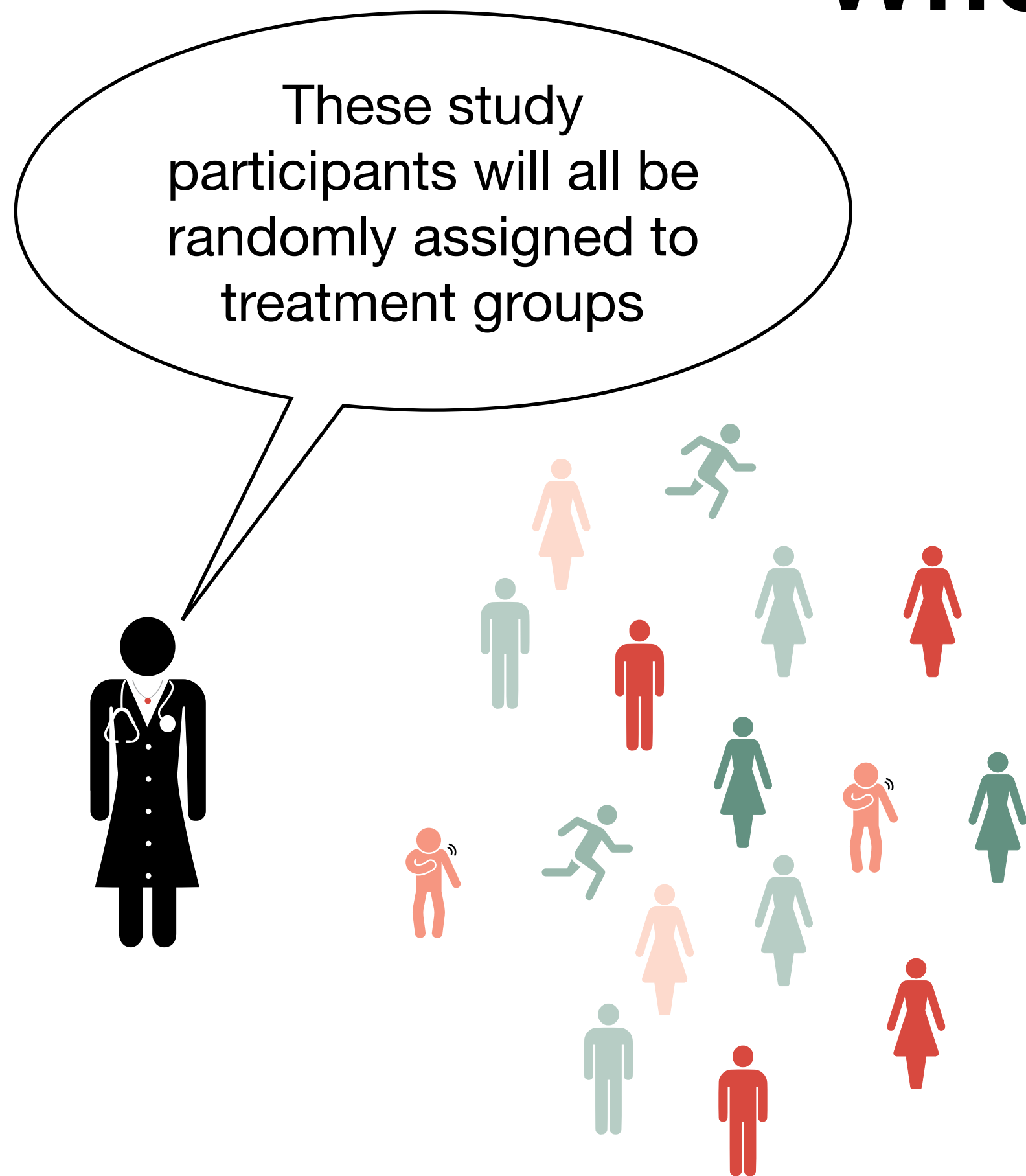
Randomized Trials & Quantifying Treatment Effects

Martina Steurer, MD MAS

10/31/19

With appreciation to Tom Newman & Michael Kohn

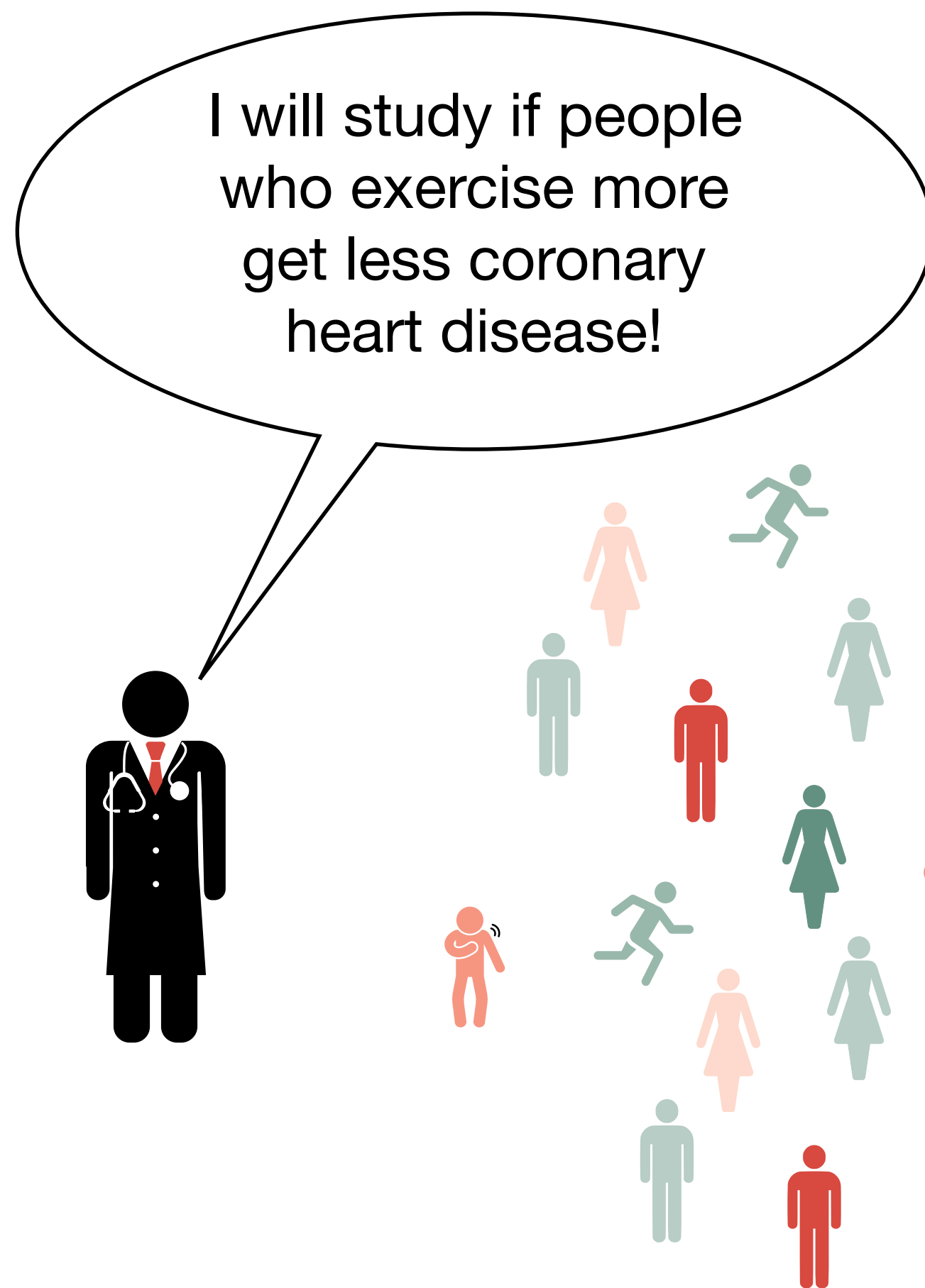
What is randomization?



Treatment/Intervention

No Treatment/Intervention

Why randomize?



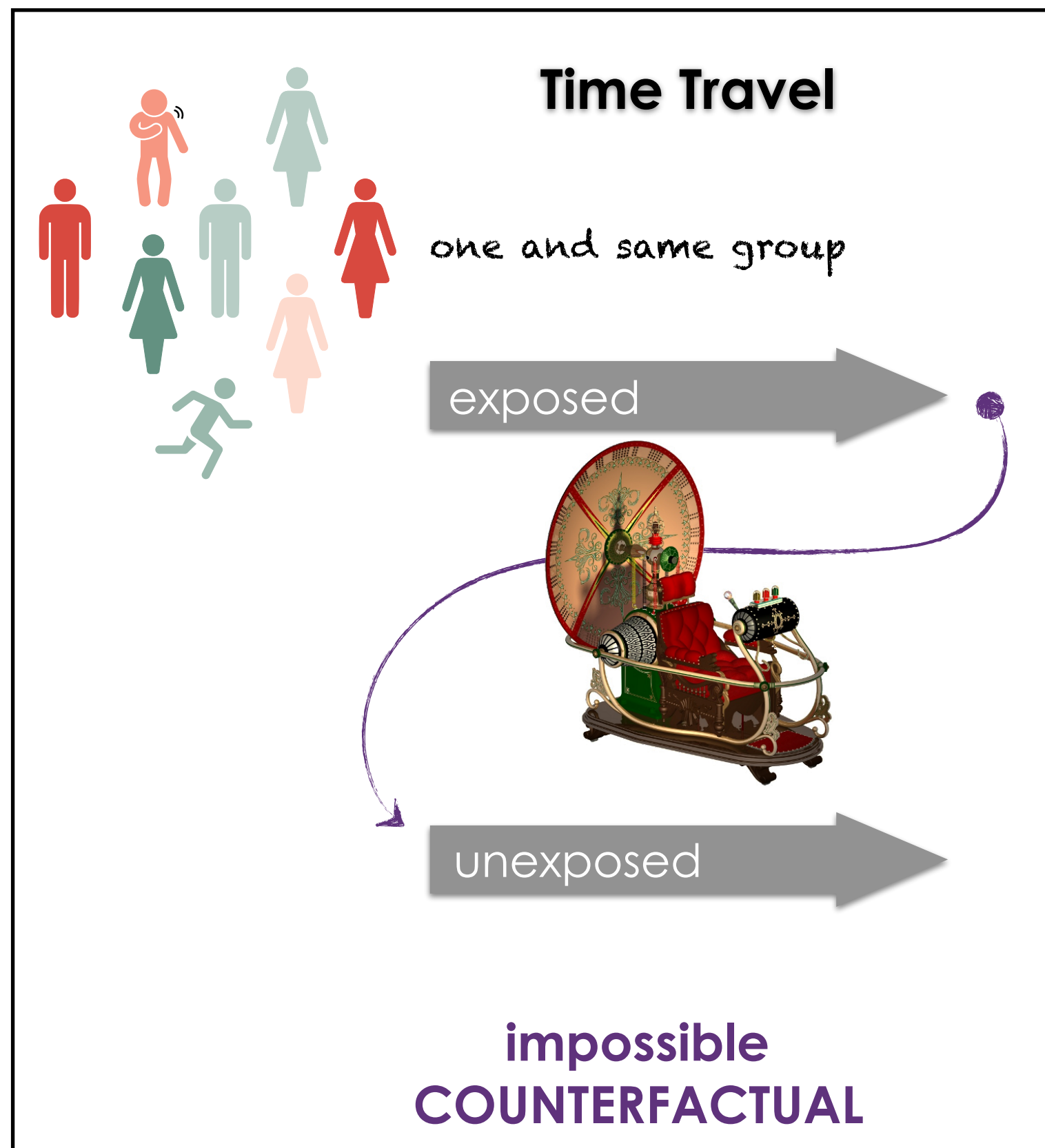
Exercise

No Exercise

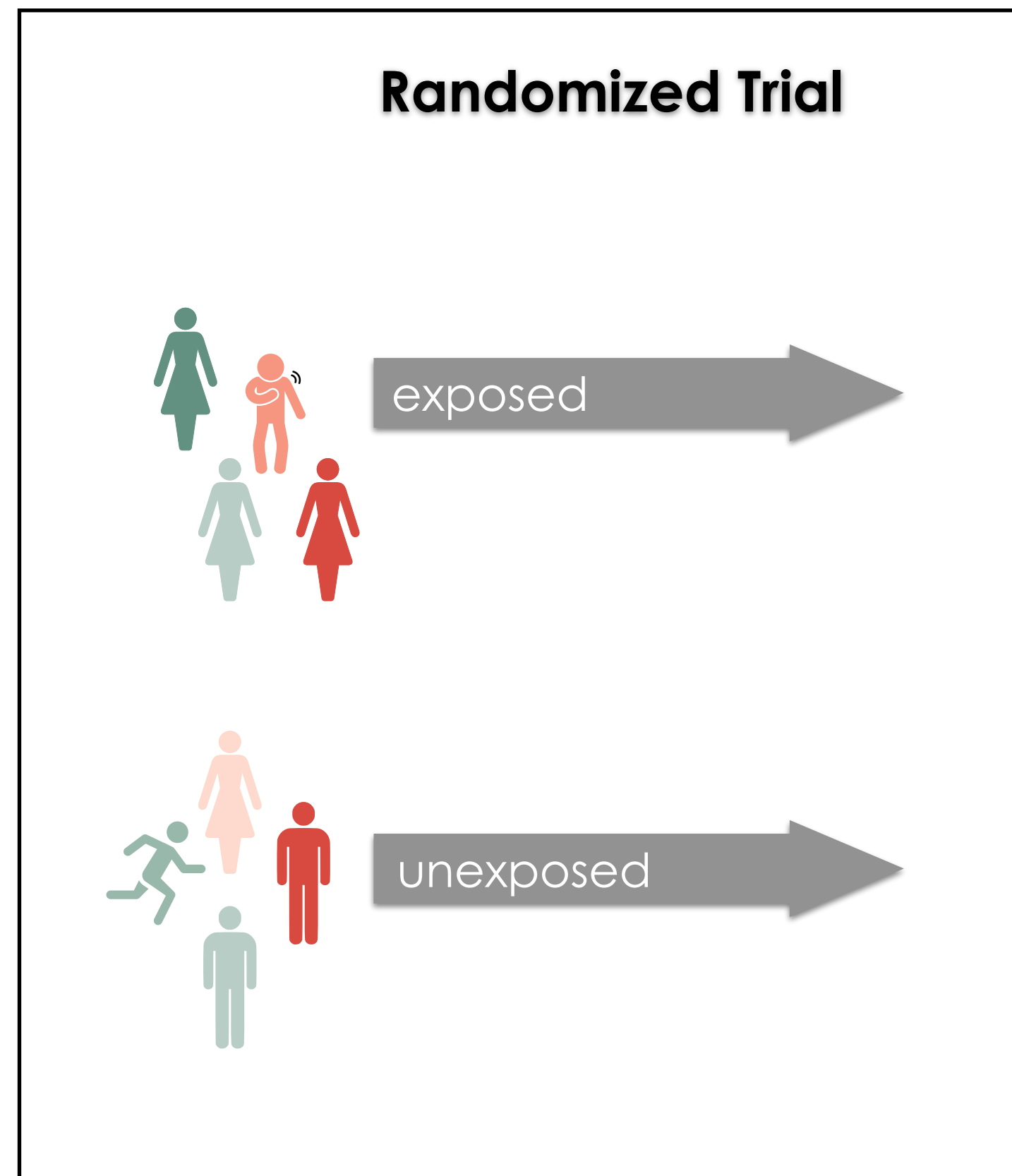
Where is the problem with this study design?

Why randomize?

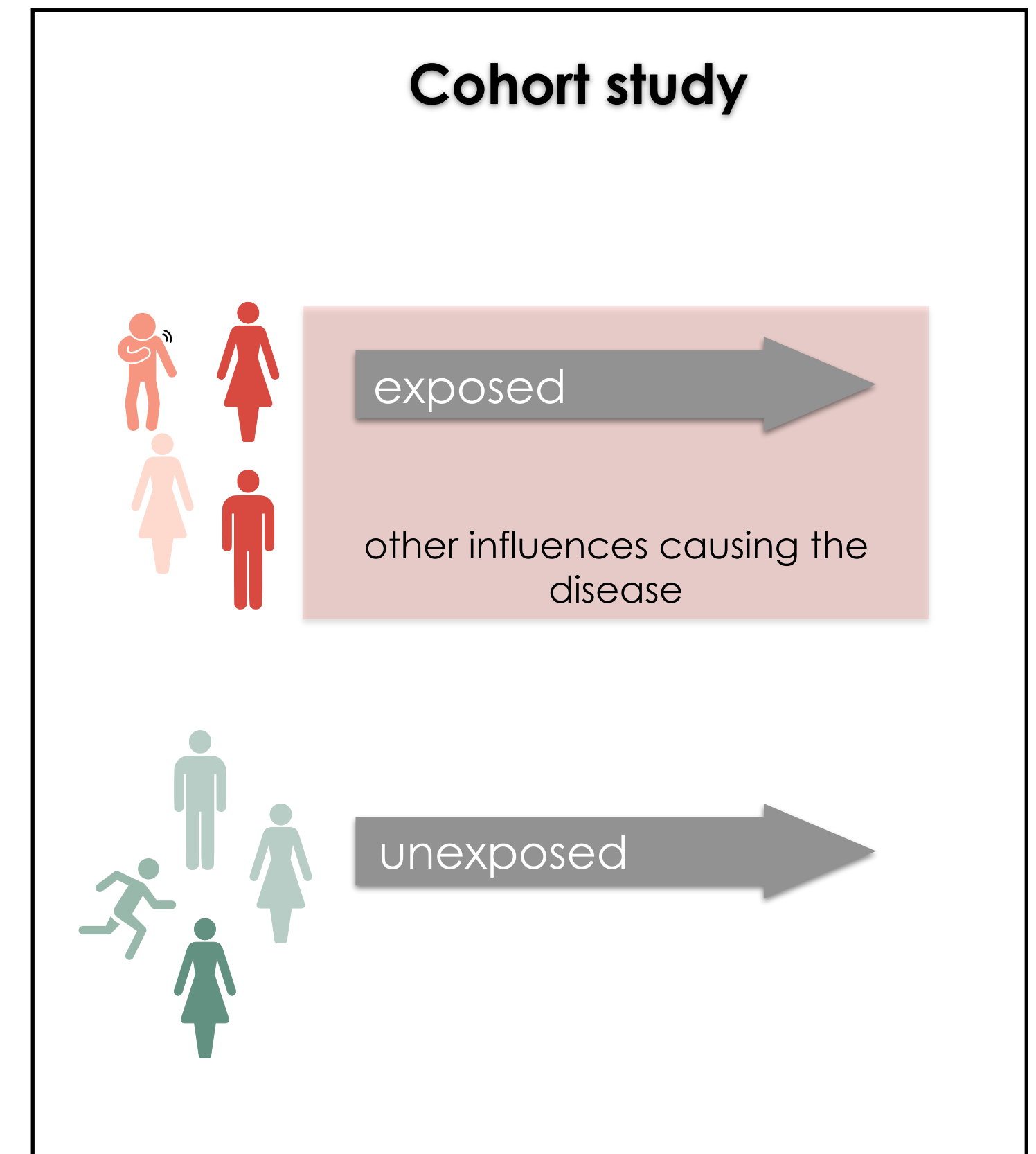
In order to proof that an effect is causal, the two populations (i.e. exposed and unexposed) need to be **EXCHANGEABLE**



We will let you know immediately when we figured this out, should be soon....



TODAY in EPI 204



In order to get at causal interference in cohort studies, you have to deal with confounding, you will learn this in EPI 203

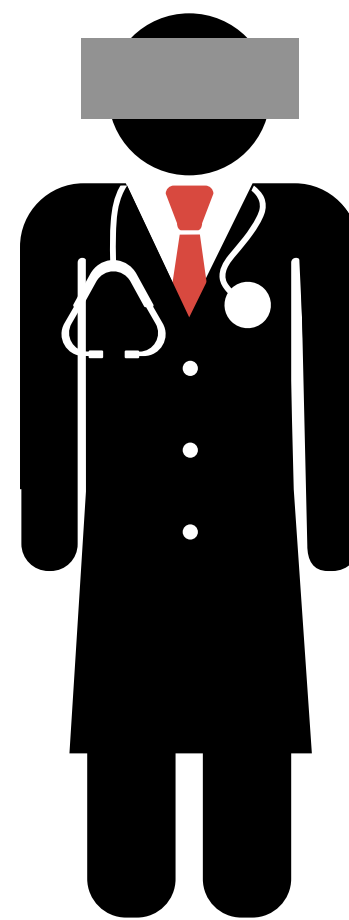
Blinding

Patient



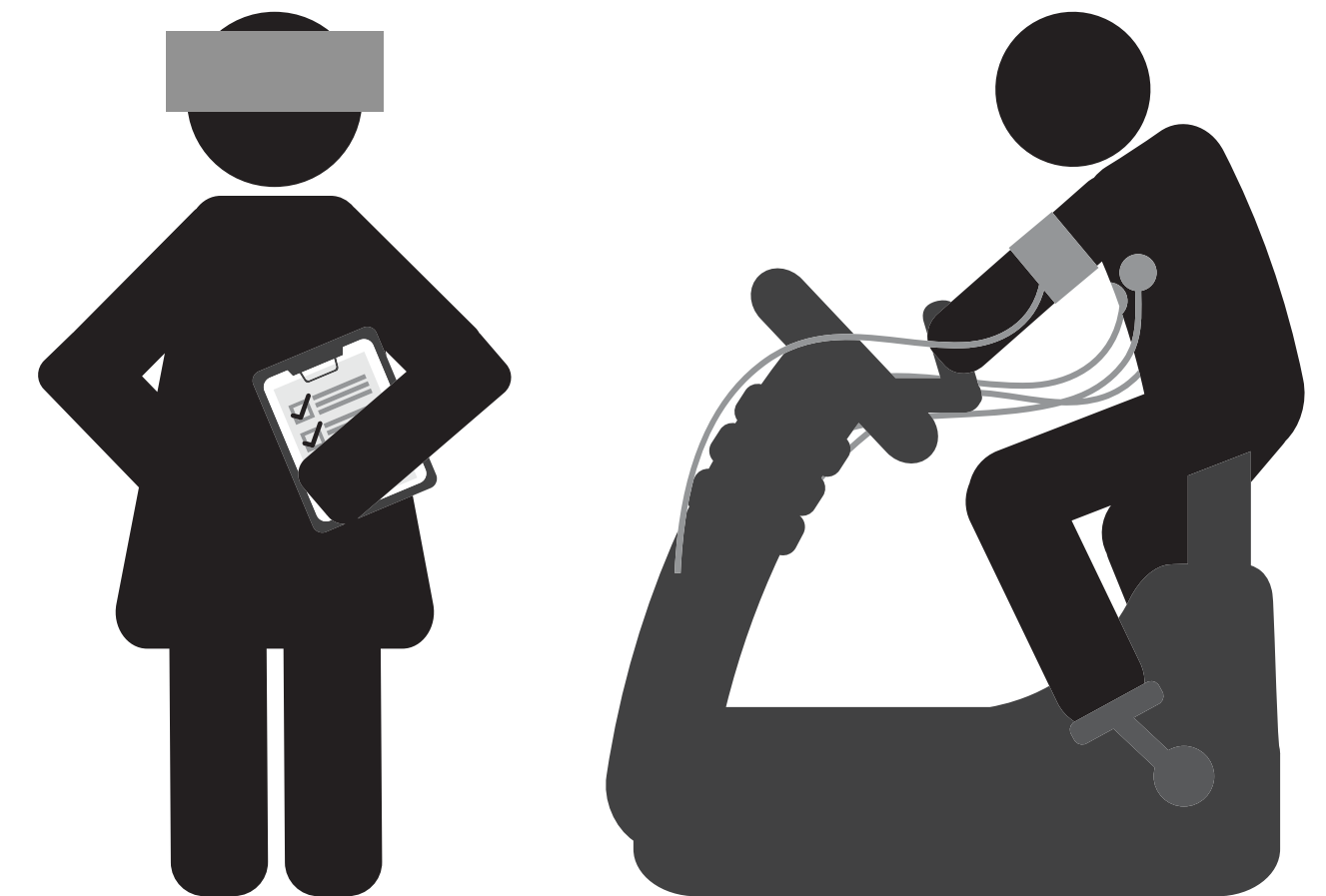
- Prevents differences due to placebo effect
- Very important for subjective measures
- Getting study drug outside of the trial (controls)

Provider



- Helps avoid co-interventions

Person assessing the outcome



- Prevents observer bias
- More important for subjective measures
- Not so important for objective measures

What about mortality?

Blinding

Patient



The NEW ENGLAND
JOURNAL of MEDICINE

ORIGINAL ARTICLE [FREE PREVIEW](#)

Arthroscopic Partial Meniscectomy versus Sham Surgery for a Degenerative Meniscal Tear

Raine Sihvonen, M.D., Mika Paavola, M.D., Ph.D., Antti Malmivaara, M.D., Ph.D., Ari Itälä, M.D., Ph.D., Antti Joukainen, M.D., Ph.D., Heikki Nurmi, M.D., Juha Kalske, M.D., and Teppo L.N. Järvinen, M.D., Ph.D. for the Finnish Degenerative Meniscal Lesion Study (FIDELITY) Group



Blinding

Person assessing the outcome

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January 01, 1994; 44 (1) [ARTICLES](#)

The impact of blinding on the results of a randomized, placebo-controlled multiple sclerosis clinical trial

J. H. Noseworthy, G. C. Ebers, M. K. Vandervoort, R. E. Farquhar, E. Yetisir, R. Roberts

First published January 1, 1994, DOI: <https://doi.org/10.1212/WNL.44.1.16>



Issues with blinding

Patients may be able to tell they are on drug



side effects

Part of the causal effect of treatment may be related to knowing you are on treatment

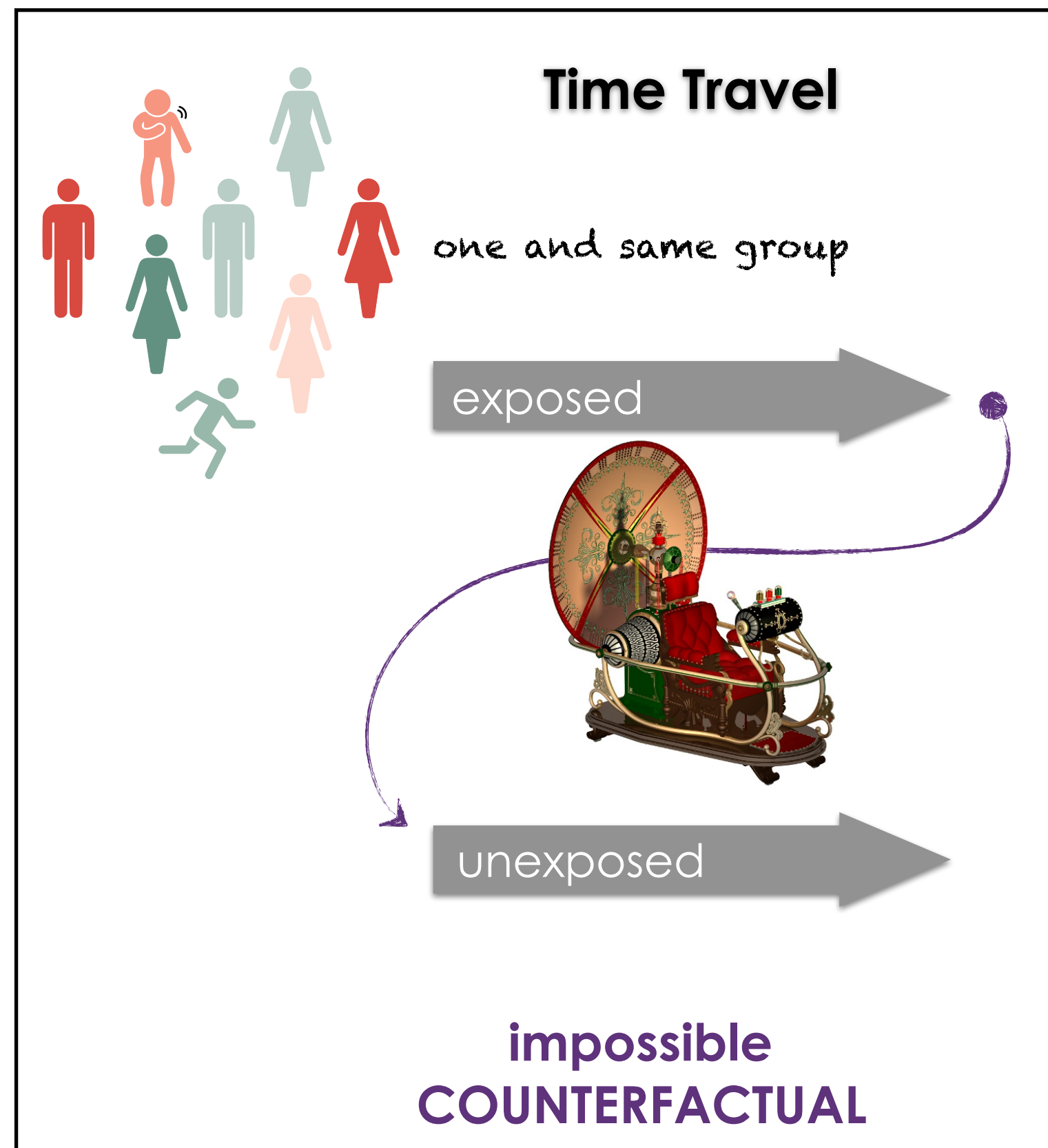


treatment looks better i.e. Meniscus surgery



Issues with RCTs?

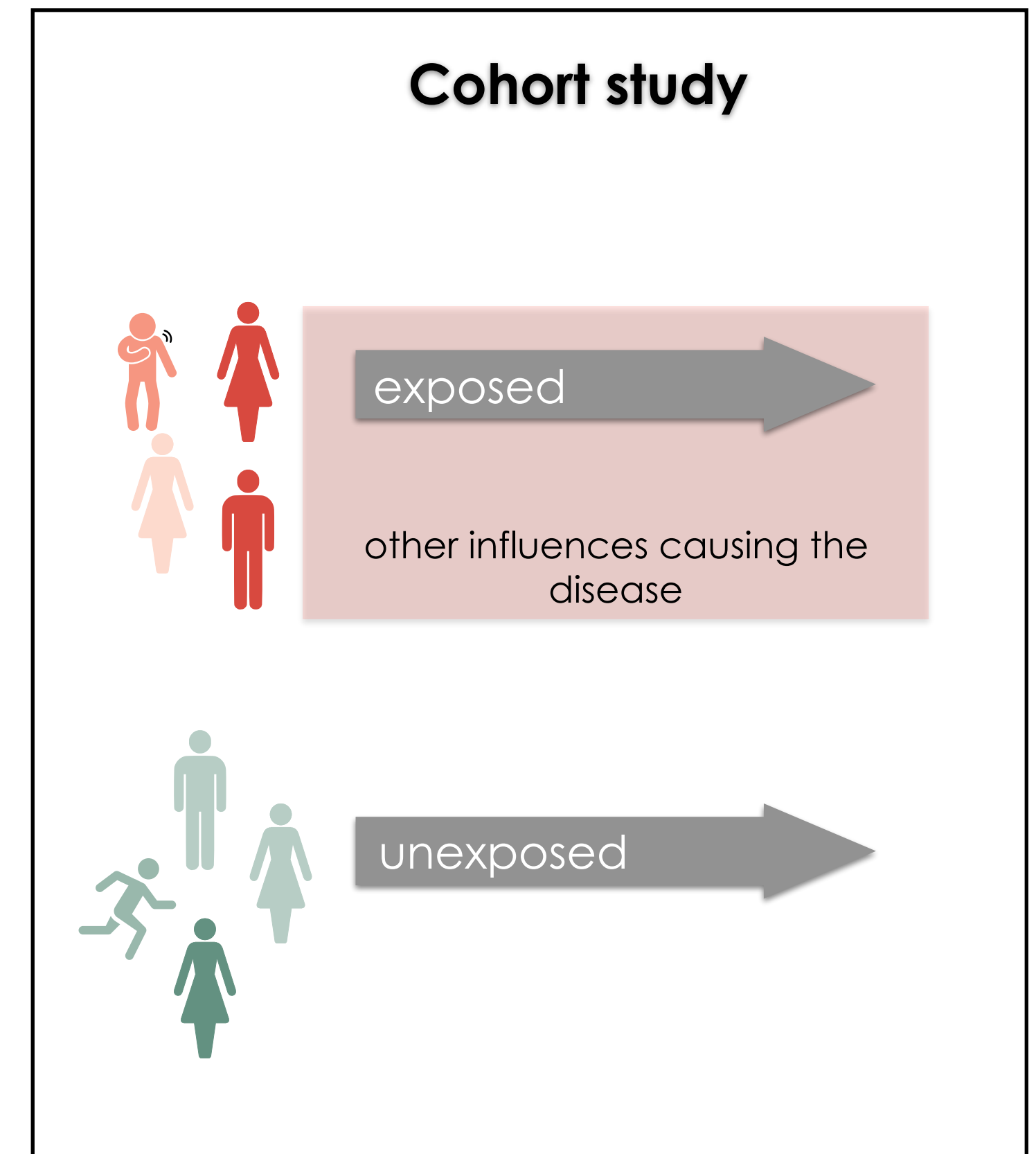
In order to proof that an effect is causal, the two populations (i.e. exposed and unexposed) need to be **EXCHANGEABLE**



We will let you know immediately when we figured this out, should be soon....



TODAY in EPI 203



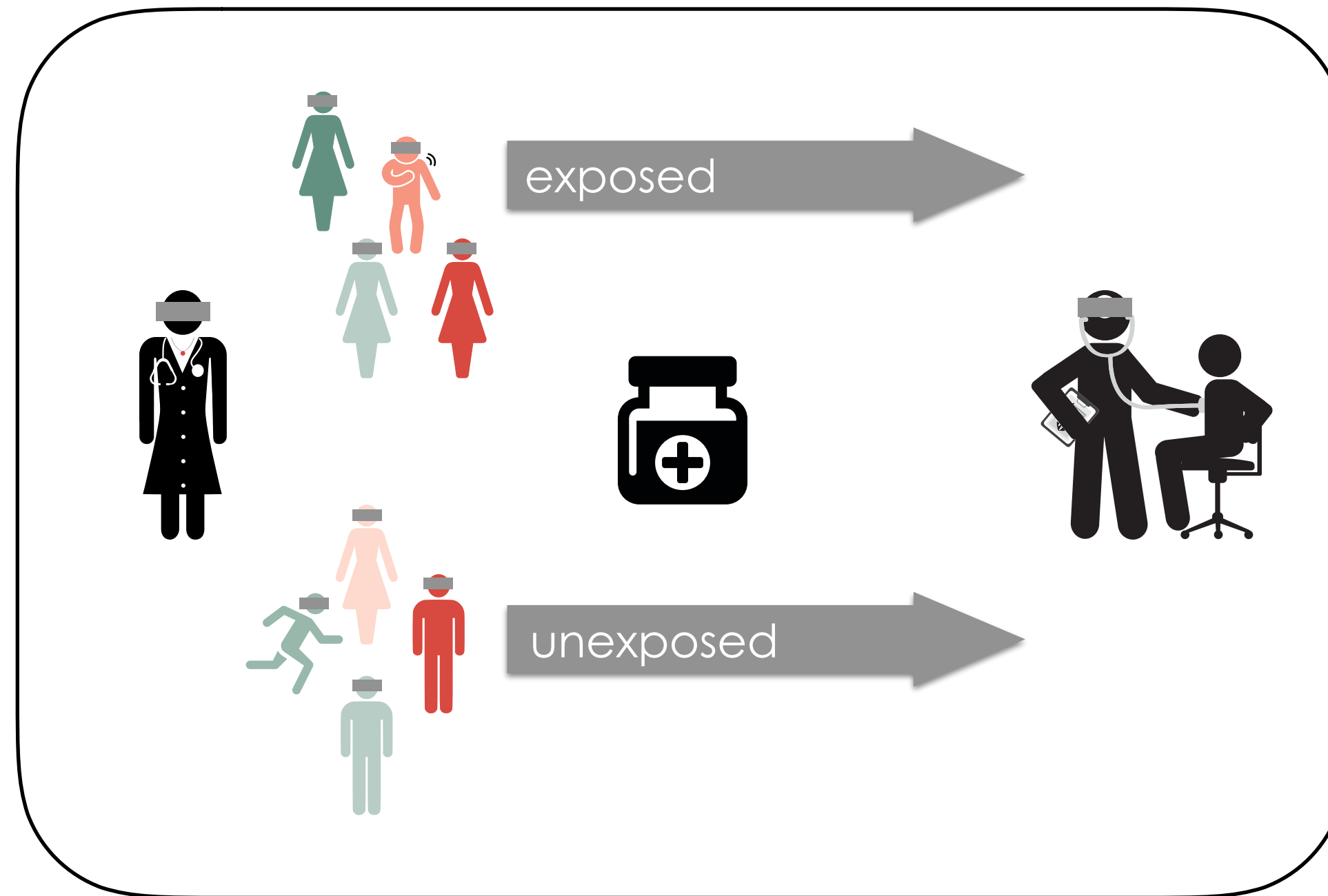
In order to get at causal interference in cohort studies, you have to deal with confounding, you will learn this in EPI 203

Issues with RCTs?

Subjects



Treatment variable

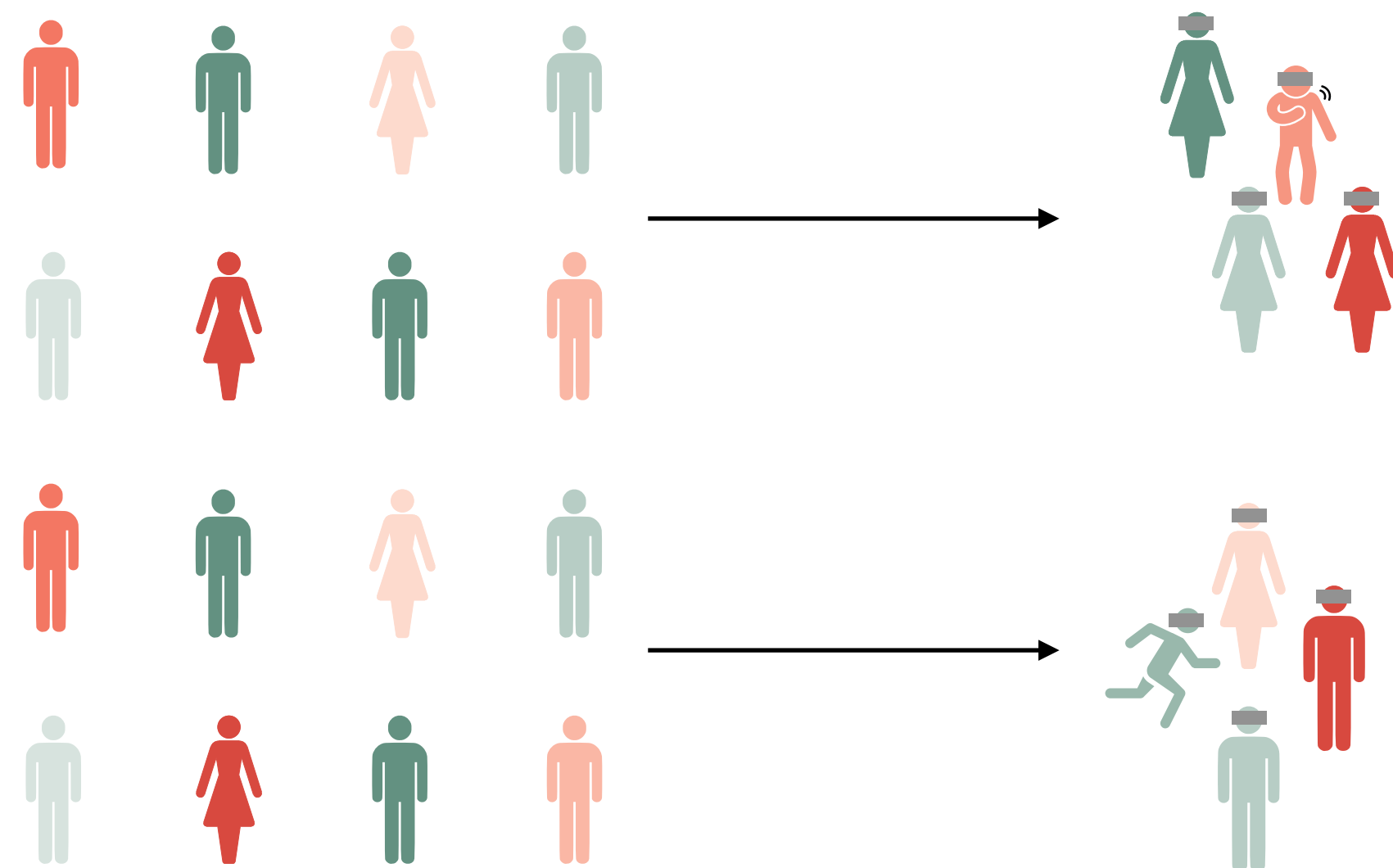


Outcome variable

Perfect randomization, perfect blinding

Issues with RCTs?

Subjects



Is the study population representative of the population the results of the study will be applied to?

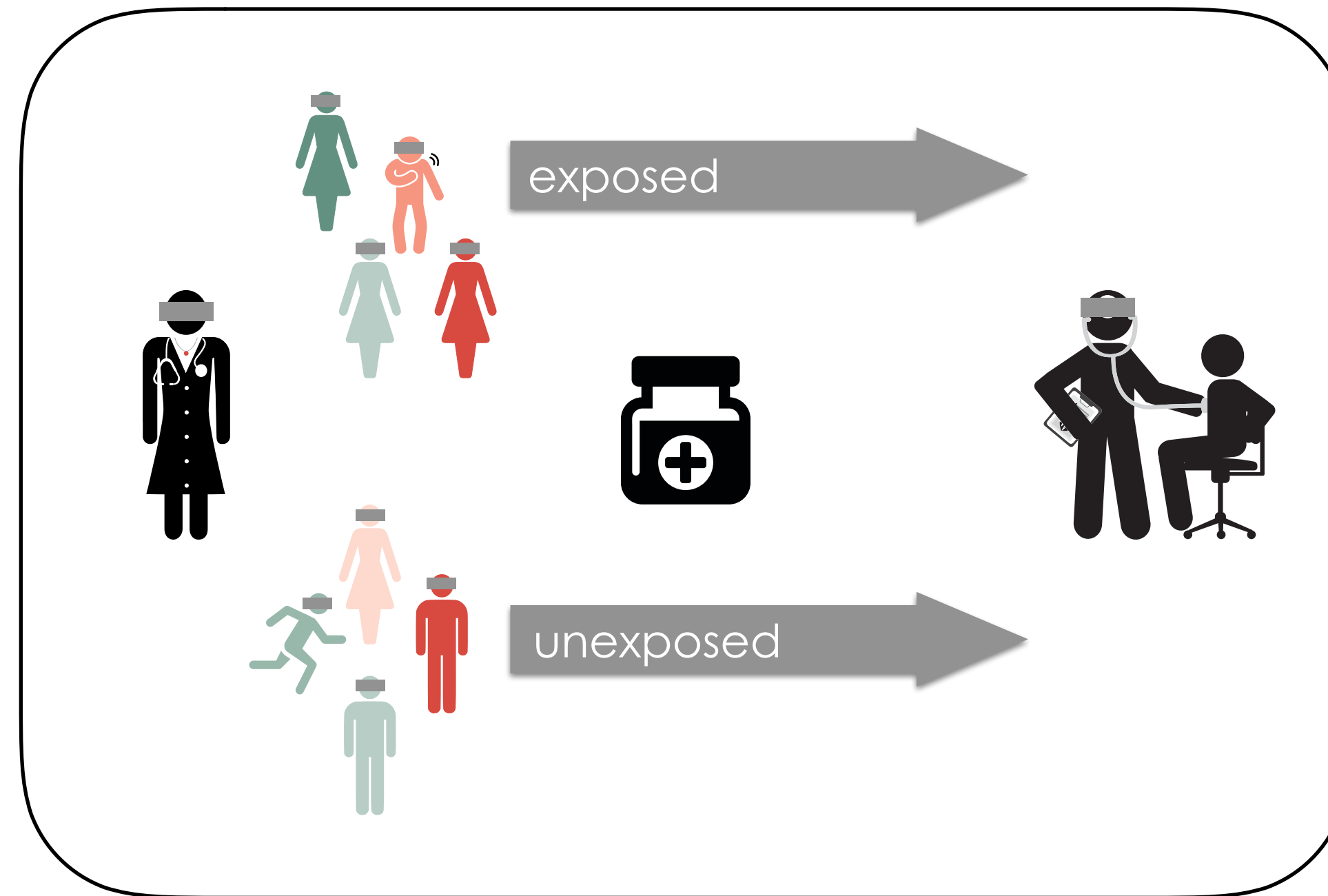
- Select patient most likely to benefit from treatment and least likely to have adverse effects
- Groups under-represented
 - Children
 - Elderly
 - Women
 - Comorbid conditions
 - On other meds

Issues with RCTs?

Subjects

Treatment variable

Outcome variable



Perfect randomization, perfect blinding

Issues with RCTs?

Treatment variable



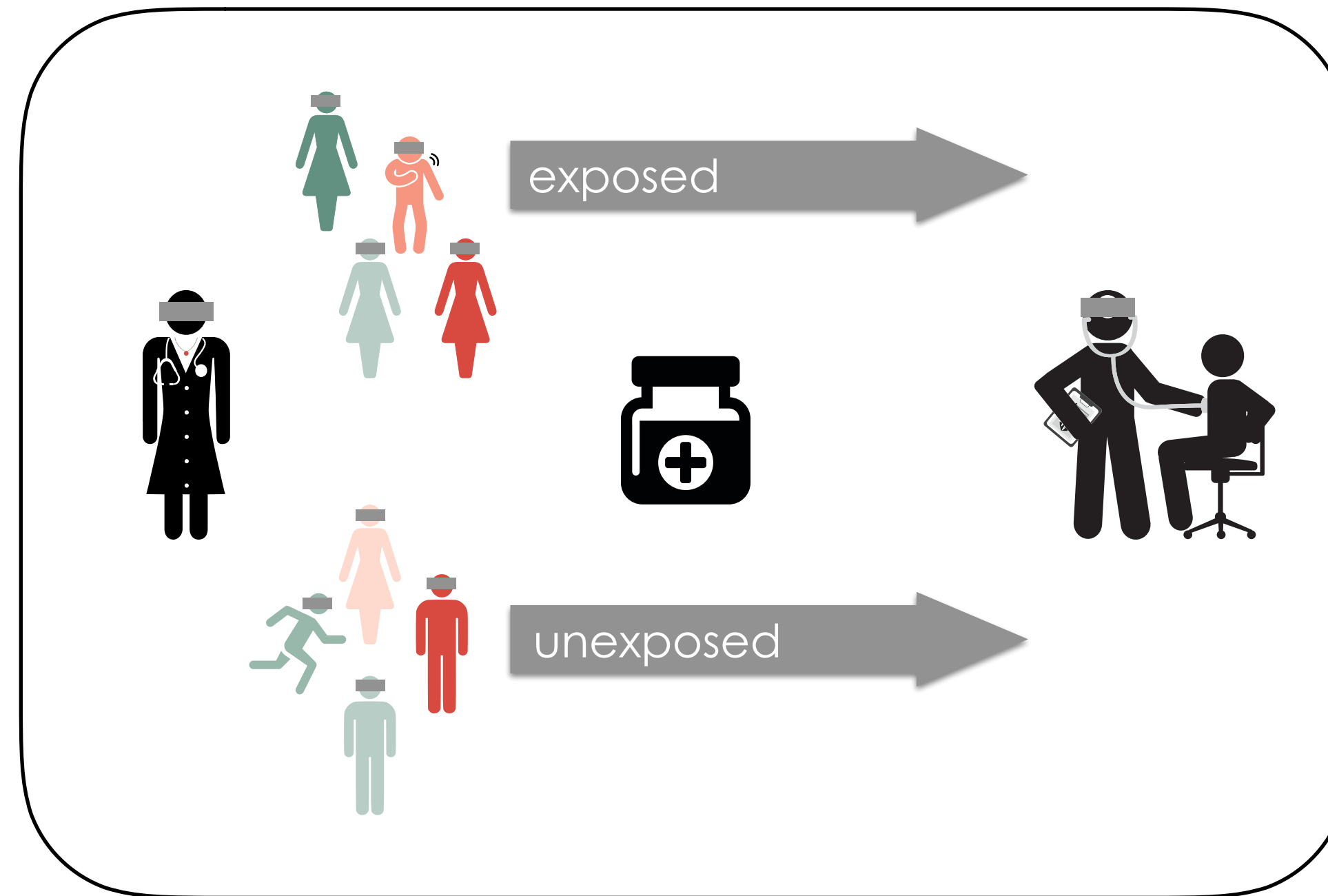
Study drug
Dose
Comparator

Issues with RCTs?

Subjects

Treatment variable

Outcome variable



Perfect randomization, perfect blinding

Issues with RCTs?

- **Surrogate outcomes**

- ✓ Presumed to be associated with the outcome of interest.
- ✓ more easily measured
- ✓ occurs sooner

- **Composite outcomes**

- ✓ Several bad outcomes grouped into one composite outcome
- ✓ Can lead to combining outcomes of varying importance.
Death vs. multiple hospital re-admissions
- ✓ Can be misleading if individual components of composite outcomes are not reported

Outcome variable



Issues with RCTs?

Surrogate outcomes: examples

- Outcome is risk factor rather than disease:
 - ✓ Blood pressure instead of stroke
- Outcome is marker of disease severity rather than morbidity/mortality from disease
 - ✓ Viral load instead of morbidity from AIDS
 - ✓ HbA1C rather than morbidity from diabetes

Issues with RCTs?

Surrogate outcome: The CAST trial



The NEW ENGLAND
JOURNAL of MEDICINE

SPECIAL REPORT [FREE PREVIEW](#) [ARCHIVE](#)

Preliminary Report: Effect of Encainide and Flecainide on Mortality in a Randomized Trial of Arrhythmia Suppression after Myocardial Infarction

The Cardiac Arrhythmia Suppression Trial (CAST) Investigators

- Asymptomatic premature ventricular contractions after myocardial infarction increase risk of sudden death
- Surrogate outcome: PVC burden

Issues with RCTs?

Composite outcome: The ACCORD trial



The NEW ENGLAND
JOURNAL of MEDICINE

ORIGINAL ARTICLE

Effects of Intensive Blood-Pressure Control in Type 2 Diabetes Mellitus

The ACCORD Study Group*

Article Figures/Media

[17 References](#) [1874 Citing Articles](#)

[April 29, 2010](#)

N Engl J Med 2010; 362:1575-1585

DOI: 10.1056/NEJMoa1001286

Pre-specified primary composite outcome

Nonfatal MI

Nonfatal stroke

Cardiovascular death

Issues with RCTs?

Composite outcome: The ACCORD trial

After 3.5 years...

Non-significant reduction in the risk of the primary composite outcome

But.....

1% decrease in the risk of nonfatal MI (3.6% vs. 4.6%; $p=0.004$)

1% increase in risk of cardiovascular death!! (5% vs. 4%, $p=0.04$)

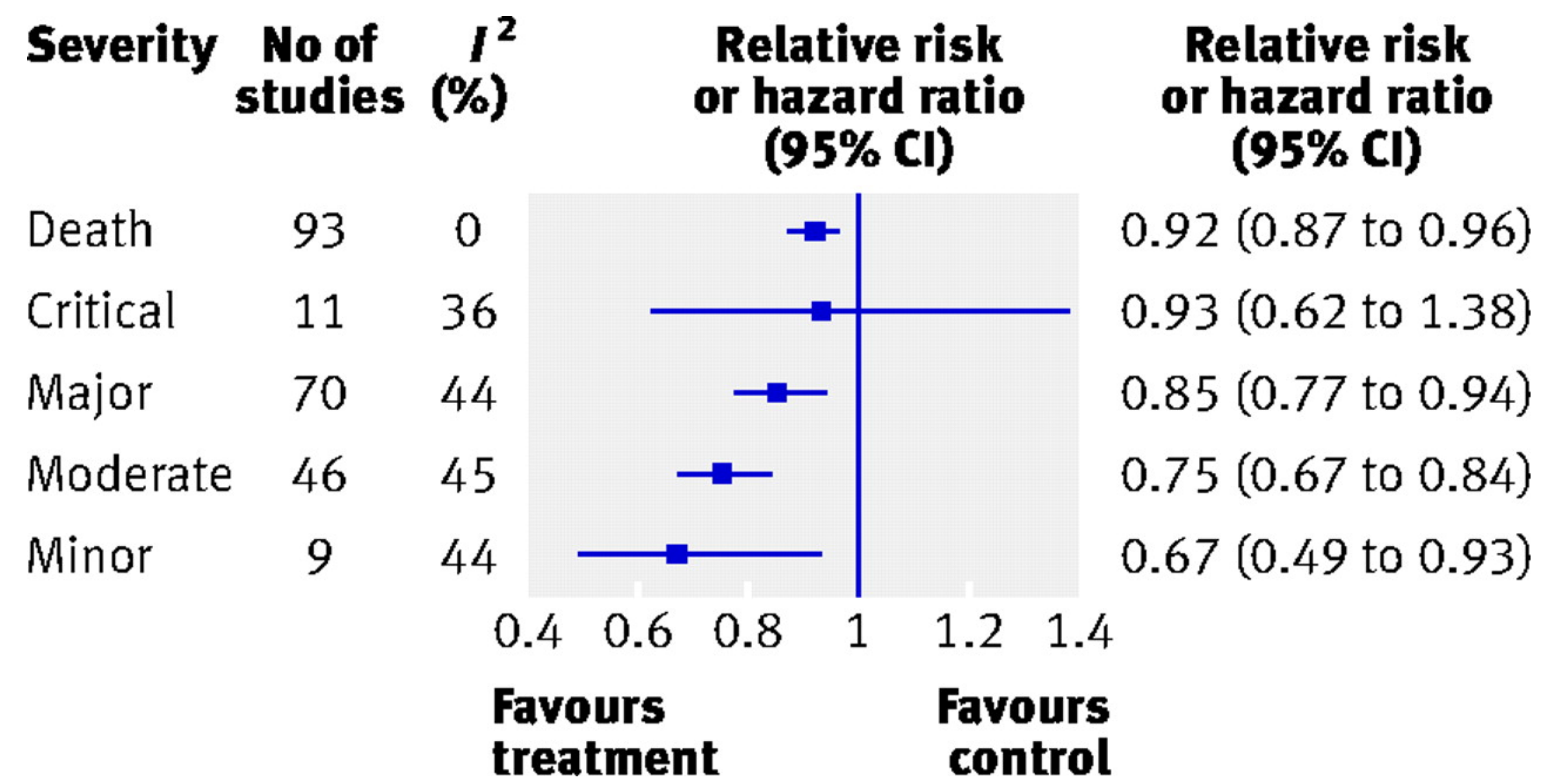
Masked an increase in total mortality!

Issues with RCTs?

Composite outcome: A common problem

I. Ferreira-González et al. Problems with use of composite end points in cardiovascular trials: systematic review of randomised controlled trials. BMJ 2007.

- Systematic Review of 114 randomized trials of cardiovascular interventions
- 68% of the studies reported results for each component of the composite outcome
- Outcomes of greater importance such as death were associated with smaller relative treatment effects than less important outcomes



Issues with RCTs?

Composite outcome

- Important to 'decompose'
- Can mask important effects of therapy
 - ✓ No effect on important outcomes
 - ✓ Negative effect on important outcomes

Issues with RCTs?

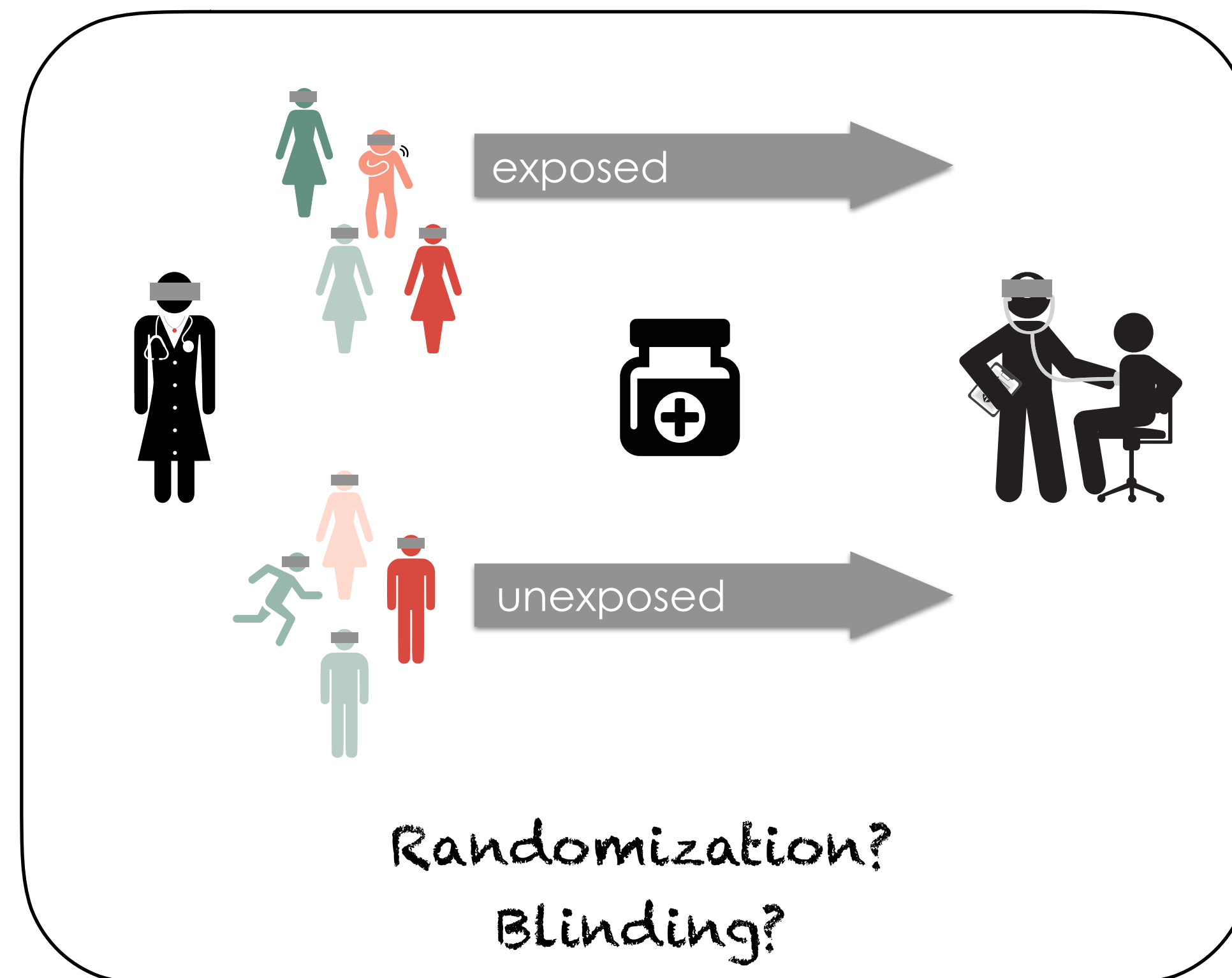
Treatment variable

Study drug?
Dose?
Comparator?

Subjects

Subjects representative
of target population?

Underrepresented
groups?



Outcome variable

Surrogate outcome?

Composite outcome?

Analysis of RCTs

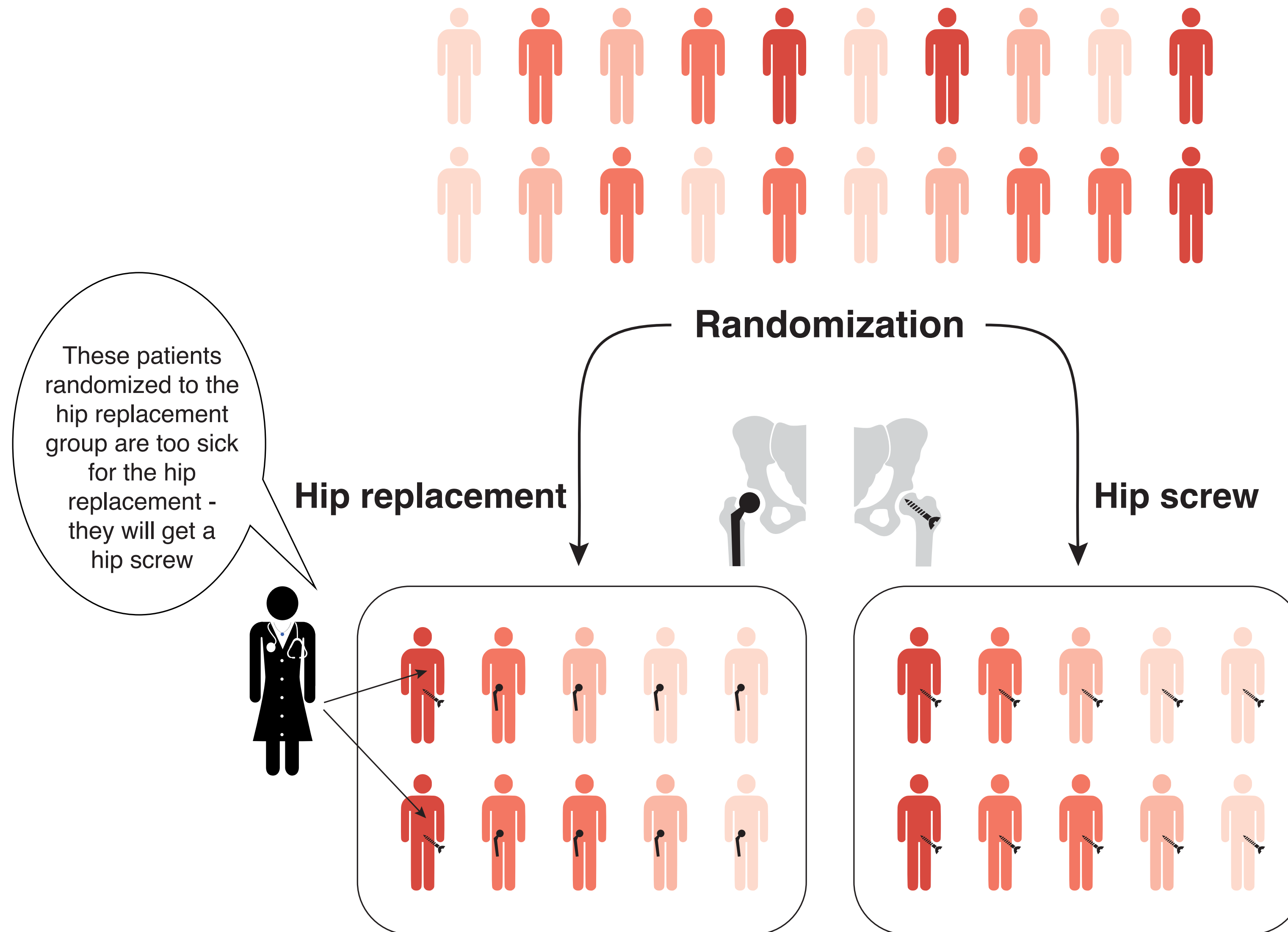
Problem: Subjects don't always stay in the assigned group

Intention to treat

As treated

Per protocol

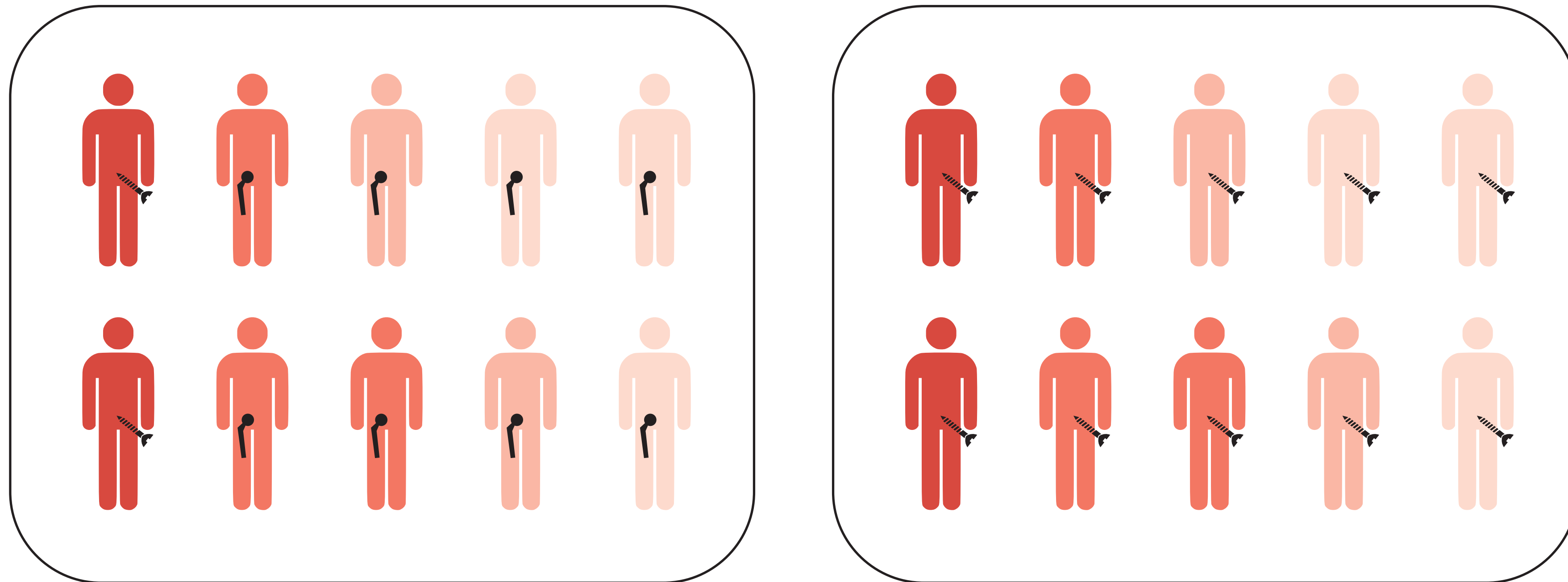
Analysis of RCTs



Analysis of RCTs

Intention to treat

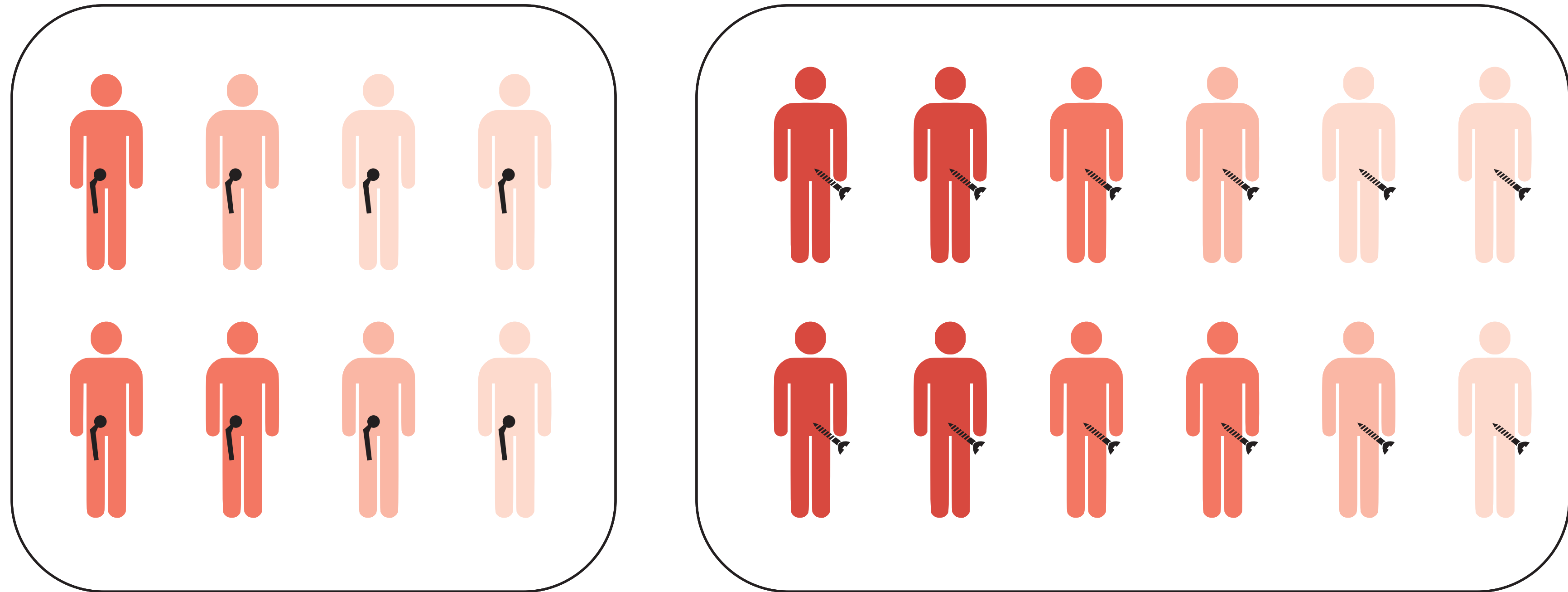
All subjects are analyzed according to the randomization group



Analysis of RCTs

As treated

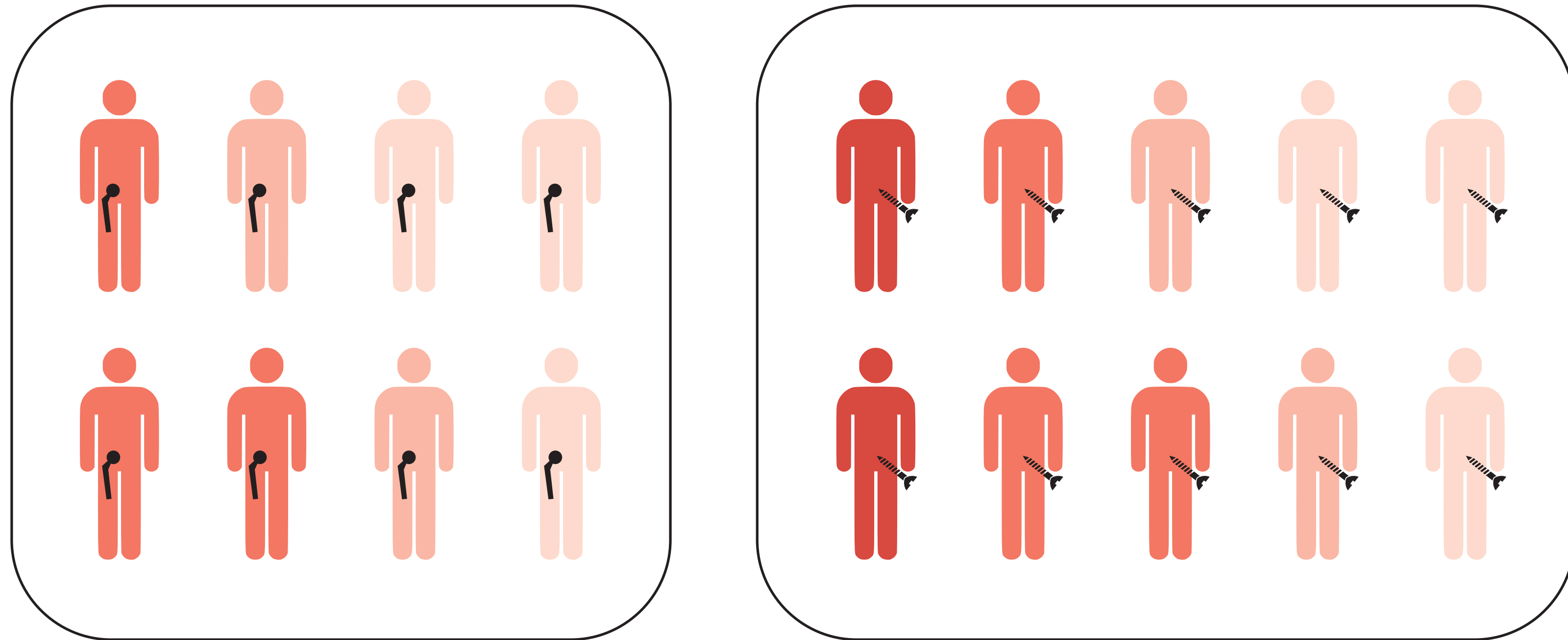
Subjects are analyzed according to treatment received



Analysis of RCTs

Per protocol

Only subjects treated according to the study protocol are analyzed

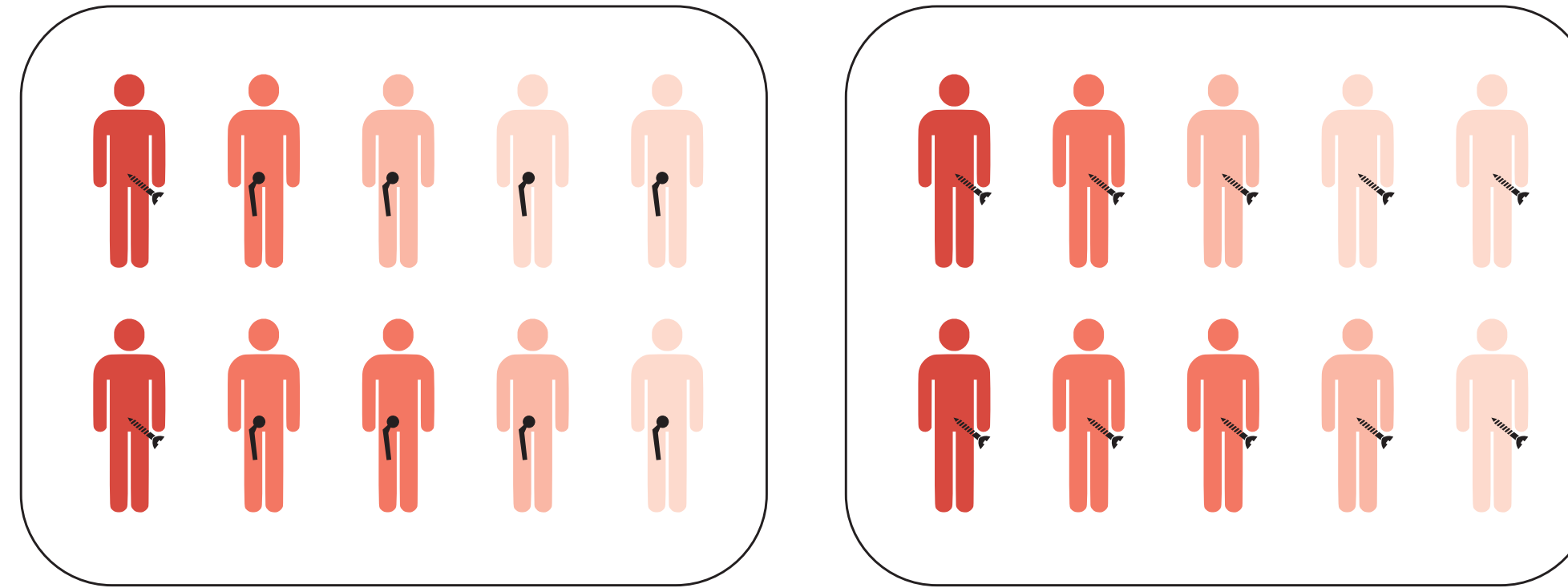


Analysis of RCTs

A

Intention to treat

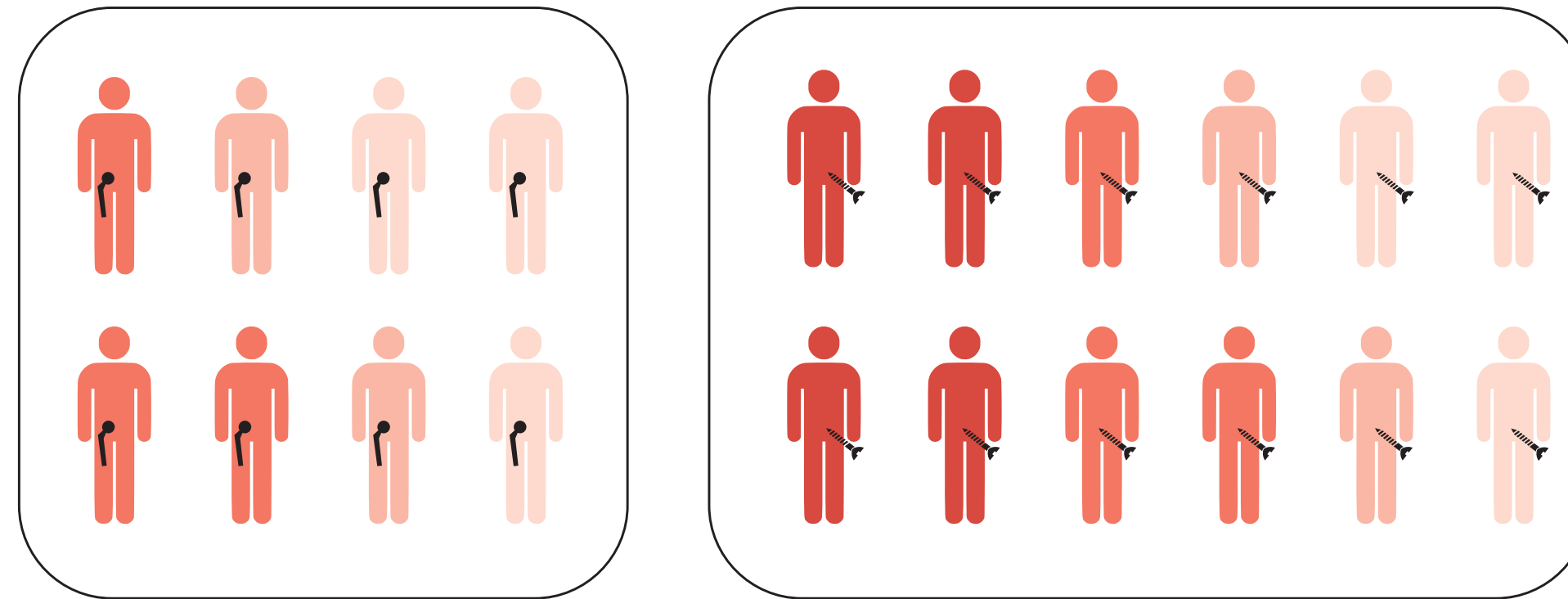
all subjects are analyzed according to the randomization group



B

As treated

Subjects are analyzed according to treatment received

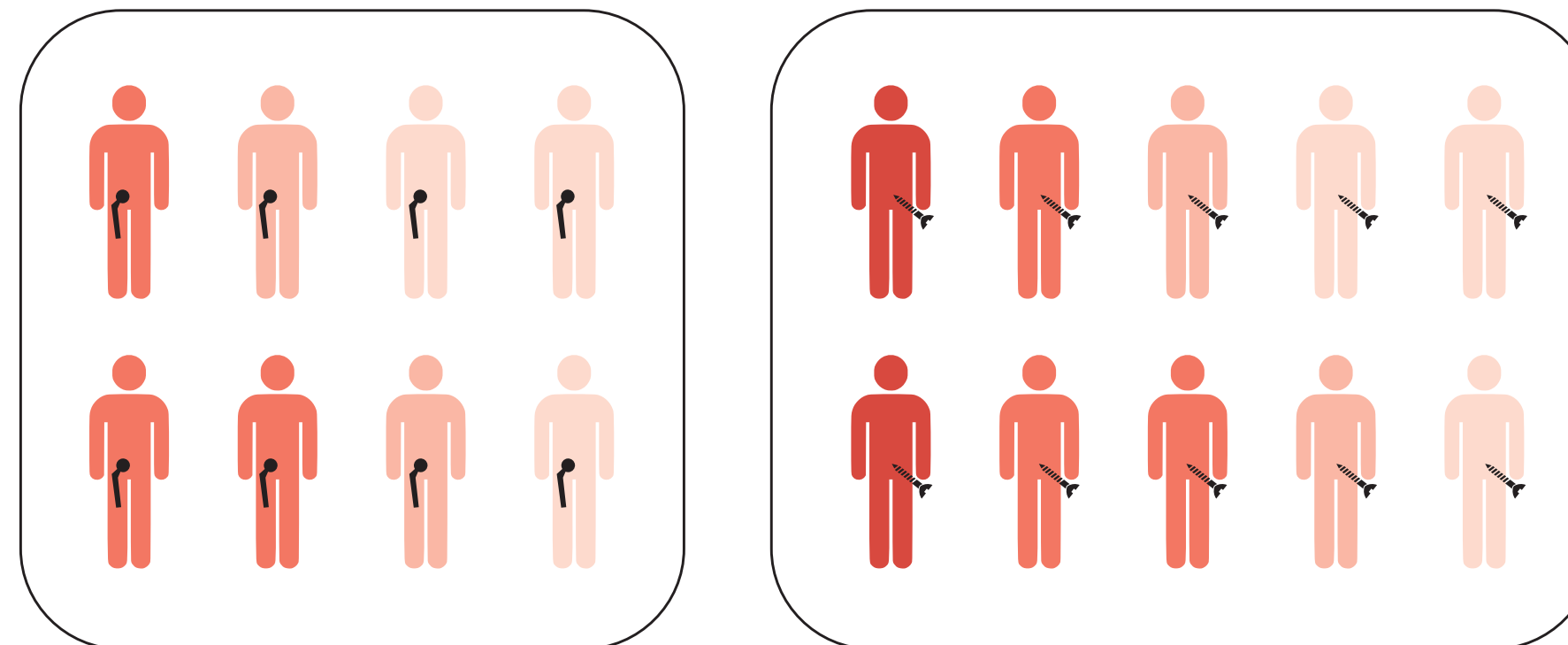


Bias in favor of hip replacement

C

Per protocol

Only subjects treated according to the study protocol are analyzed



Bias in favor of hip replacement

Analysis of RCTs: Pitfalls

- Sub-group analysis
- Multiple Comparisons
- Between- vs. within-group comparisons

Analysis of RCTs: Pitfalls

Sub-group analysis

Presenting results within a particular subgroup of subjects (bound to find something interesting)

THE LANCET
Volume 332, Issue 8607, 13 August 1988, Pages 349-360



RANDOMISED TRIAL OF INTRAVENOUS
STREPTOKINASE, ORAL ASPIRIN, BOTH, OR
NEITHER AMONG 17 187 CASES OF
SUSPECTED ACUTE MYOCARDIAL
INFARCTION: ISIS-2

Aspirin therapy (one month of 160 mg/day) in patients with acute myocardial infarction (AMI) reduced 30-day cardiovascular mortality from 11.8% in the placebo group to 9.4% in the aspirin group.

Analysis of RCTs: Pitfalls

Sub-group analysis

Presenting results within a particular subgroup of subjects (bound to find something interesting)

ISIS-2	30-day mortality		
	Overall	Geminis and Libras	Other signs
Aspirin	9.4%	11.1%	9.0%
Placebo	11.8%	10.2%	12.1%

Analysis of RCTs: Pitfalls

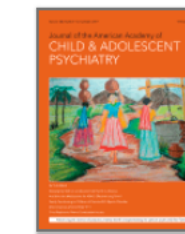
Multiple comparison

Looking at multiple different outcomes in multiple different groups in multiple different ways (need to adjust for this)



Journal of the American Academy of Child & Adolescent Psychiatry

Volume 40, Issue 7, July 2001, Pages 762-772



Articles

Efficacy of Paroxetine in the Treatment of Adolescent Major Depression: A Randomized, Controlled Trial

MARTIN B. KELLER M.D. ¹, NEAL D. RYAN M.D., MICHAEL STROBER PH.D., RACHEL G. KLEIN PH.D., STAN P. KUTCHER M.D., BORIS BIRMAHER M.D., OWEN R. HAGINO M.D., HAROLD KOPLEWICZ M.D., GABRIELLE A. CARLSON M.D., GREGORY N. CLARKE PH.D., GRAHAM J. EMSLIE M.D., DAVID FEINBERG M.D., BARBARA GELLER M.D., VIVEK KUSUMAKAR M.D., GEORGE PAPTAEODOROU M.D., WILLIAM H. SACK M.D., MICHAEL SWEENEY PH.D., KAREN DINEEN WAGNER M.D., PH.D. ... JAMES P. MCCAFFERTY B.S.

Glaxo-Smith Kline funded study on depression in adolescents concluded that Paroxetine "generally well-tolerated and effective for major depression in adolescents"

Analysis of RCTs: Pitfalls

Multiple comparison

Looking at multiple different outcomes in multiple different groups in multiple different ways (need to adjust for this)

- Independent analysis: 20 outcome measures had been added to the 8 original outcomes
- Only highlighted those that were favorable!
- GSK paid a fine of \$3 billion (which was a fraction of the sales from this drug– \$27.9 billion)

Analysis of RCTs: Pitfalls

Within group comparison

Comparing multiple time points within the treatment group

JAMA Network

November 5, 2003

Effect of Recombinant ApoA-I Milano on Coronary Atherosclerosis in Patients With Acute Coronary Syndromes A Randomized Controlled Trial

Steven E. Nissen, MD; Taro Tsunoda, MD; E. Murat Tuzcu, MD; [et al](#)

» [Author Affiliations](#)

JAMA. 2003;290(17):2292-2300. doi:10.1001/jama.290.17.2292

Analysis of RCTs: Pitfalls

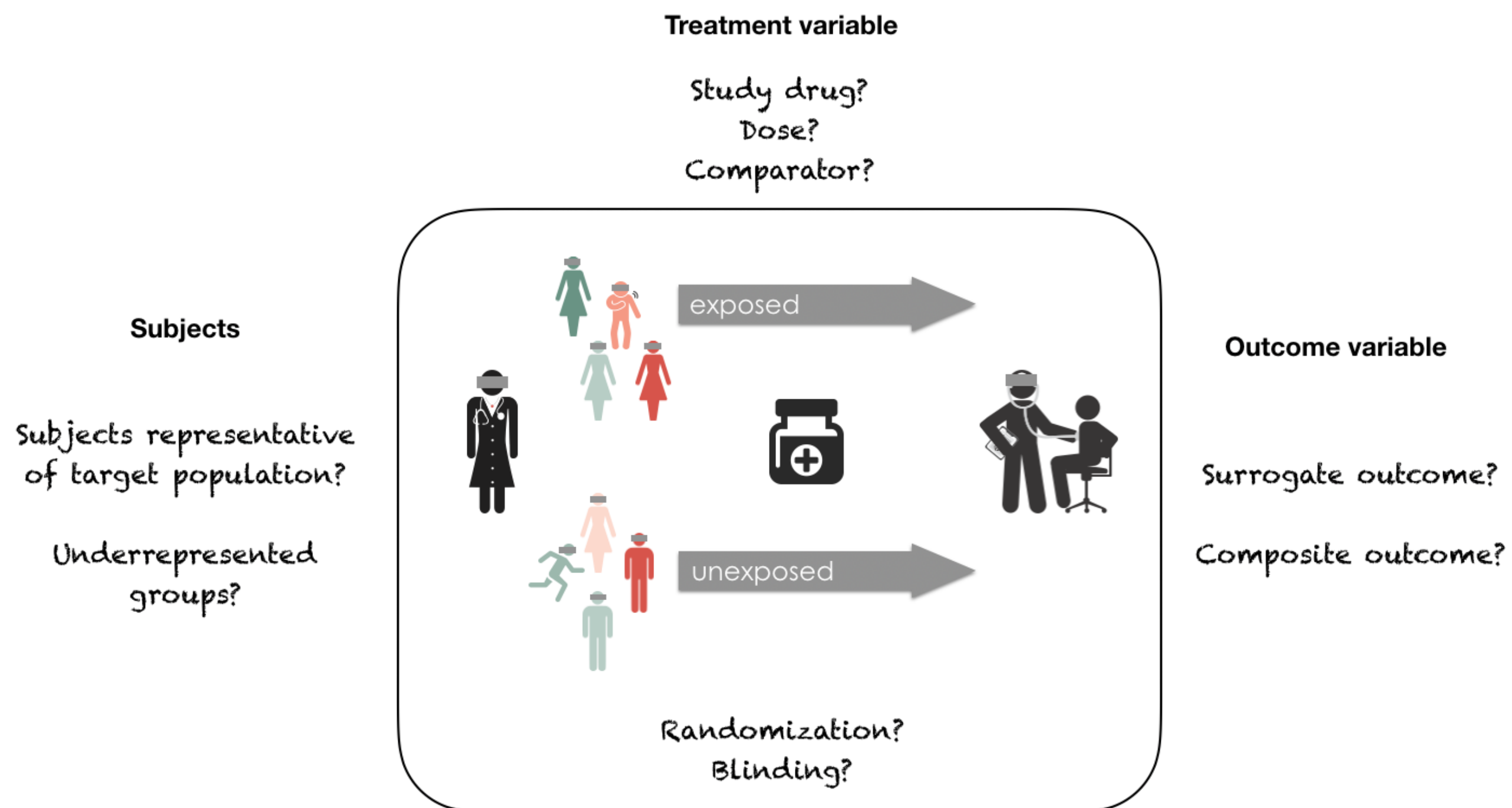
Within group comparison

Comparing multiple time points within the treatment group

Percent Atheroma Volume in the Target Coronary Segment					
	N	Baseline	Follow-Up	Change	P Value*
Placebo	11	34.8	34.9	0.1	0.97
ApoA-I Milano					
15 mg/kg	21	39.7	38.4	-1.3	0.03
45 mg/kg	15	37.9	37.2	-0.7	0.45
Combined	36	39.0	37.9	-1.1	0.02
*P values for within-group comparison from Wilcoxon signed rank test. For between-group comparison, P = 0.29					

Issues with RCTs?

Design and Execution



Analysis



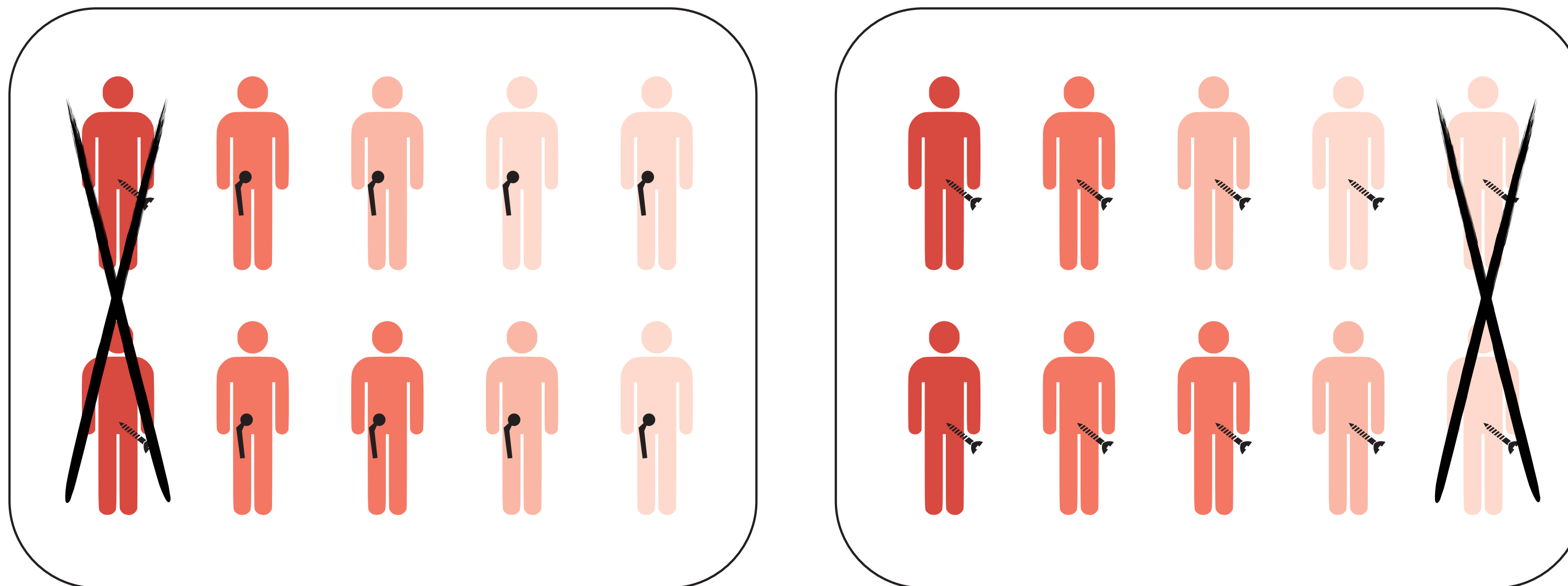
Subgroup analysis?

Multiple comparison?

Within group comparison?

Loss to follow-up in RCTs

- Reduces power
- May introduce bias



Loss to follow-up in RCTs

Solution:

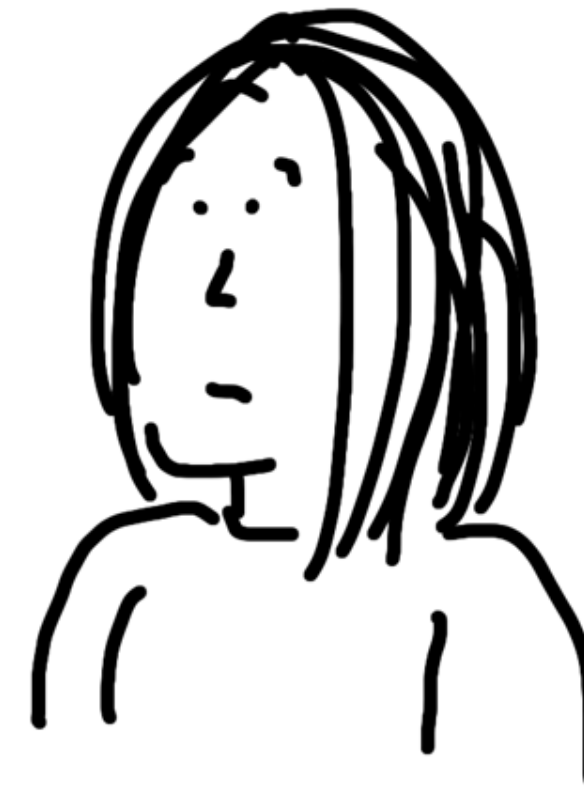
1. Sensitivity Analysis

Worst case scenario analysis (extreme): Poor outcomes for all losses in Tx group & good outcomes for all losses in control group

2. Compare subjects that remained in the study vs. lost to follow up (similar or different?)

Break

Do you know about any RCTs that provide evidence that we should use RCTs?



Quantifying Treatment Effects

	Poor outcome	No poor outcome
Treatment	A	B
Control	C	D

$R_T = \text{Risk in Treatment Group} = a/(a+b)$

$R_C = \text{Risk in Control Group} = c/(c+d)$

$RR = R_T/R_C = a/(a+b)/c/(c+d)$

$RRR = 1 - RR$

$ARR = - \text{Risk Difference} = - (R_T - R_C) = R_C - R_T = c/(c+d) - a/(a+b)$

$NNT = 1/ARR$

$OR = ad/bc$

Quantifying Treatment Effects

ORIGINAL ARTICLE

Serious Asthma Events with Fluticasone plus Salmeterol versus Fluticasone Alone

David A. Stempel, M.D., Ibrahim H. Raphiou, Ph.D., Kenneth M. Kral, M.S., Anne M. Yeakey, M.D., Amanda H. Emmett, M.S., Charlene M. Prazma, Ph.D., Kathleen S. Buaron, B.S.N., and Steven J. Pascoe, M.B., B.S. for the AUSTRI Investigators*

Article	Figures/Media	Metrics	May 12, 2016
28 References	80 Citing Articles		N Engl J Med 2016; 374:1822-1830
	Letters		DOI: 10.1056/NEJMoa1511049

	Astma exacerbation	No Asthma exacerbation	Total
Fluticasone + Salmeterol	480	5354	5834
Fluticasone alone	597	5248	5845

$R_T = \text{Risk (Fluticasone+Salmeterol)} = 480/5834 = 8.2\%$
 $R_c = \text{Risk (Fluticasone alone)} = c/(c+d) = 597/5845 = 10.2\%$

$ARR = - \text{Risk Difference} = - (R_T - R_c) = -(8.2\%-10.2\%) = 2.0\%$
(over 6 months)

$RR = R_t/R_c = a/(a+b)/c/(c+b) = 8.2\%/10.2\% = 0.81$

$NNT = 1/ARR = 1/0.02 = 50$ for 6 months

$RRR = 1 - RR = 1 - 0.81 = 19\%$

Meaning of NNT?

Quantifying Treatment Effects

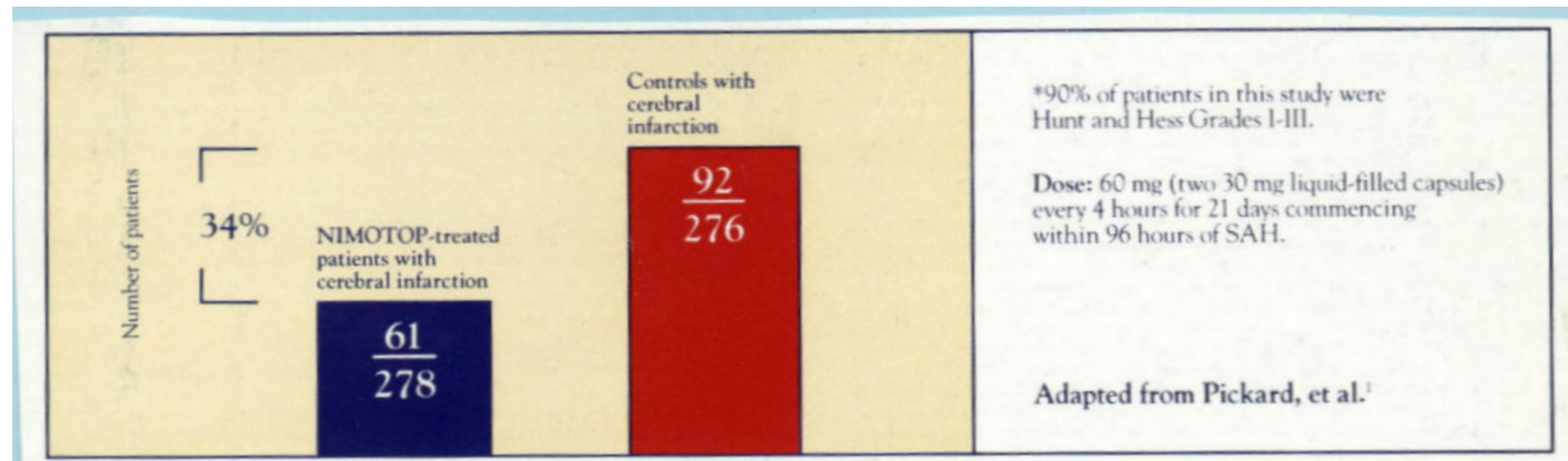
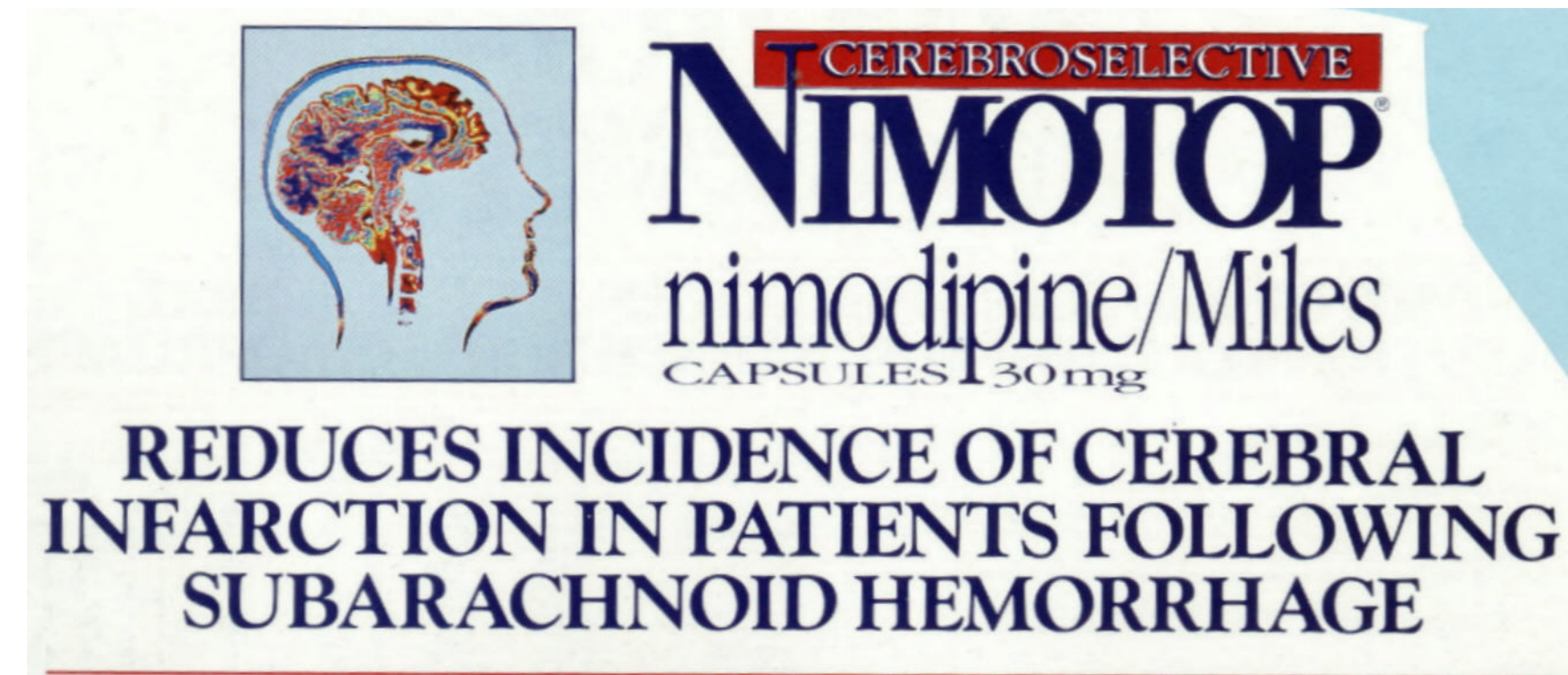
Number Needed to Treat (NNT)

- Number needed to treat to prevent one bad outcome
- Lower the # the better
- Specify the follow-up time, the bad outcome being prevented and the characteristics of the people being treated
- Round to the nearest integer (since talking about people) especially when talking to patients

.

Quantifying Treatment Effects

Relative and Absolute Risk Reduction



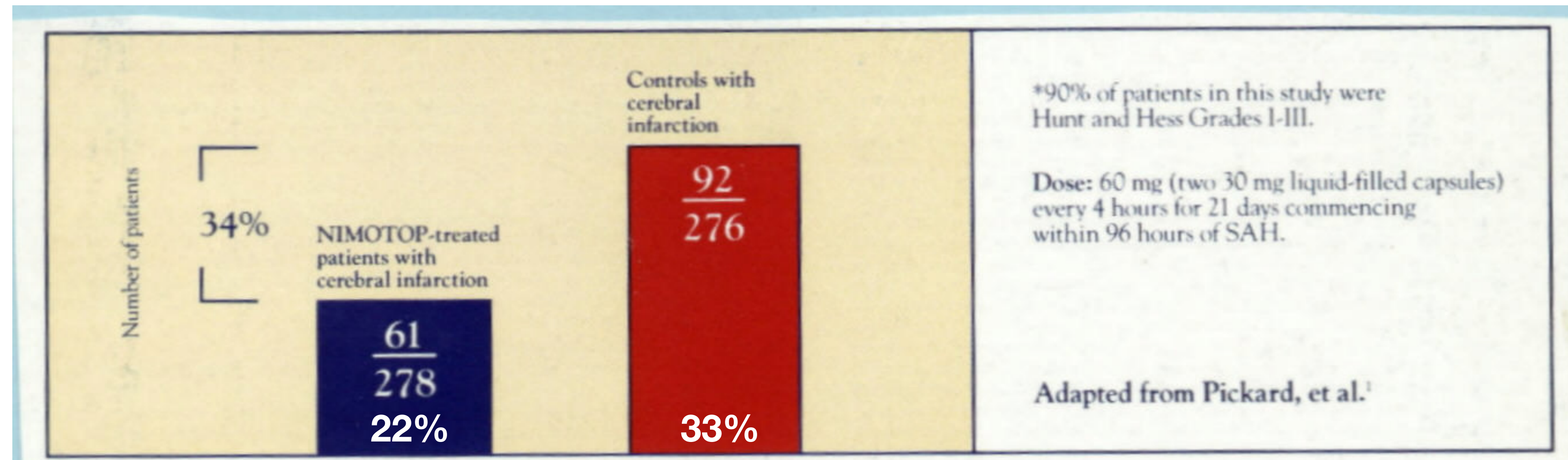
Quantifying Treatment Effects

Relative and Absolute Risk Reduction

$$RR = 22\% / 33\% = 0.66$$

$$RRR = 1 - 0.66 = 34\%$$

$$ARR = 33\% - 22\% = 11\%$$



Quantifying Treatment Effects

Relative and Absolute Risk Reduction

- Relative risk reduction will always appear more impressive than absolute risk reduction
- Absolute risk reduction takes into account baseline risk (of control group)

Quantifying Treatment Effects

Relative and Absolute Risk Reduction

ORIGINAL ARTICLE

Intravascular Gamma Radiation for In-Stent Restenosis in Saphenous-Vein Bypass Grafts

Ron Waksman, M.D., Andrew E. Ajani, M.D., R. Larry White, M.D., Rosanna C. Chan, M.D., Lowell F. Satler, M.D., Kenneth M. Kent, M.D., Augusto D. Pichard, M.D., Ellen E. Pinnow, M.S., Anh B. Bui, M.D., Steven Ramee, M.D., Paul Teirstein, M.D., and Joseph Lindsay, M.D.

[Article](#) [Figures/Media](#)

April 18, 2002

N Engl J Med 2002; 346:1194-1199

DOI: 10.1056/NEJMoa012579

[19 References](#) [100 Citing Articles](#) [Letters](#)

“At 12 months, the rate of revascularization of the target lesion was 70 percent lower in the iridium-192 group than in the placebo group”.

After cleaning out blocked bypass graft, how many do we have to treat with iridium to prevent one revascularization procedure (of the target lesion)?

You need to calculate NNT

Quantifying Treatment Effects

Number Needed to Treat

$$RRR = 1 - (R_t / R_c)$$

$$RRR \times R_c = 1 \times R_c - 1 (R_t / R_c) \times R_c$$

$$RRR \times R_c = R_c - R_t$$

$$ARR = RRR \times Risk_{control}$$

$$NNT = 1 / ARR$$



Quantifying Treatment Effects

Relative and Absolute Risk Reduction

ORIGINAL ARTICLE

Intravascular Gamma Radiation for In-Stent Restenosis in Saphenous-Vein Bypass Grafts

Ron Waksman, M.D., Andrew E. Ajani, M.D., R. Larry White, M.D., Rosanna C. Chan, M.D., Lowell F. Satler, M.D., Kenneth M. Kent, M.D., Augusto D. Pichard, M.D., Ellen E. Pinnow, M.S., Anh B. Bui, M.D., Steven Ramee, M.D., Paul Teirstein, M.D., and Joseph Lindsay, M.D.

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Quantifying Treatment Effects

Relative and Absolute Risk Reduction

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[Article](#) [Figures/Media](#)

[19 References](#) [100 Citing Articles](#) [Letters](#)

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N Engl J Med 2002; 346:1194-1199

DOI: 10.1056/NEJMoa012579

“At 12 months, the rate of revascularization of the target lesion was 70 percent lower in the iridium-192 group than in the placebo group”.

The **baseline risk** (i.e., risk in the placebo group) was **57%**.

After cleaning out blocked bypass graft, how many do we have to treat with iridium to prevent one revascularization procedure (of the target lesion)?

Quantifying Treatment Effects

Relative and Absolute Risk Reduction

ORIGINAL ARTICLE

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Ron Waksman, M.D., Andrew E. Ajani, M.D., R. Larry White, M.D., Rosanna C. Chan, M.D., Lowell F. Satler, M.D., Kenneth M. Kent, M.D., Augusto D. Pichard, M.D., Ellen E. Pinnow, M.S., Anh B. Bui, M.D., Steven Ramee, M.D., Paul Teirstein, M.D., and Joseph Lindsay, M.D.

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Baseline 12-month risk of revascularization = 57%

RRR = 0.70

$ARR = RRR \times Risk_{control}$

$ARR = 0.70 \times 0.57 = 0.40$

$NNT = 1/0.40 = 2.5$

Need to treat 2.5 unclogged grafts with radiation to prevent 1 from needing revascularization of the target lesion in the next 12 months.

Quantifying Treatment Effects

Odds Ratio

- OR not necessary to report in clinical trials
- Often reported since always further from 1 than the RR
- Inflates the apparent effect size
-

ORIGINAL ARTICLE

Intravascular Gamma Radiation for In-Stent Restenosis in Saphenous-Vein Bypass Grafts

Ron Waksman, M.D., Andrew E. Ajani, M.D., R. Larry White, M.D., Rosanna C. Chan, M.D., Lowell F. Satler, M.D., Kenneth M. Kent, M.D., Augusto D. Pichard, M.D., Ellen E. Pinnow, M.S., Anh B. Bui, M.D., Steven Ramee, M.D., Paul Teirstein, M.D., and Joseph Lindsay, M.D.

	Major Cardiac Event	No Major Cardiac Event	Total
Irradiation	19	41	60
Placebo	38	22	60

$$\text{RR} = (19/60)/(38/60) = 0.50$$

$$\text{OR} = (19/41)/(38/22) = 0.27$$

Quantifying Treatment Effects

Incorporating Cost

NNT= # needed to treat to prevent one bad outcome

C= cost to treat one patient

NNT x C= treatment cost to prevent one bad outcome = CBOP

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Quantifying Treatment Effects

CBOP



<https://www.shopcobop.com>



<https://cbop.audubon.org>

Quantifying Treatment Effects

JAMA Network

February 14, 2001

Effectiveness of Oseltamivir in Preventing Influenza in Household Contacts A Randomized Controlled Trial

Robert Welliver, MD; Arnold S. Monto, MD; Otmar Carewicz, MD; [et al](#)

Index case Proven Flu +

	Flu + Contact	Flu – Contact	Totals
Tamiflu	3	206	209
Placebo	26	180	206

$$\text{Risk}_T = 3/209 = 1.4\%$$

$$\text{Risk}_C = 26/206 = 12.6\%$$

$$\text{RR} = 1.4\%/12.6\% = 11\%$$

$$\text{RRR} = 1 - 11\% = 89\%$$

$$\text{ARR} = 12.6\% - 1.4\% = 11.2\%$$

Why is ARR
different?

Index case NOT Proven Flu +

	Flu + Contact	Flu – Contact	Totals
Tamiflu	1	283	284
Placebo	8	248	256

$$\text{Risk}_T = 1/284 = 0.35\%$$

$$\text{Risk}_C = 8/256 = 3.1\%$$

$$\text{RR} = 0.35\%/3.1\% = 11\%$$

$$\text{RRR} = 1 - 11\% = 89\%$$

$$\text{ARR} = 3.1\% - 0.35\% = 2.7\%$$

Quantifying Treatment Effects

Incorporating Cost

Tamiflu prophylaxis \$40

Index case Proven Flu +

	Flu + Contact	Flu – Contact	Totals
Tamiflu	3	206	209
Placebo	26	180	206

$$\text{ARR} = 12.6\% - 1.4\% = 11.2\%$$

$$\text{NNT} = 1 / 11.2\% = 9$$

$$\text{Cost} = 9 * \$40 = \$360$$

It costs about \$360 to prevent a case of flu in household contacts

Index case NOT Proven Flu +

	Flu + Contact	Flu – Contact	Totals
Tamiflu	1	283	284
Placebo	8	248	256

$$\text{ARR} = 3.1\% - 0.35\% = 2.7\%$$

$$\text{NNT} = 1 / 2.7\% = 37$$

$$\text{Cost} = 37 * \$40$$

It costs about \$1480 to prevent a case of flu in household contacts

Quantifying Treatment Effects

Incorporating Cost

CBOP: Is it worth it?

INTRODUCING BBOP

INTRODUCING BBOP



HearthSong
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Bump with the stars—or feel like you are! The fun begins the minute kids step inside our colorful, confetti-filled Bright Lights BBOP (Buddy Bounce Outdoor Play) ball—and skyrockets when they start bumping their chums (nanoseconds later). The more kids run and bump, the more the balls' motion-activated LEDs and colorful foil confetti light up like starbursts! Keeps kids on the move and laughter on a roll (available in Blue and Red). Use our Electric Pump for quick, easy inflation and deflation. Adult supervision recommended.

Quantifying Treatment Effects

Incorporating Cost

CBOP: Is it worth it?

Depends on BBOP= Benefit per Bad Outcome Prevented

BBOP= \$1080

BBOP > CBOP = treat!

Quantifying Treatment Effects

Incorporating Cost

Treatment Threshold

If $BBOP > CBOP$, at what probability of disease would benefits outweigh costs?

What is the treatment threshold (P_{TT})?

$$P_{TT} = C / (C + B)$$

We know C (\$40), but what is B ?

Quantifying Treatment Effects

Incorporating Cost

What is B?

B = expected benefit of treatment

$$B = \text{BBOP} \times \text{ARR} - C$$

**Benefit of
preventing a
bad outcome**

**Absolute Risk
Reduction (how
much does
treatment reduce
the risk)**

Cost

Quantifying Treatment Effects

Incorporating Cost

Treatment Threshold

$$B = BBOP \times ARR - C$$

$$ARR = 1/NNT$$

$$B = BBOP/NNT - C$$

$$P_{TT} = C / (C + BBOP/NNT - C) = C / (BBOP/NNT)$$

$$P_{TT} = C \times NNT / BBOP$$

$$P_{TT} = CBOP / BBOP$$

Quantifying Treatment Effects

Incorporating Cost

Back to the Flu example

Treatment cost = \$40

BBOP = \$1080

NNT= 9

$P_{\pi} = 9 * 40 / 1080 = 0.33$

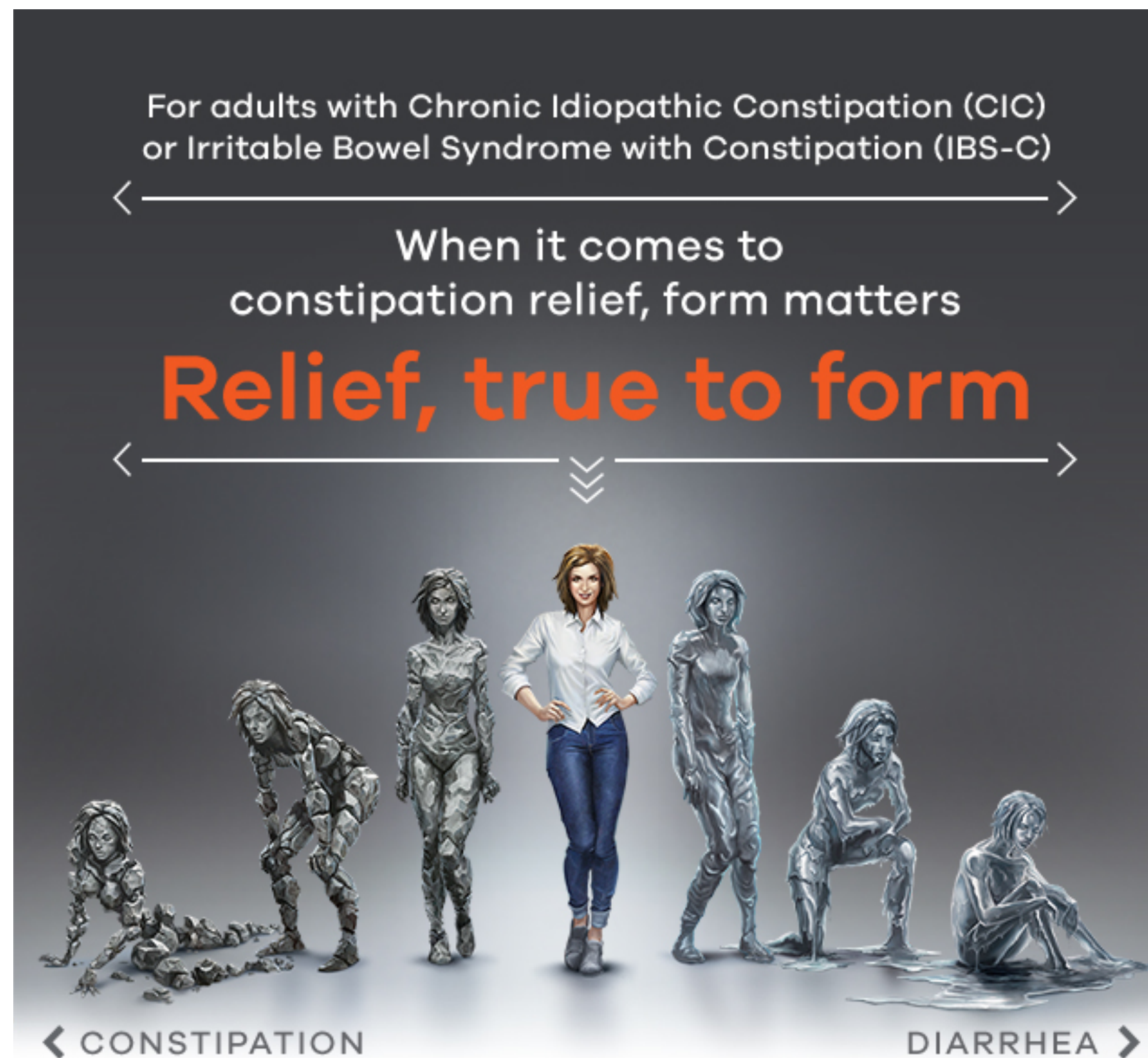
If you treat household contacts when the index cases's probability of the flu is 33% or higher, you will not spend more than the BBOP of \$1080 to prevent a case of flu

Quantifying Treatment Effects

Number Needed to Harm

Adverse effects of treatment

Trade-off between side effects and outcome prevention



Example:

Plecanatide (Trulance) to treat chronic constipation

Unintended side effect: Diarrhea

Quantifying Treatment Effects

Number Needed to Harm

	Diarrhea	No Diarrhea	Total	Risk
Plecanatide	28	446	474	5.9%
Placebo	6	452	458	1.3%

$$RR = 5.9\% / 1.3\% = 4.5$$

$$ARR = 1.3\% - 5.9\% = -4.6\%$$

$$ARI = 5.9\% - 1.3\% = 4.6\%$$

$$NNH = 1 / ARI = 21.7$$

For every 22 patients treated , we will cause one additional case of diarrhea

Quantifying Treatment Effects

Number Needed to Harm

The Trade Off

Trade-off between side effects and primary outcome prevention
Harms/Bad outcome prevented = ARI/ARR or NNT/NNH

Trulance example:

NNT for primary outcome 9.3

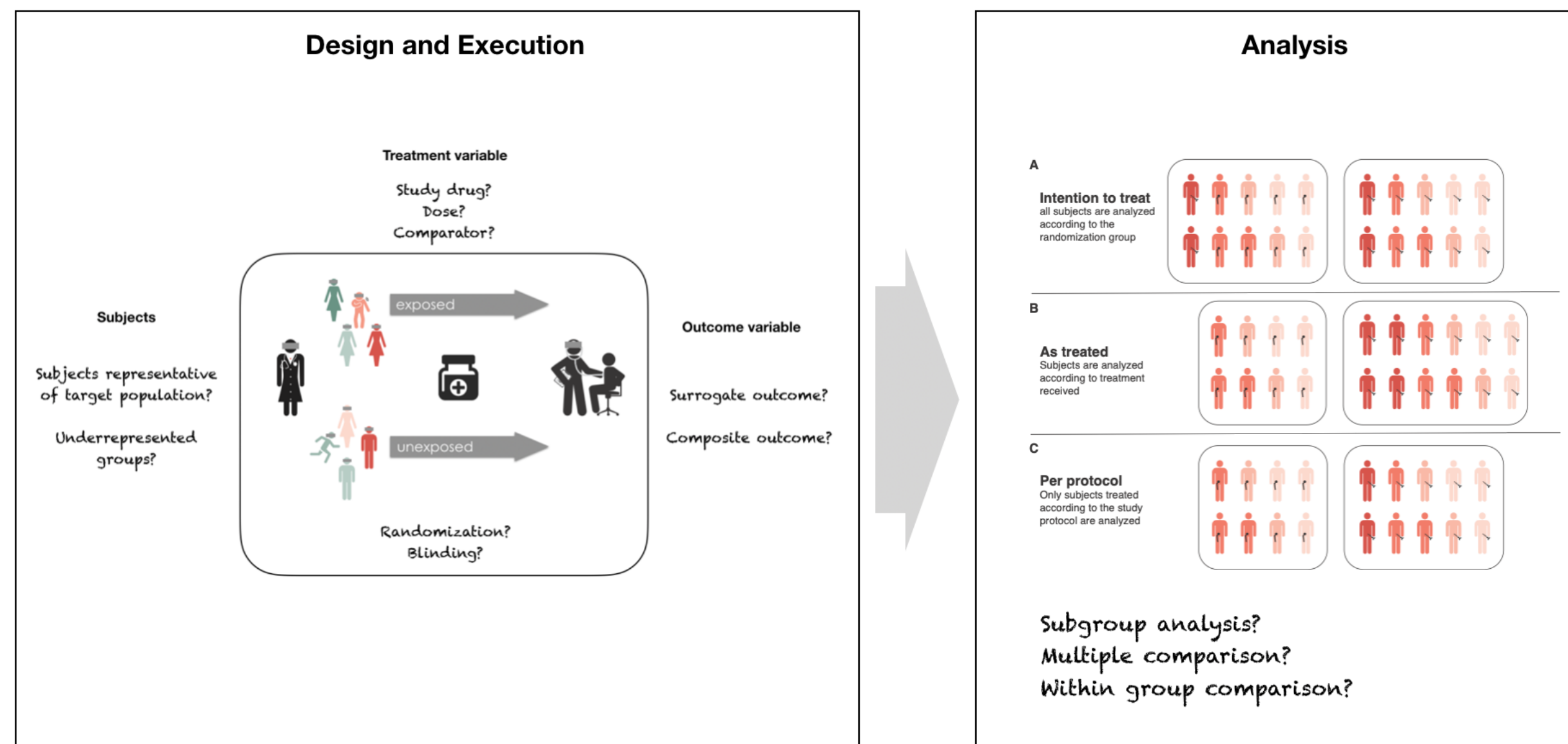
NNH: 22

$NNT/NNH = 9.3/21.7 = 0.43$

We cause 0.43 cases of diarrhea for each durable responder

Conclusions

Design, Execution and Analysis of RCTs



Treatment Effects in RCTs

- ARR takes baseline risk into account
- Practical results:
NNT
CBOP, BBOP
ARI and NNH