

A complex network diagram on the left side of the slide. It consists of numerous nodes of varying sizes and colors (black, grey, yellow, red) connected by thin black lines. A dense, blue, semi-transparent mesh of lines is overlaid on the network, suggesting a data flow or a specific subset of connections.

Model-Based Meta-Analysis (MBMA):

Optimizing Drug Development with Public Data and Predictive Models

Your Speakers



Jaap Mandema, PhD
Chief Innovation Officer



Matthew Zierhut
Vice President, Integrated Drug
Development

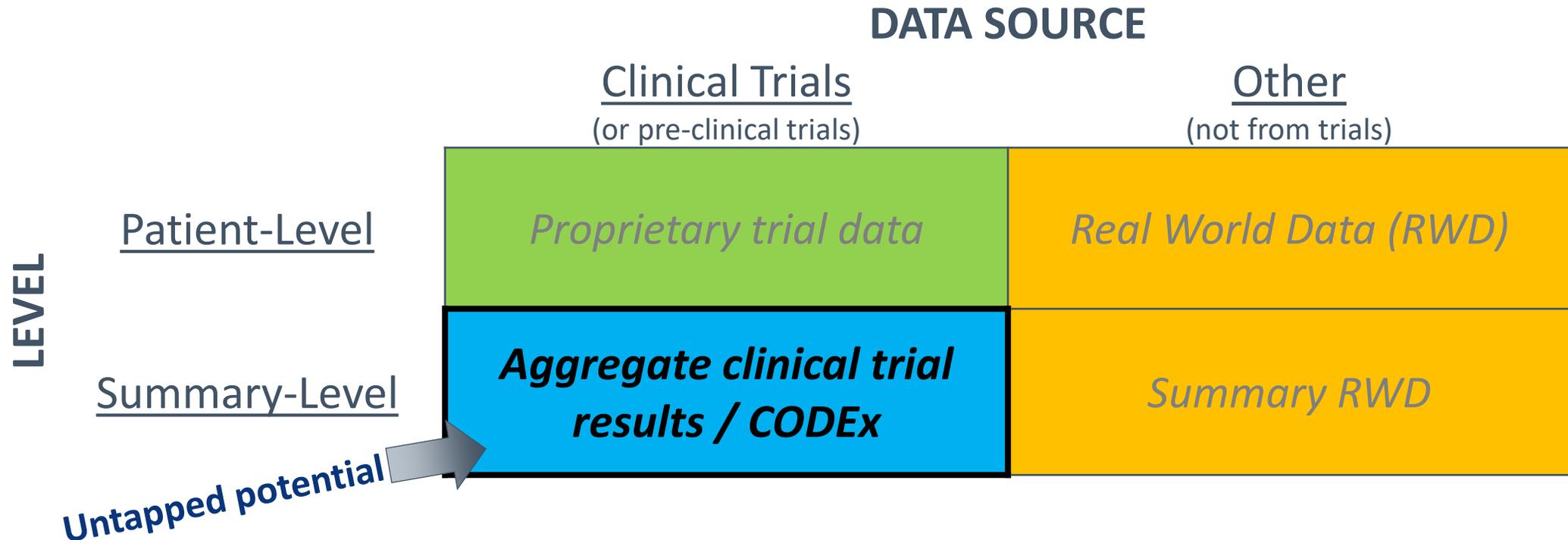
Questions We Will Answer Today

- What is the value proposition for meta-analysis and MBMA?
- How is MBMA different from other forms of meta-analysis?
- How has MBMA been used to answer real-world drug development questions?

Where does meta-analysis and MBMA fit into drug development?

Utilizing all available data can lead to better decisions:

- Drug development decisions (R&D)
- Commercial/Marketing or other critical decisions



Meta-Analysis Incorporates Valuable External Information Into Drug Development Decisions

- Comparative safety and efficacy
 - There is a need to evaluate new treatment options against other existing or emerging treatment options (indirectly) for go/no-go decisions, dose selection, trial strategy
- Endpoint-to-Endpoint relationships
 - Biomarker to clinical endpoint predictions
 - Bridging across indications
- Create synthetic control arms
 - Adjusted for known and accounting for unknown factors that impact heterogeneity
- Leveraging existing information
 - Similar shape of dose response relationships of drugs within class
 - Similar impact of disease severity on treatment effect
- Optimize Trial design
 - Impact of trial design features on placebo, treatment effect, and variability

How can meta-analysis and MBMA improve drug development?

- Rigorously establish safety/efficacy targets needed for differentiation
 - Understand what is truly needed to compete in market and achieve technical success
- Design cost-effective trials that improve probability of early correct development decisions
 - Link early endpoints/biomarkers to registration endpoints
 - Simulate future trials to optimize design and understand probability of trial success
- Iteratively quantify probability of technical success (PTS) as new data are available for each compound in portfolio
 - Quantify risk and get a data-based estimate of PTRS and eNPV
 - Maximize ROI on a trial-by-trial or program-by-program basis
- Additional benefits across the organization
 - E.g., simplify knowledge transfer from detailed trial networks

Key Steps in the Data Curation Process

- Formulation of study objectives and protocol
- Literature search
 - Pubmed, clintrials/eudract, company websites/CSRs, conference abstracts/posters/talks
- Review and quality assessment of data sources
- Selection of data sources for inclusion
- Data extraction into CODEx outcomes database
 - CODEx: Clinical Outcomes Database Explorer
- Standard procedures for this process are well defined in the Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) statement (<http://www.prisma-statement.org/>).

Clinical Outcomes Database Explorer (CODEx)

CODEx is a web-based software platform that stores curated clinical outcomes databases and provides an interactive platform to quickly investigate indication-specific databases and perform initial exploratory analytics

codex.certara.com/codex

The screenshot shows the CODEx web application interface. At the top is a blue navigation bar with the CODEx logo (By CERTARA) on the left and navigation links for Home, About Us, CODEx (underlined), Contact Us, and Account on the right. Below the navigation bar is the main content area. The heading 'CODEx' is followed by a welcome message: 'Welcome to CODEx. Use 'Your CODEx library' section to quickly find the databases you have access to. For general information about other databases, you can check the gallery underneath that. You can access the general descriptions of all databases in the gallery, but only the ones that your organization has a subscription for will be accessible via the CODEx interface.' Below this is the 'Your CODEx Library' section, which includes a message to 'Dear Certara IDD Consultants member' and a gallery of databases. The gallery has filter tabs for 'all', 'immunology', 'oncology', 'metabolic', 'cardiovascular', 'CNS', 'other', and 'sample'. A search bar with 'Submit' and 'Reset' buttons is present. The gallery displays four database cards: 'Acute Coronary Syndrome' (with a heart illustration), 'Acute Myeloid Leukemia' (with a blood cell illustration), 'Age-Related Macular Degeneration' (with a brain illustration), and 'Alzheimer's Disease' (with a brain illustration). The bottom row of the gallery is partially visible, showing four more database cards.

Results from over 400 RA studies are available

Filtering to relevant data enables impactful exploratory analyses

**CODEX**

[Overview](#) [Database](#) [Studies](#) [Exploration](#) [Pairwise MA](#) [Network MA](#) [Settings](#)

[Contact the team](#)

Rheumatoid Arthritis Clinical Outcomes Database

Data filters
[Apply filter](#) [Reset filter](#)
Filter variable: randomized.treatment
Filter values: tofacitinib 10 mg
 Filter on study-level
[Reset all filters](#)

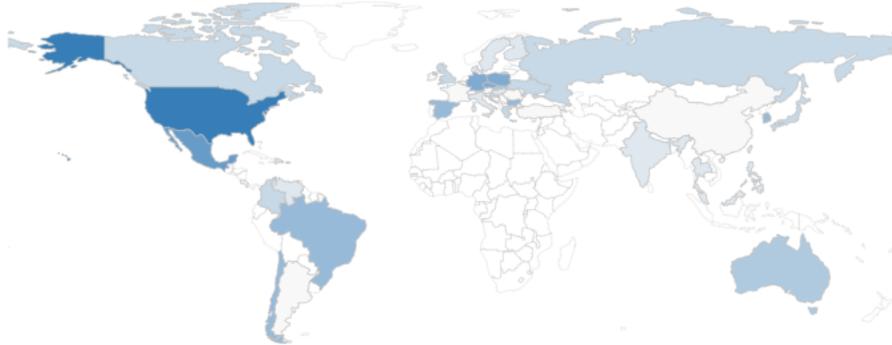
Active filters
Measure settings
Default measure: value
Default binary comparison: risk difference
Default continuous comparison: mean difference

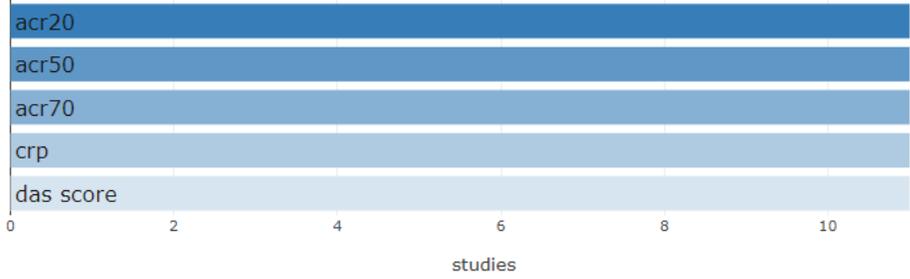
Data settings

Summary

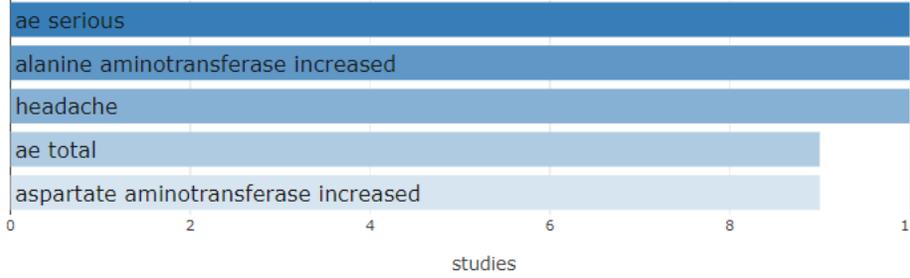
The full CODEX Rheumatoid Arthritis (RA) database contains summary-level endpoint data from 408 studies reported in 901 references. The most commonly reported efficacy and biomarker endpoints are acr20 (350 studies), das score (341 studies), acr50 (336 studies), haq (331 studies), and acr70 (319 studies).

Key	Complete	Selected*
References	901	40
Studies	408	11
Study arms	1273	11
Patients	137061	1546
Data Rows	179094	1648

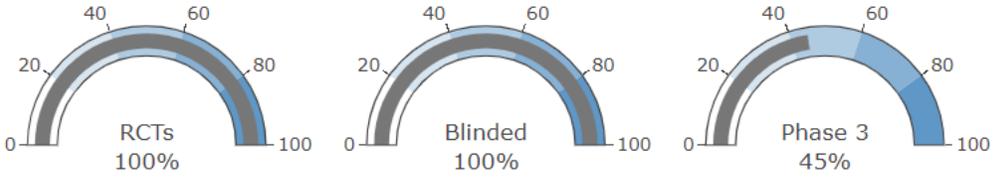
Study sites

Most common efficacy endpoints

Endpoint	Number of Studies
acr20	11
acr50	11
acr70	11
crp	11
das score	11

Most common safety endpoints

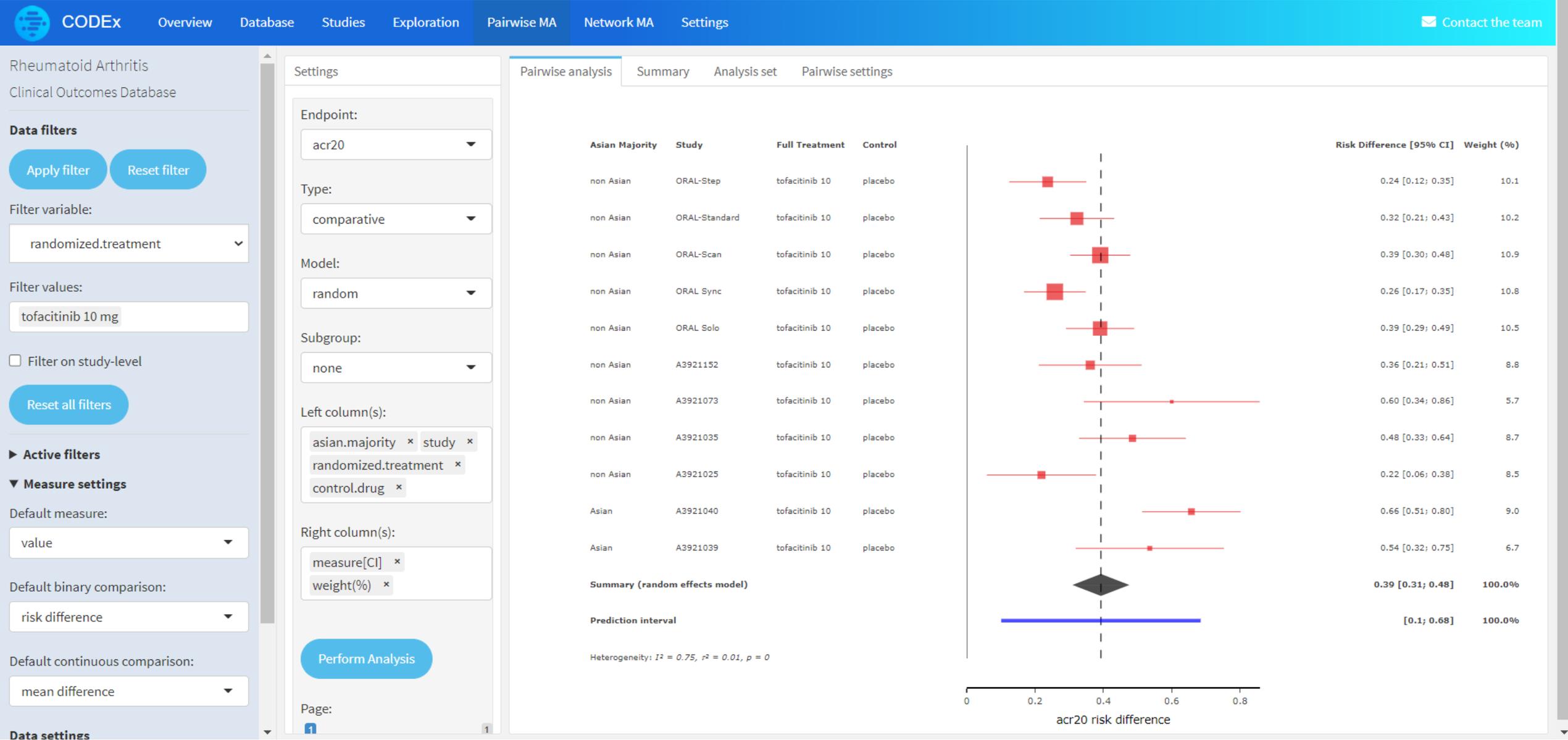
Endpoint	Number of Studies
ae serious	11
alanine aminotransferase increased	11
headache	11
ae total	10
aspartate aminotransferase increased	10

Evidence characteristics

Characteristic	Percentage
RCTs	100%
Blinded	100%
Phase 3	45%

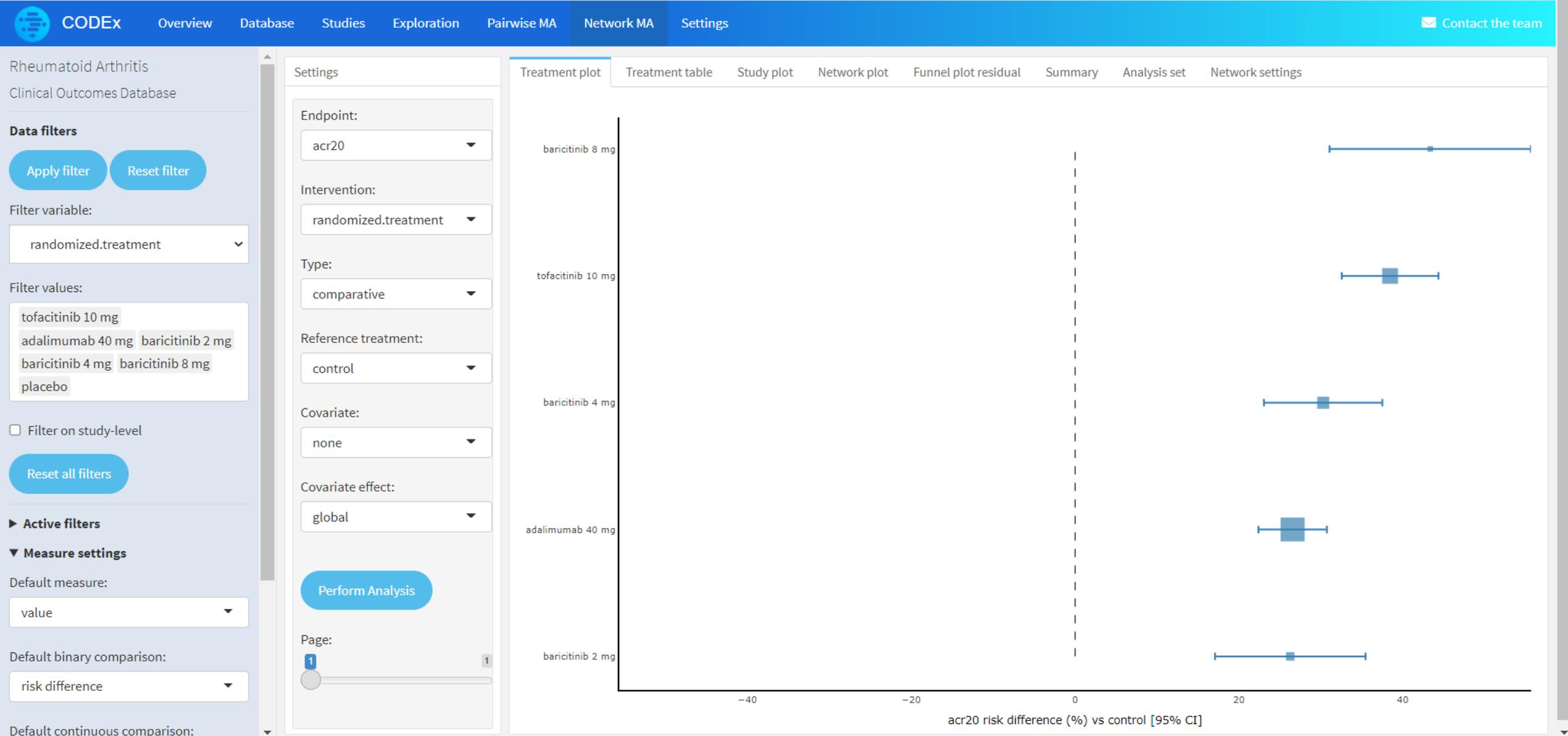
Significant between-trial heterogeneity in tofacitinib treatment effect

Asian majority trials may help explain some of this heterogeneity



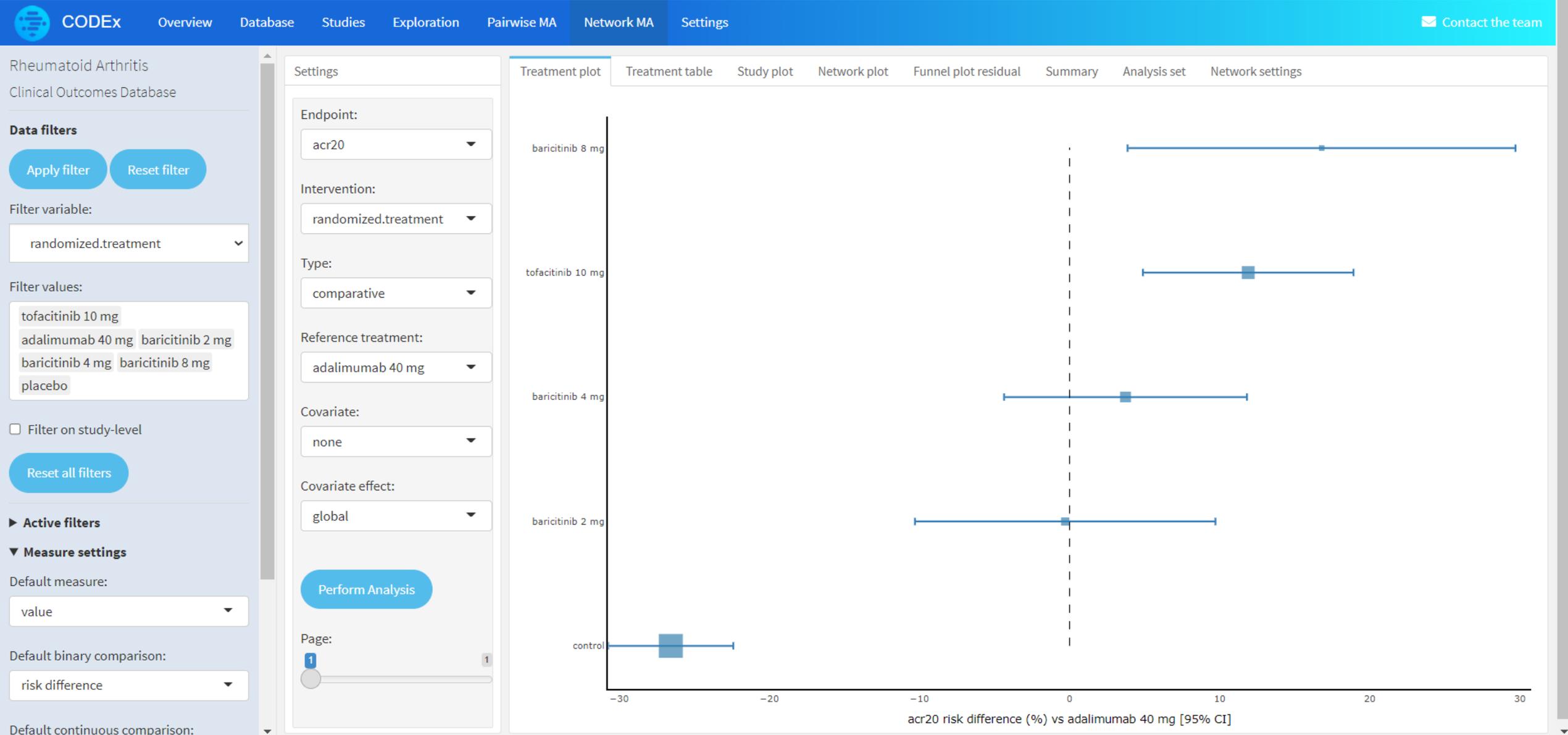
Adalimumab, baricitinib, and tofacitinib are more effective than placebo

Baricitinib has apparent dose-response

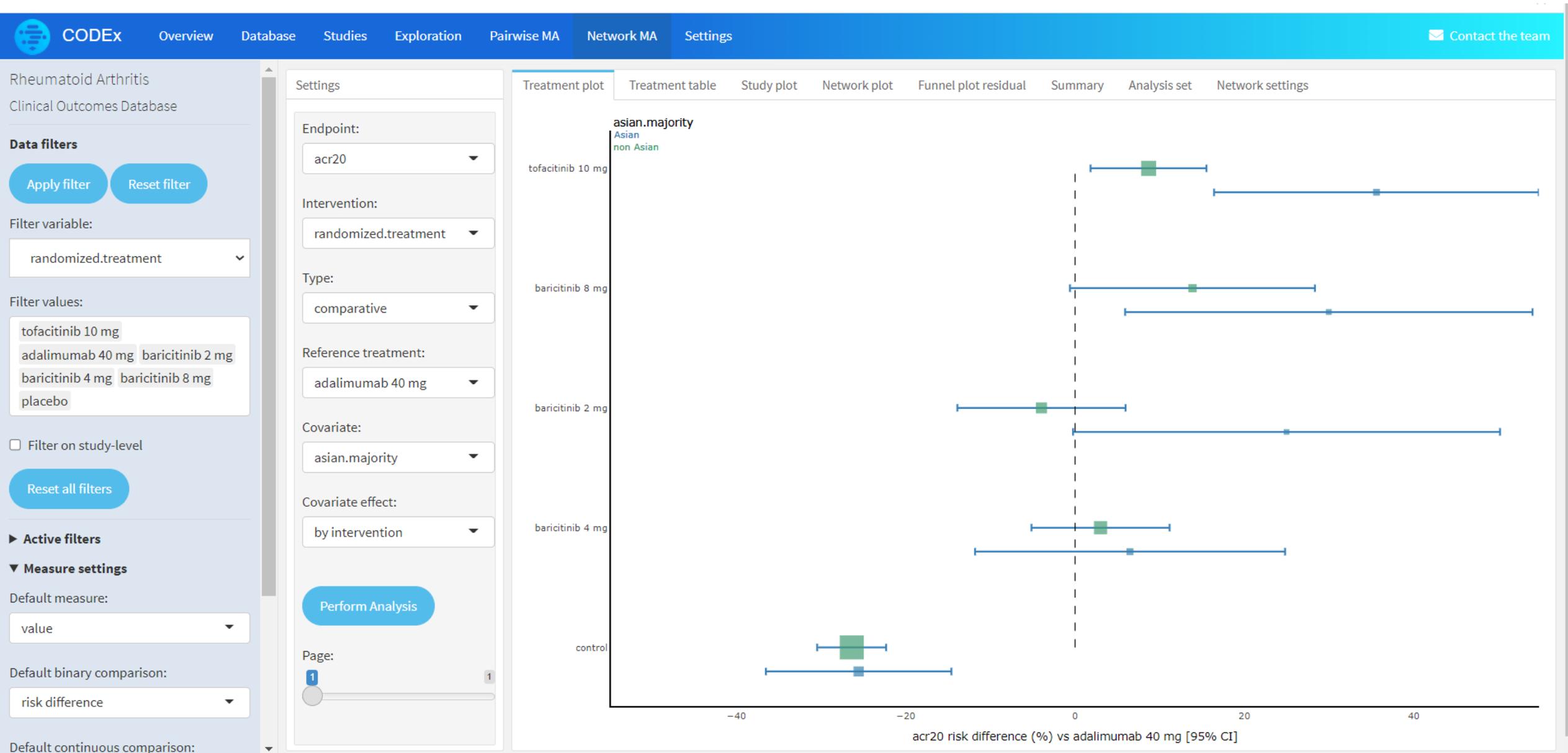


Baricitinib dose-response more apparent versus adalimumab

Baricitinib 8mg and tofacitinib 10mg appear more effective than adalimumab

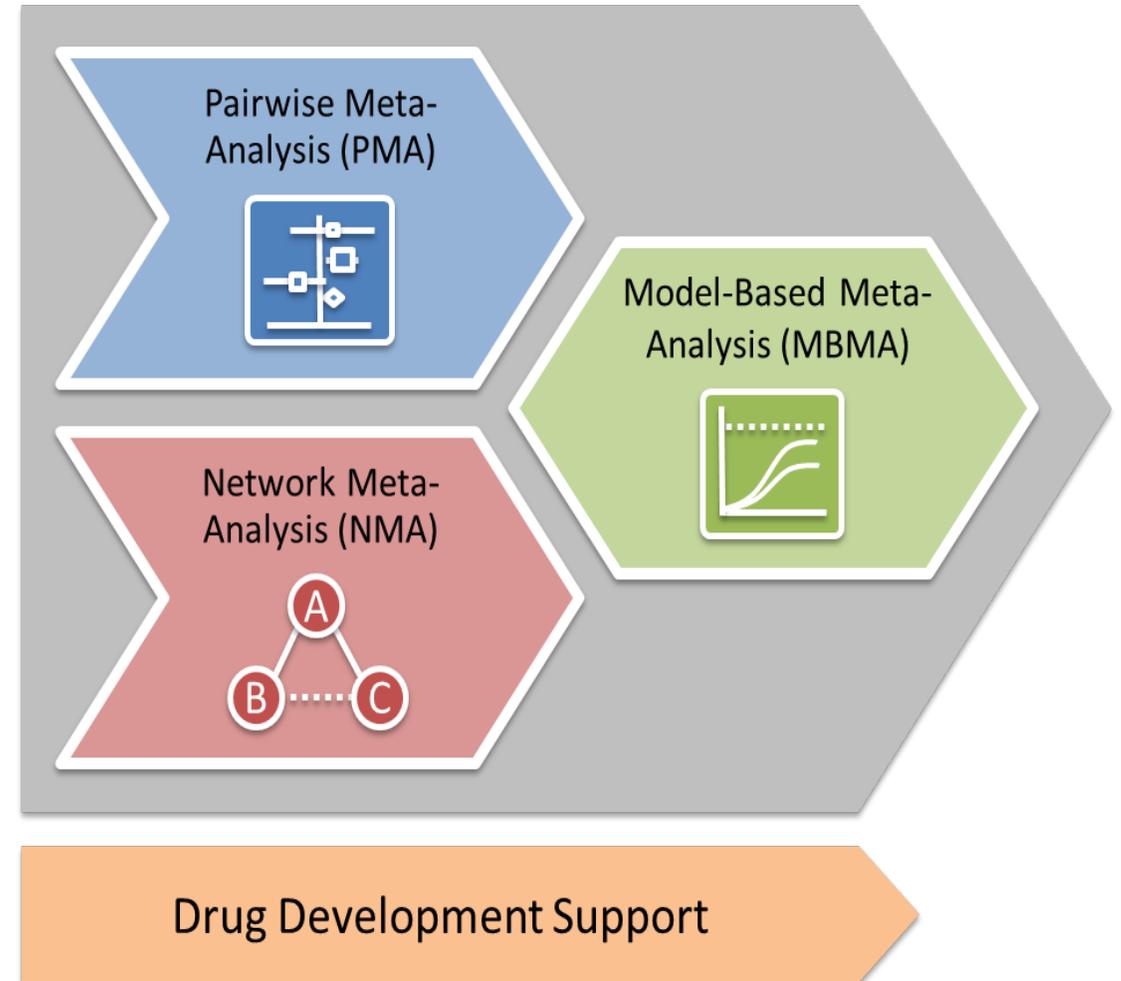


Majority Asian trials have an impact on anti-TNF treatment effect



MBMA takes advantage of pharmacological/physiological insights

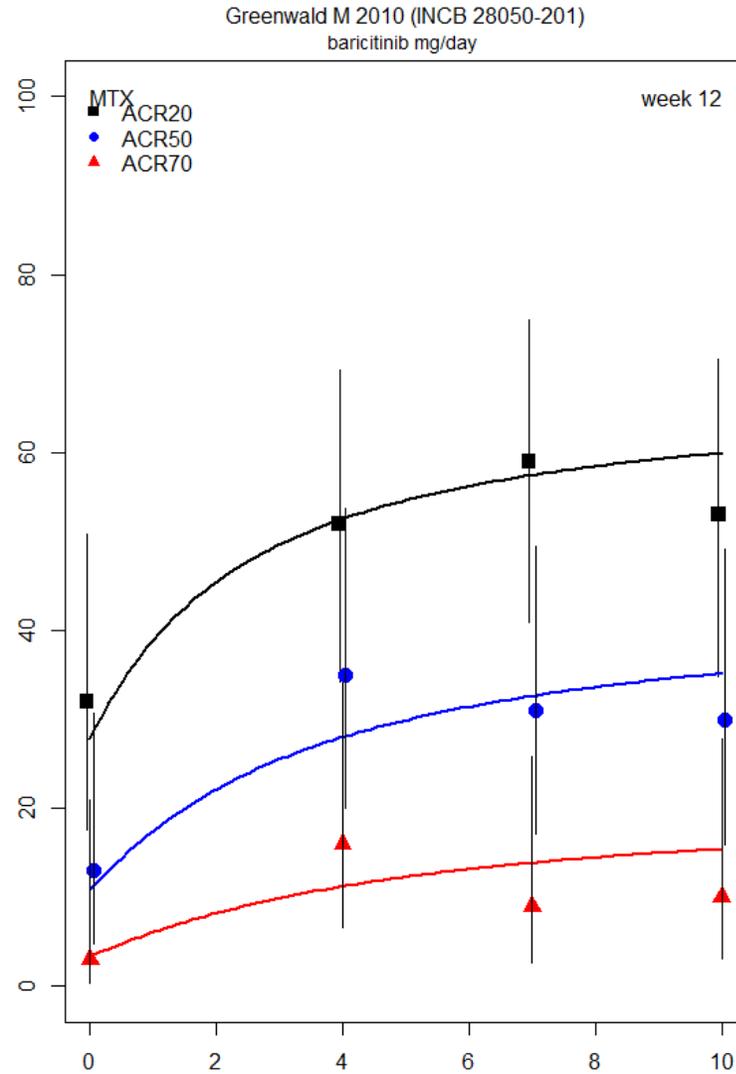
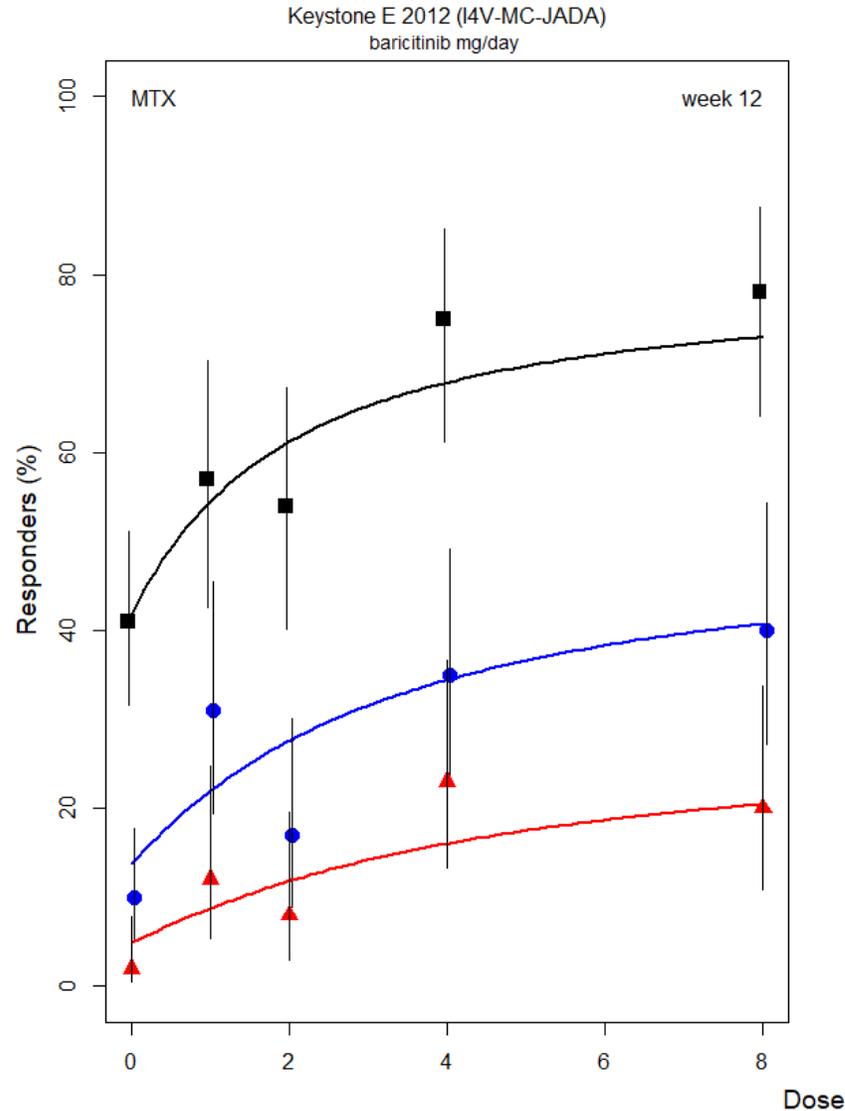
- Model-based meta-analysis (MBMA) is a type of meta-analysis that incorporates parametric models for the effect of treatment, time, and patient population characteristics on the outcomes
 - Explicitly incorporates the effect of dose and duration using standard pharmacology models and assumptions
 - Can include trial-level covariate relationships on the dose-response models to account for between trial differences in patient populations
- Allows simultaneous modeling of multiple endpoints and can therefore link biomarkers to clinical endpoints or early to late endpoints
- Like network meta-analysis, MBMA can provide indirect comparisons and simulations of head-to-head trials, but may use (longitudinal) dose-response models for individual drugs or drug classes
- Can be used for simulations of trials and predictions of trial success



PHASE II to III decision point for baricitinib in RA

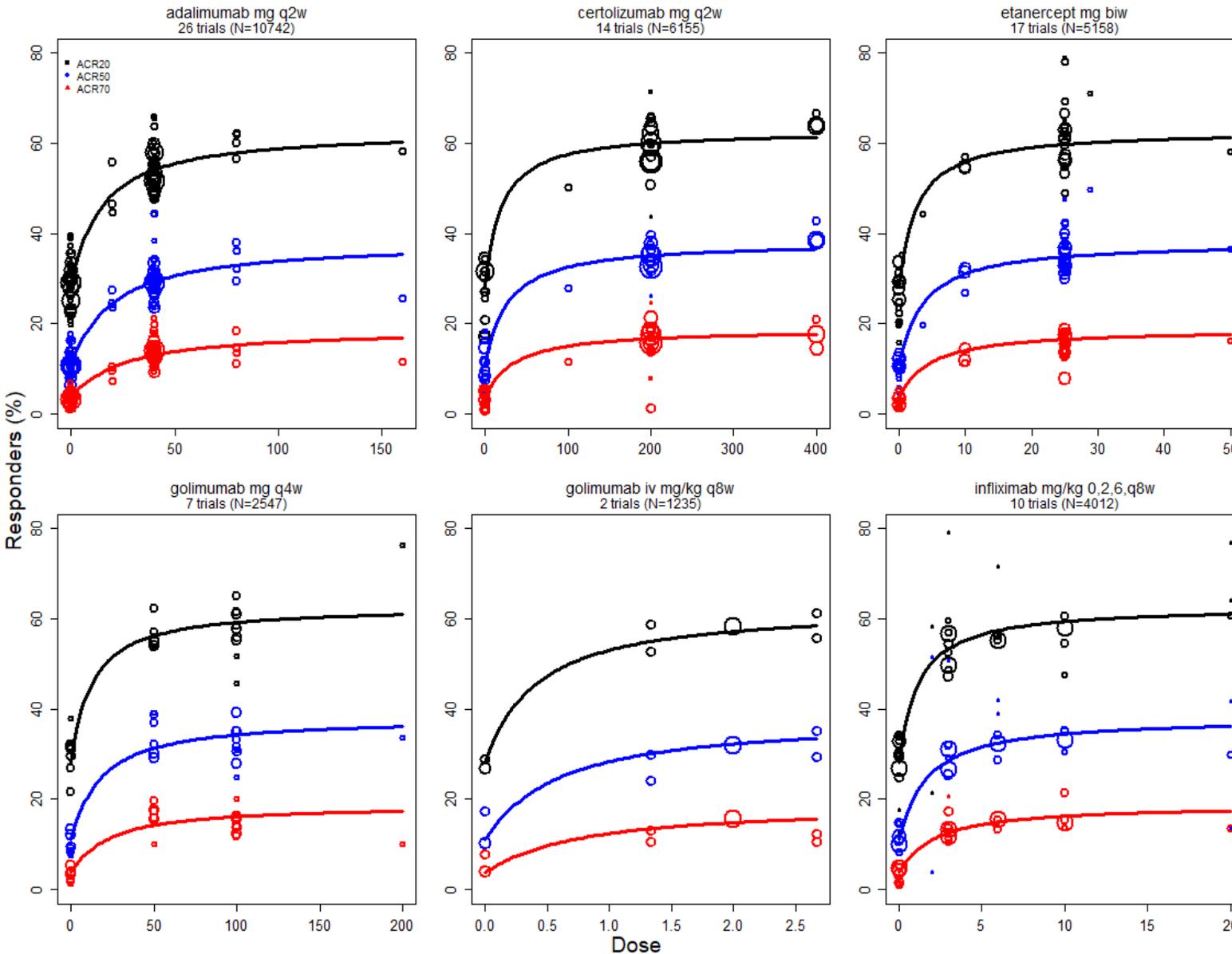
Typical MBMA application is to compare phase II efficacy/safety results for novel treatment options against other treatment options; marketed or in development

Phase II results in RA for baricitinib, a JAK inhibitor



- What is the probability to have a better efficacy/safety profile vs. other treatments?
- What is the best endpoint to differentiate our compound?
- Can we increase precision of dose response by assuming similar Emax as other JAKs?
- Should we run a phase III trial vs. competitor? Which one? Superiority strategy? or NI? at what margin?
- Is there a difference in DR (Emax or ED50) between TNF experienced and naïve patients? MTX experienced vs. naïve?
- Would 1 phase II have been sufficient?
- Could we have reduced the time frame (4 week vs. 12-week study)?
- Could we have use a synthetic control?
- Can we quickly expand to other indications: psoriatic RA, psoriasis, Crohn's, UC?

MBMA of time course of ACR20/50/70 shows that ACR dose response relationship is consistent and well characterized for all TNF inhibitor across trials (symbol size~precision)



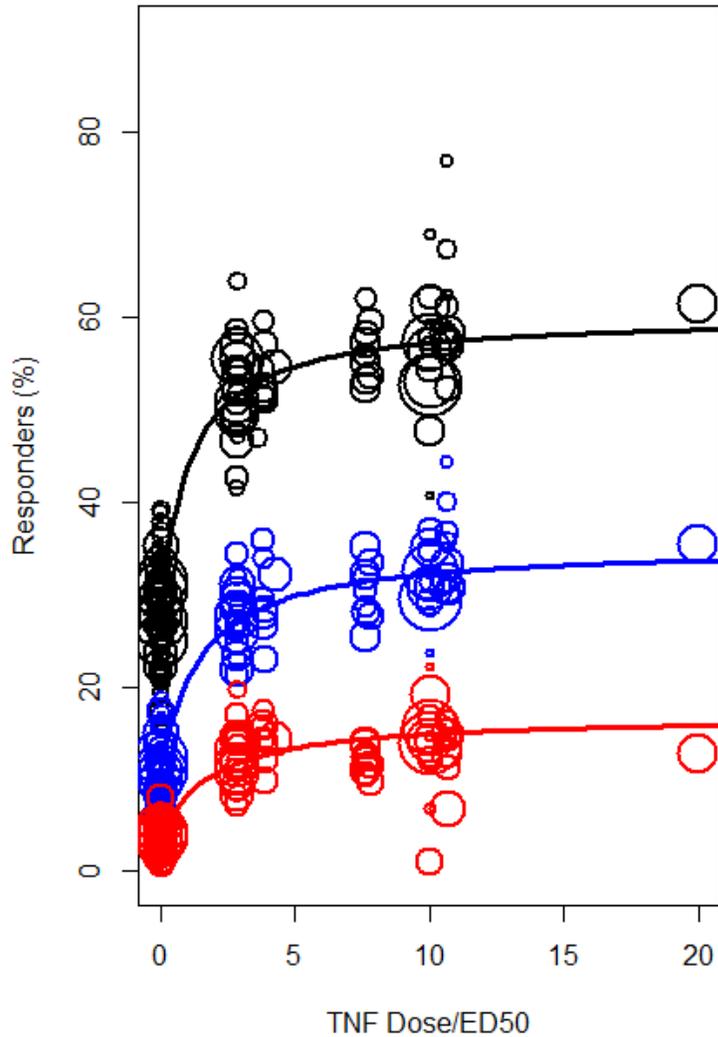
$$P(ACR)_{ijkt} = \text{logit}^{-1} \left(Eo_{itk} + f(\text{Drug}_{ij}, \text{Dose}_{ij}, t, X_{ij}, \theta_{(i)}) * e^{\eta_{ai}} \right)$$

- Eo_{itk} represent the placebo response (intercept) accounting for the trial-to-trial variability in overall response. Eo_{itk} represents a fixed-effect estimate for every time point in a study and a fixed-effect for every endpoint in a study (i.e. shift between ACR20, ACR50, and ACR70).
- $f(\text{Drug}_{ij}, \text{Dose}_{ij}, t, X_{ij}, \theta_{(i)})$ was the model for the treatment effect for each drug, based on dose, time and relevant covariates.
- Trial-to-trial variability in treatment effect for each drug was described by study specific random effects η_{ai} with mean 0 and variance ω^2 .

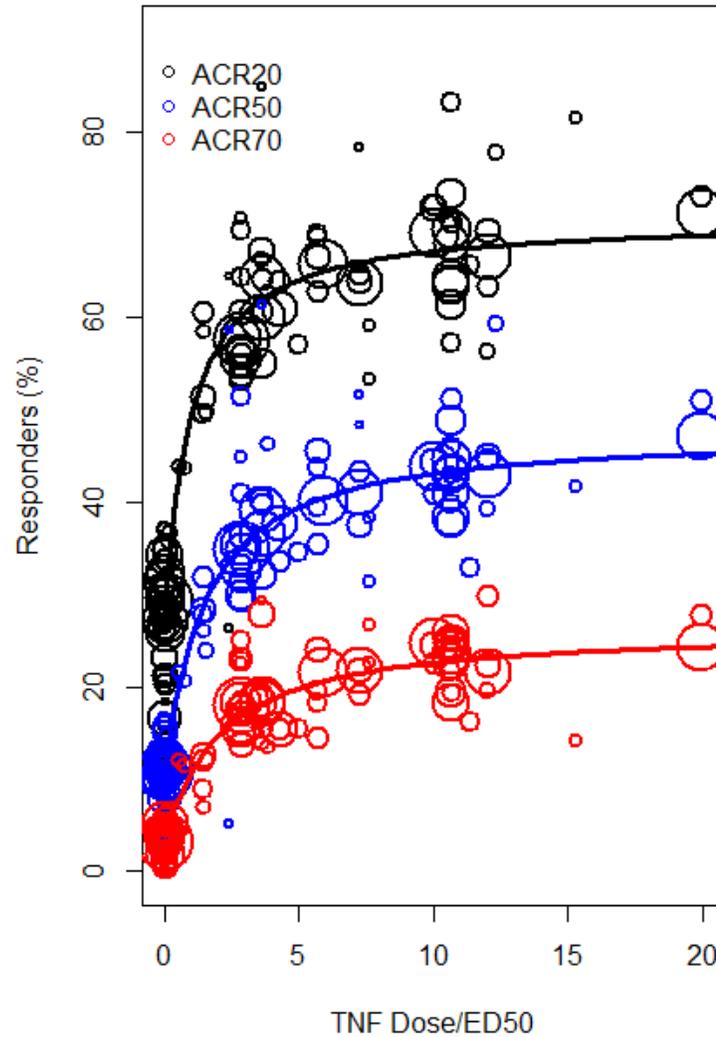
Treatment response for ACR is significantly dependent on baseline CRP

The decline of baseline CRP over the past 20 years explains the drift in treatment effect for anti-TNFs

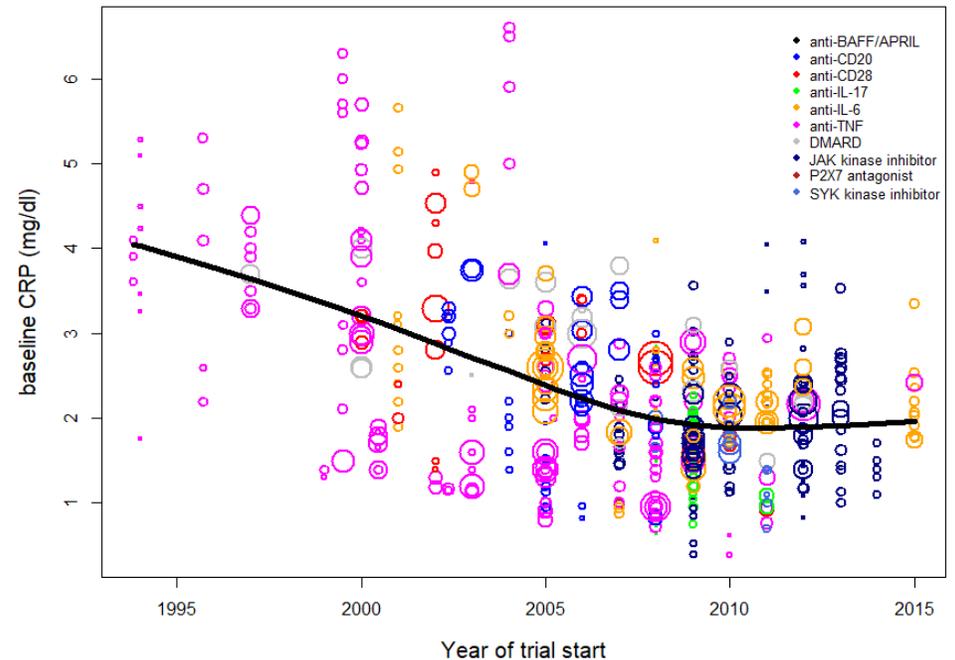
CRP 0.9 to 2.5 mg/dl



CRP 2.5 to 6.9 mg/dl

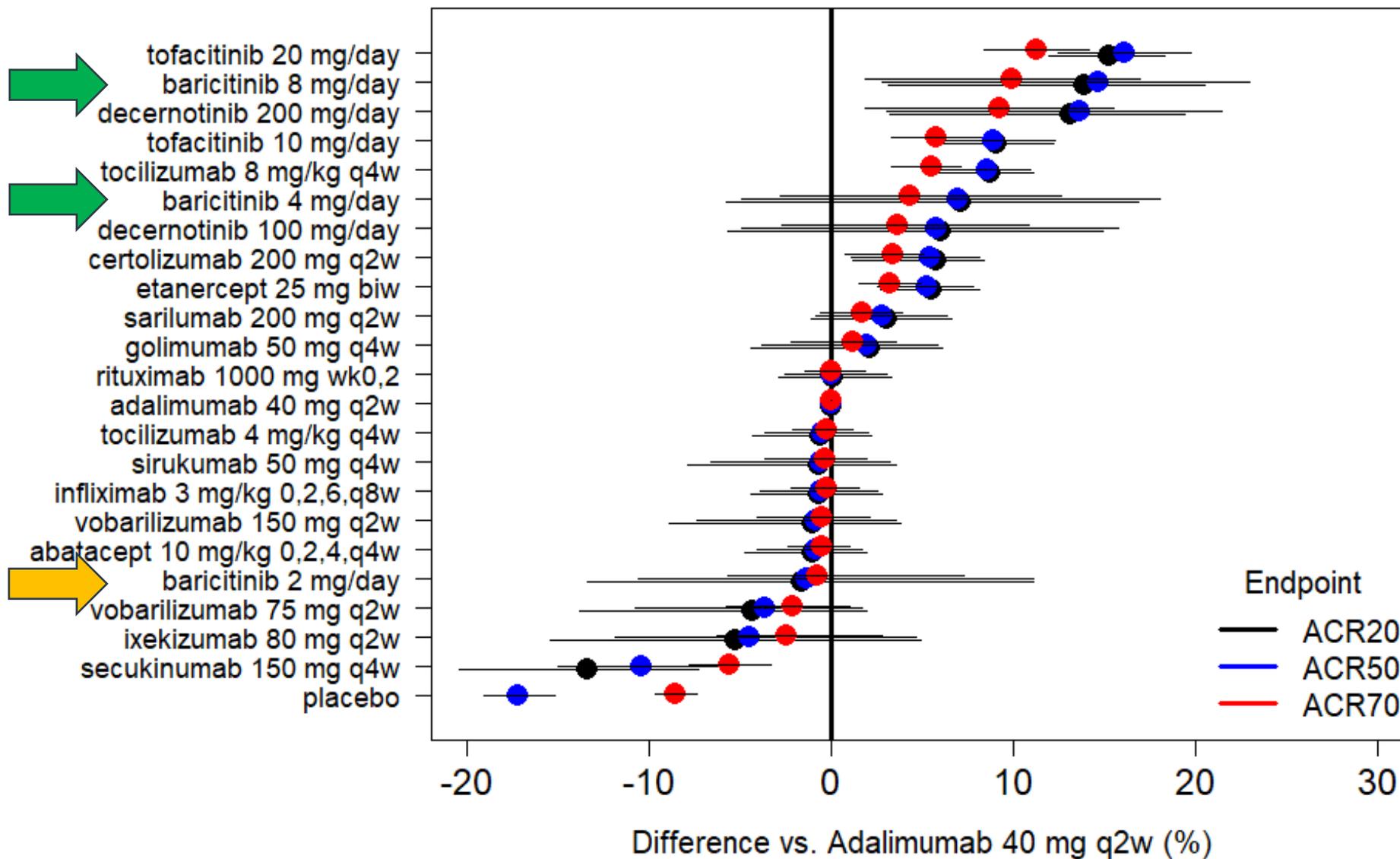


- Treatment effect was dependent on: baseline CRP, % of patients on background MTX, and Asian/non-Asian (especially for JAKs).
- 8.5% more ACR20; 7.9% more ACR50 responders for 1 mg/dl increase in mean baseline CRP



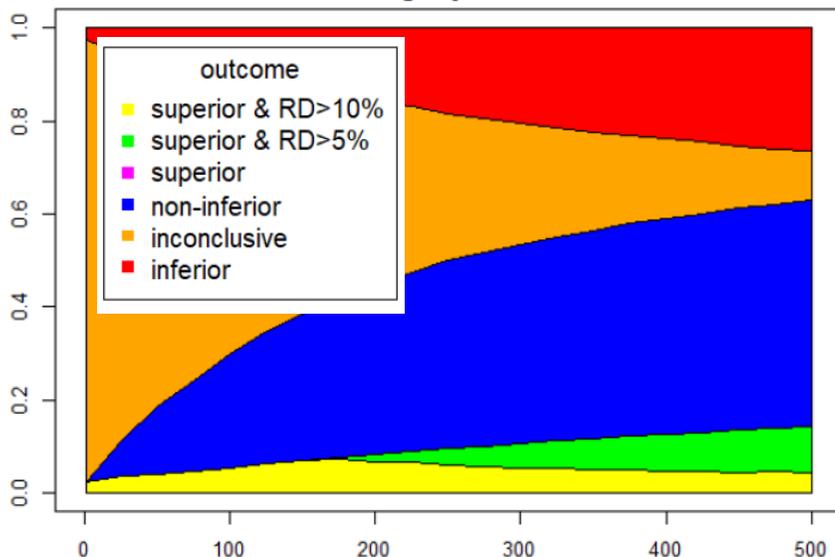
Comparative efficacy for baricitinib relative to other treatment options in RA [95% CI]

baseline CRP=2 mg/dl; MTX background; placebo=11% for ACR50

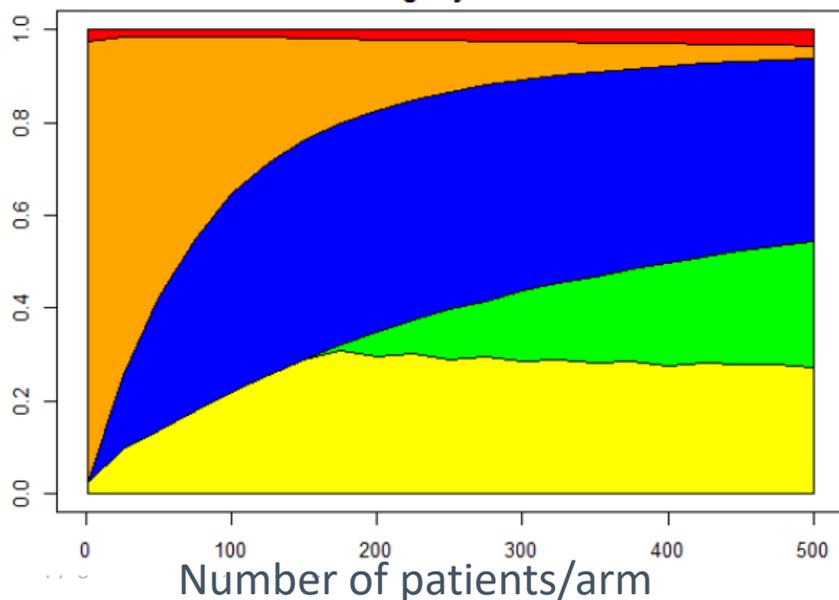


Probability of phase III trial outcome for ACR20 comparing baricitinib to 40 mg q2w adalimumab compared to observed outcome of RA-BEAM trial

baricitinib 2 mg/day vs. adalimumab

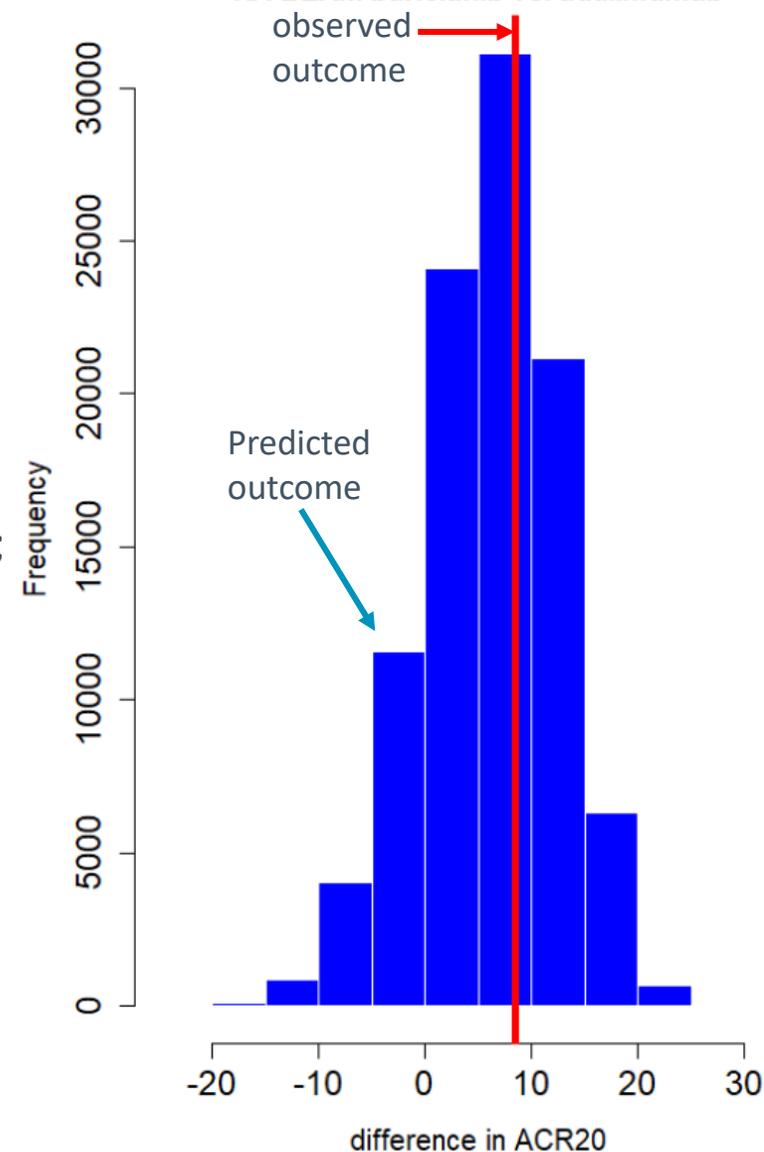


baricitinib 4 mg/day vs. adalimumab

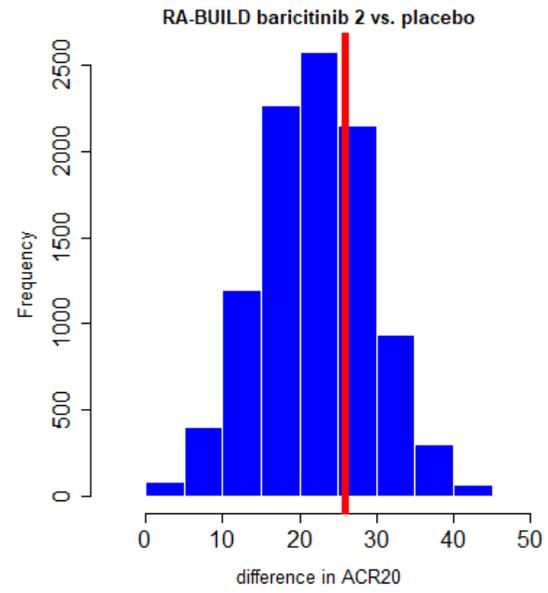
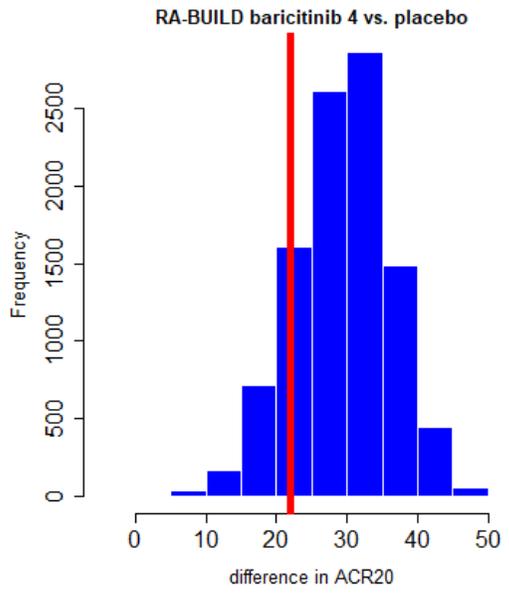
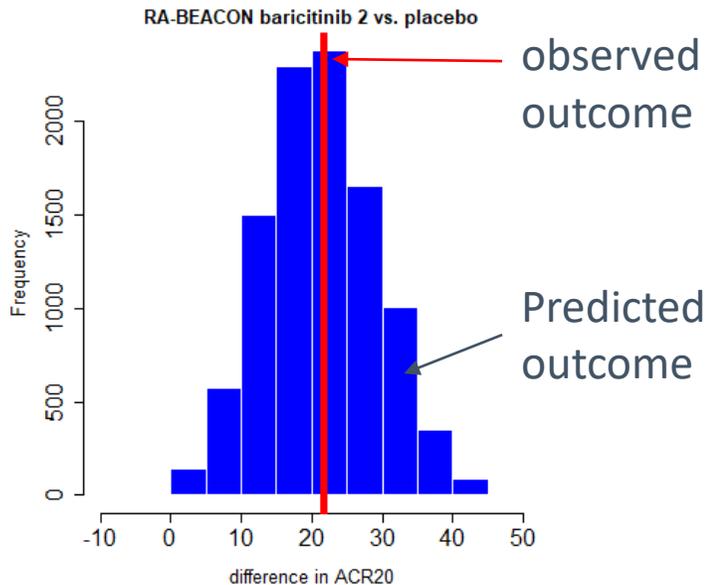
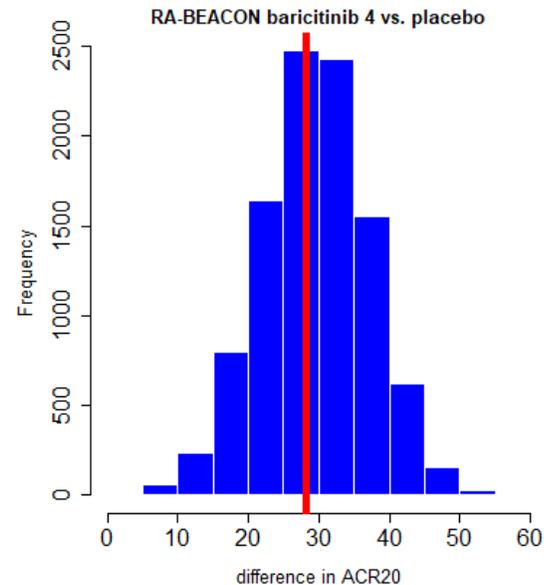
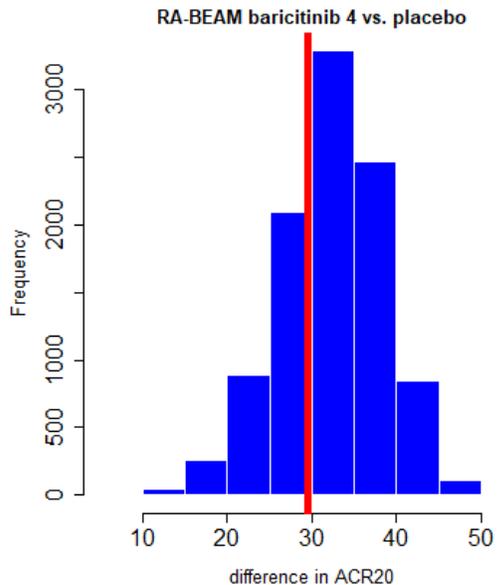


- Baricitinib 4 mg/day vs. adalimumab 40 mg q2w
- Trial simulation predicts a high, 92%, probability of non-inferiority at 400/arm and 10% margin;
- and a high probability of 50% for superiority
- Predicted trial outcome for
 - ACR20 6.7% [-5.0 to 16.0; 90% PI];
 - ACR50 7.9% [-5.0 to 19.5],
 - ACR70 5.8% [-3.8 to 16.0];
- Actual trial results showed superiority with a treatment difference of
 - ACR20 8.4% [1.7 to 15.1; 95% CI];
 - ACR50 10.1% [3.3 to 16.9]
 - ACR70 6.2% [1.2 to 11.2]

RA-BEAM baricitinib vs. adalimumab



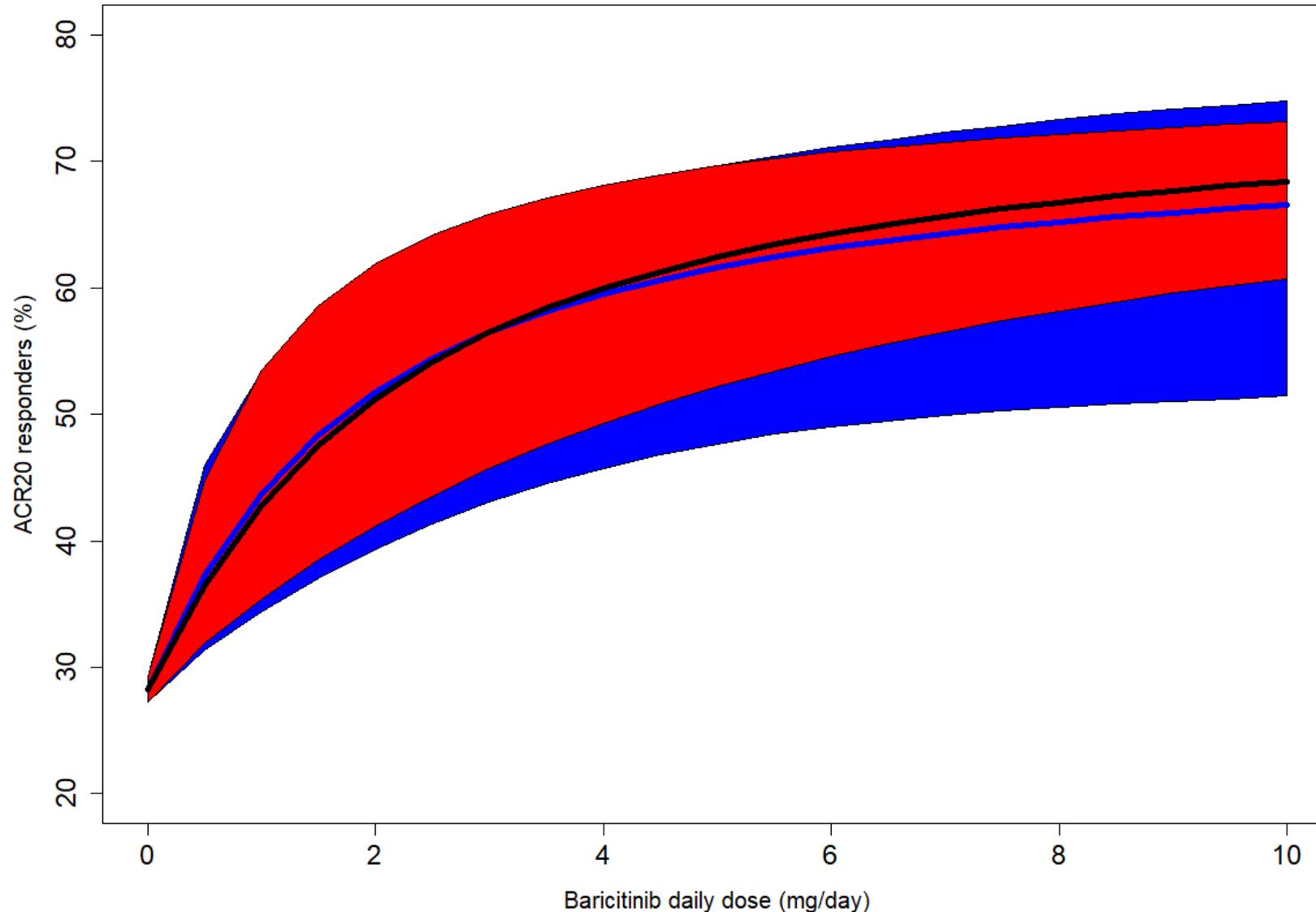
Close agreement between simulated and observed trial outcomes of phase III program confirms that we can evaluate design options based on probability models (all simulations based on phase II data only)



Phase III differed from phase II in trial duration; failed prior treatments; background treatments; patient characteristics

trial	Phase	dose	control	background treatment	failed treatment	duration	N
I4V-MC-JADA	2	1,2,4,8 mg	placebo	MTX	MTX	12	301
INCB 28050-201	2	4,7,10 mg	placebo	MTX	MTX	12	124
RA-BEACON	3	2,4 mg	placebo	MTX	biologic	24	527
RA-BEAM	3	4 mg	adalimumab	MTX	MTX	52	1305
RA-BEGIN	3	4 mg	MTX	none	none	52	584
RA-BUILD	3	2,4 mg	placebo	MTX	MTX	24	684

Pharmacological assumption of similar maximal effect among JAK inhibitors could have increased confidence in phase III decisions

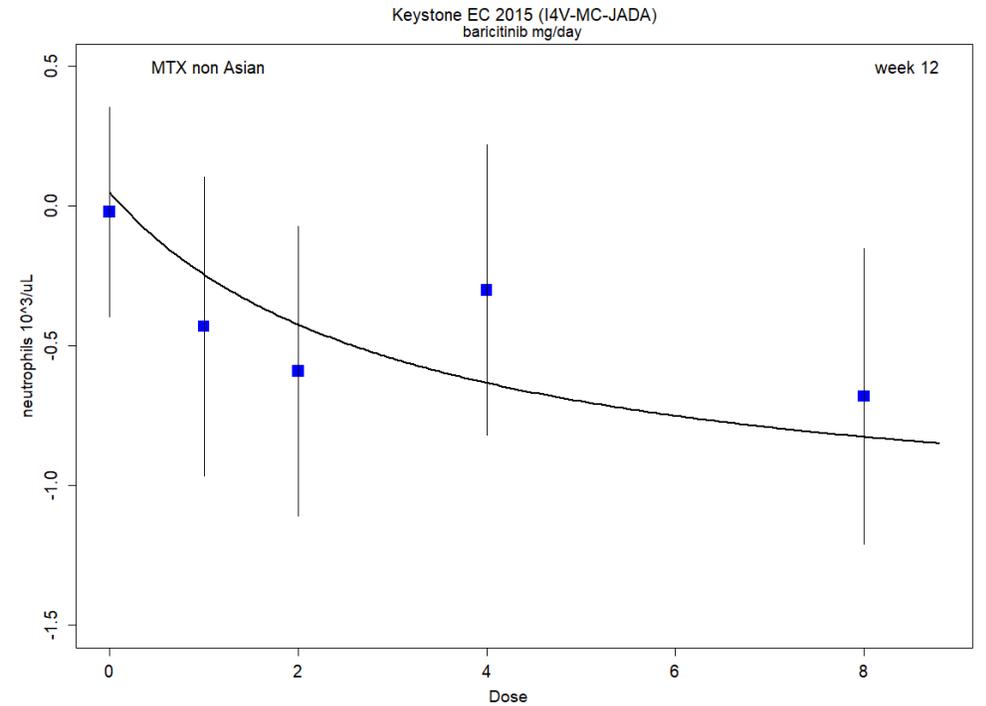
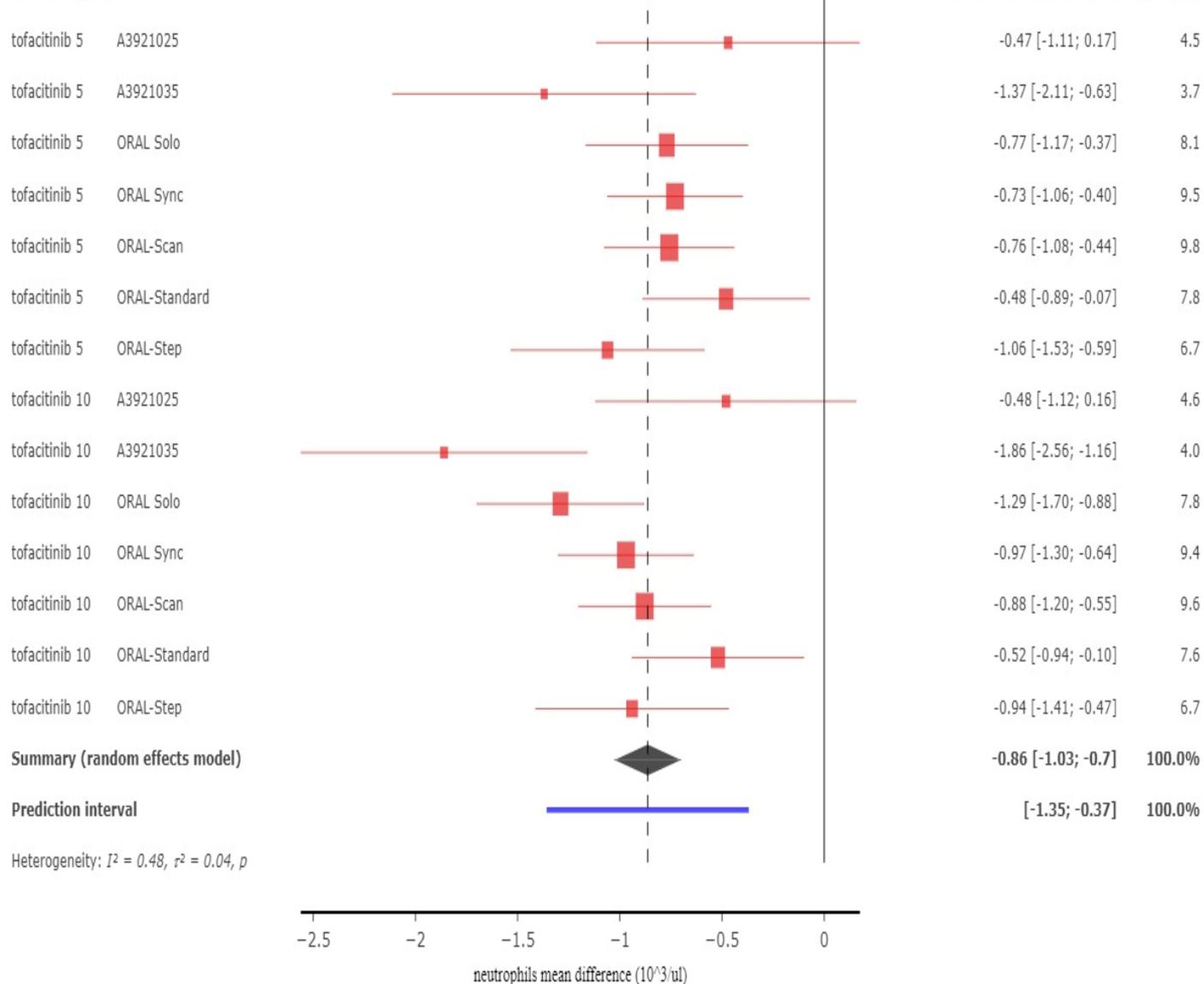


BLUE: estimated dose response at end of phase II based on baricitinib data alone

RED: estimated dose response at end of phase II assuming a similar Emax among JAK inhibitors

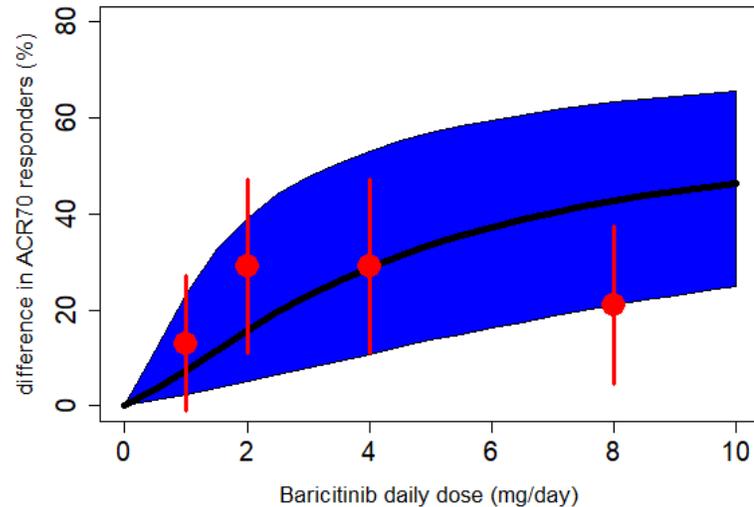
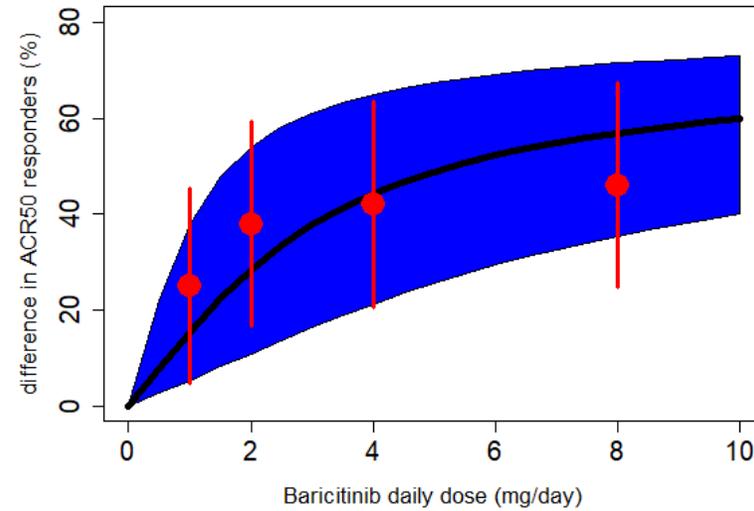
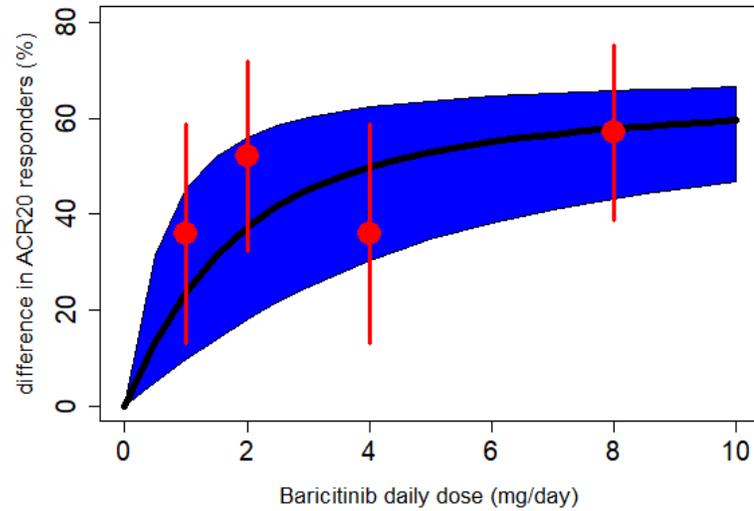
Comparison of neutrophils vs. ACR Dose Response shows a difference in Therapeutic Index among the JAK inhibitors

Full Treatment Study



Drug	ED50 neutrophils (mg/day)	ED50 ACR (mg/day)
Baricitinib (JAK 1/2)	3.2 [1.5 - 6.8]	1.8 [1.3 - 2.6]
Tofacitinib (JAK 1/3)	8.4 [5.0 - 14.3]	4.3 [3.1 - 6.1]
Filgotinib (JAK 1)	109 [43 - 272]	115 [64 - 205]

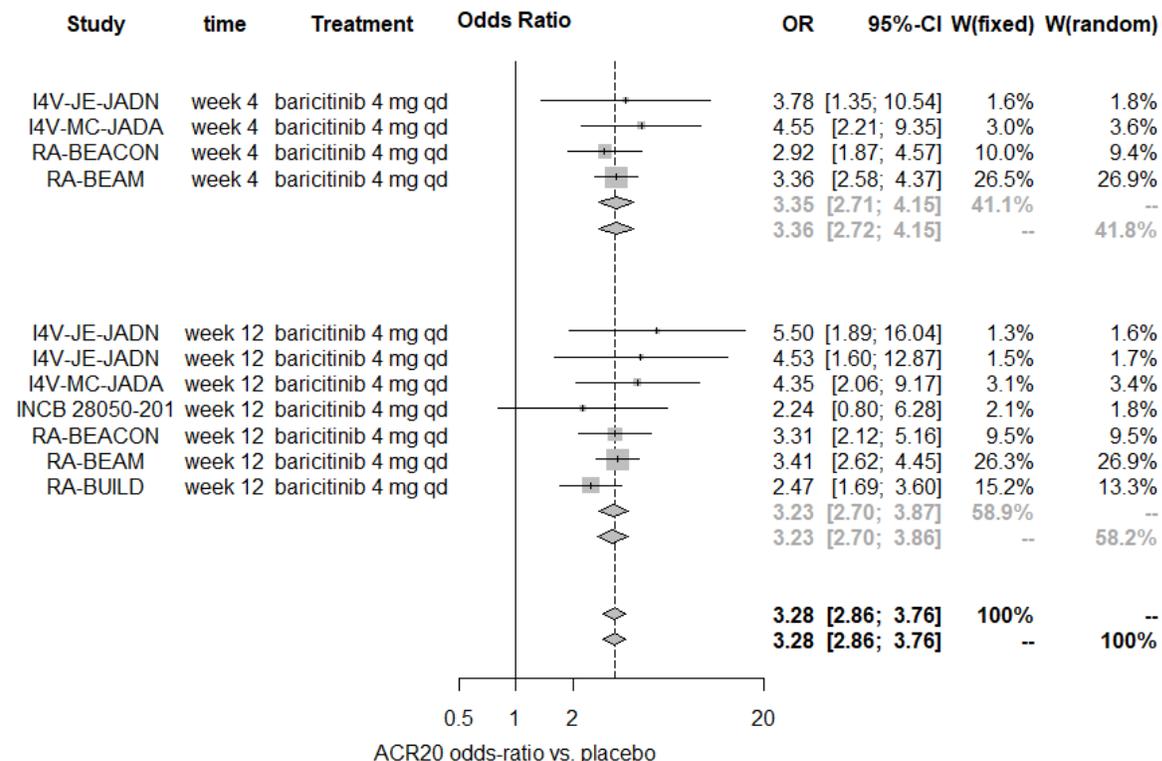
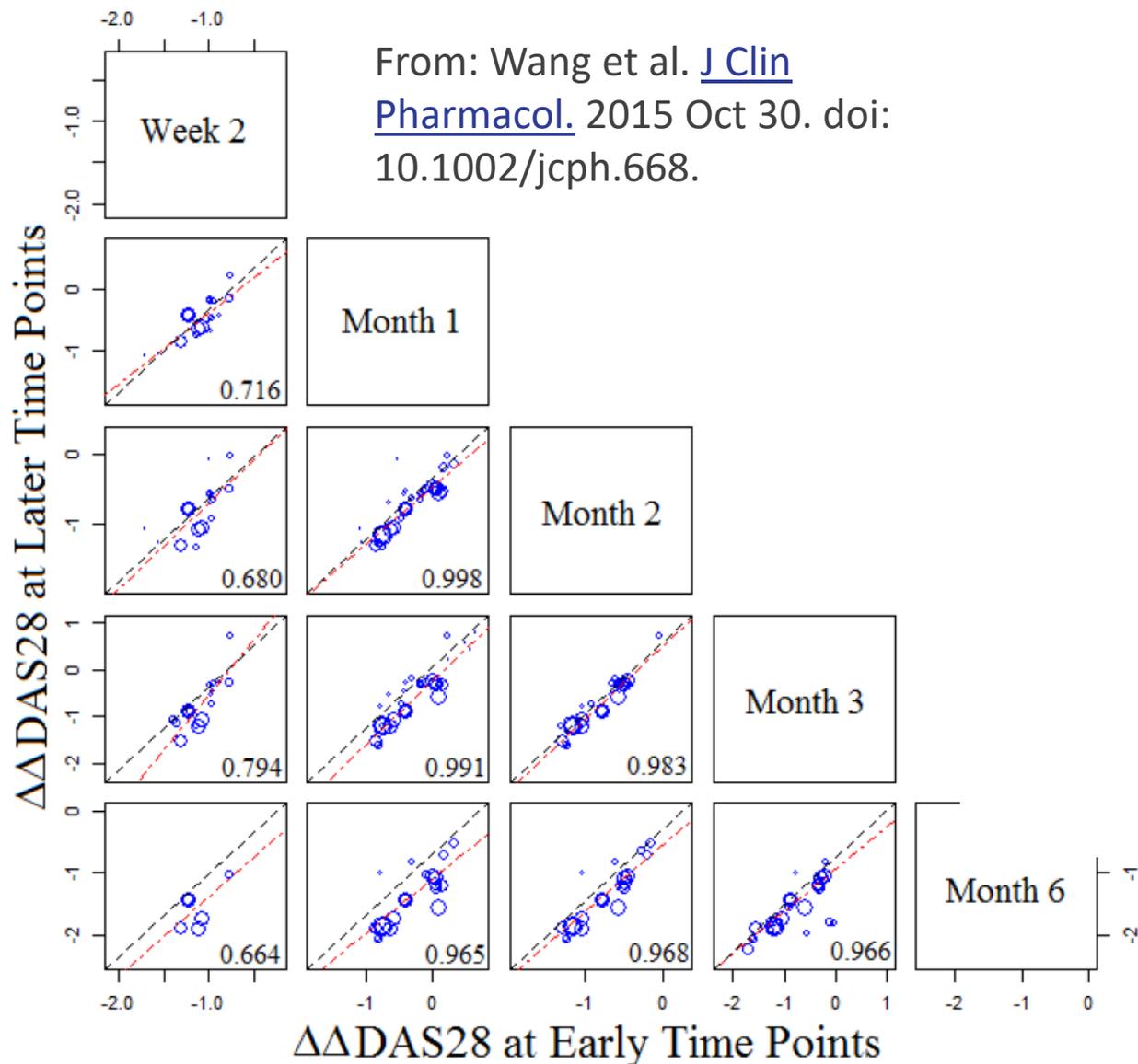
The difference from placebo in ACR response in the Japanese phase II study I4V-JE-JADN is well predicted based on tofacitinib data



The predicted effect is 46% [28 to 64%; 95% CI] greater in Japanese studies vs. ROW; this difference was estimated from tofacitinib and applied to baricitinib

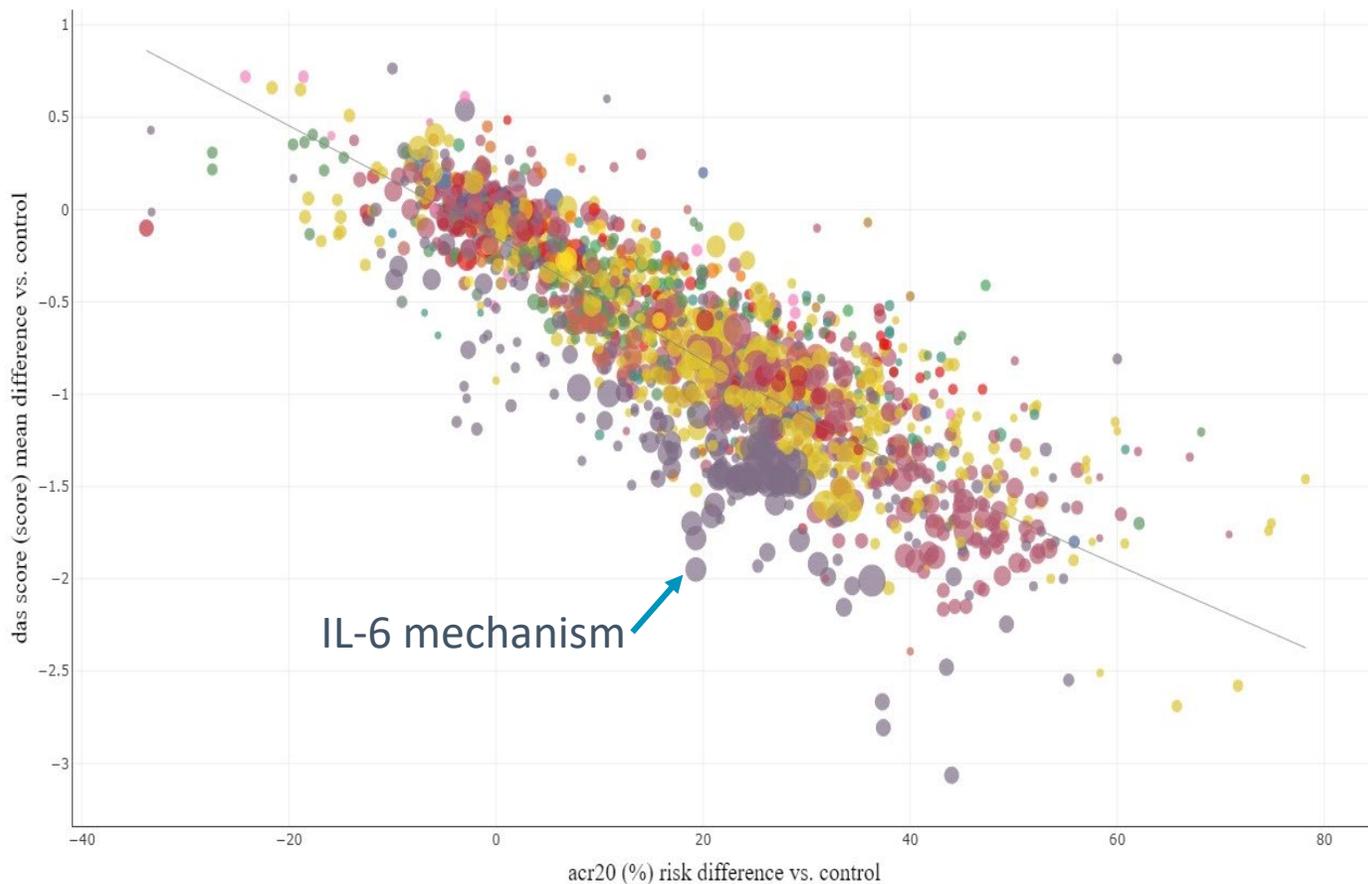
Week 4 treatment effect is predictive for later time-points

From: Wang et al. [J Clin Pharmacol.](#) 2015 Oct 30. doi: 10.1002/jcph.668.

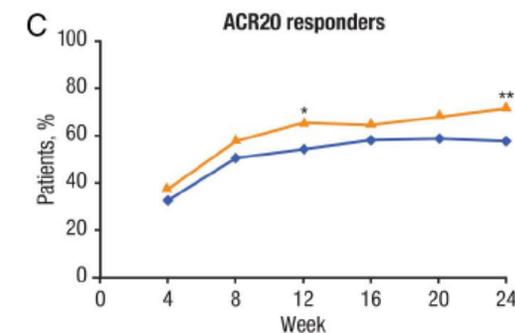
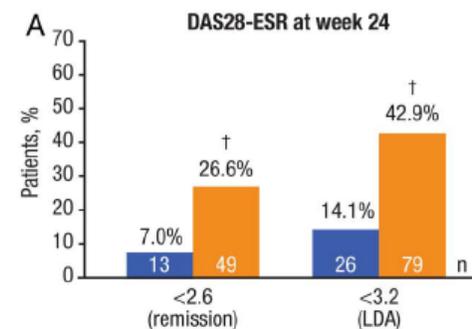


There is strong correlation between DAS score and ACR response that is different for IL-6 vs. other mechanisms

Burmester GR, et al. *Ann Rheum Dis* 2017;76:840–847. doi:10.1136/annrheumdis-2016-210310



Resulting in greater treatment differentiation based on DAS vs. ACR endpoints for IL-6 mechanism (purple dots)



Risk difference sarilumab (orange) vs. adalimumab (blue):

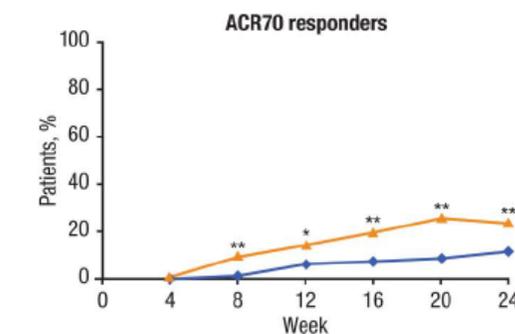
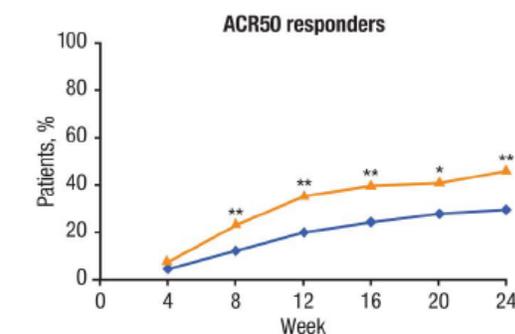
ACR20: 13.4 [3.8 - 23]

ACR50: 15.9 [6.2 - 26]

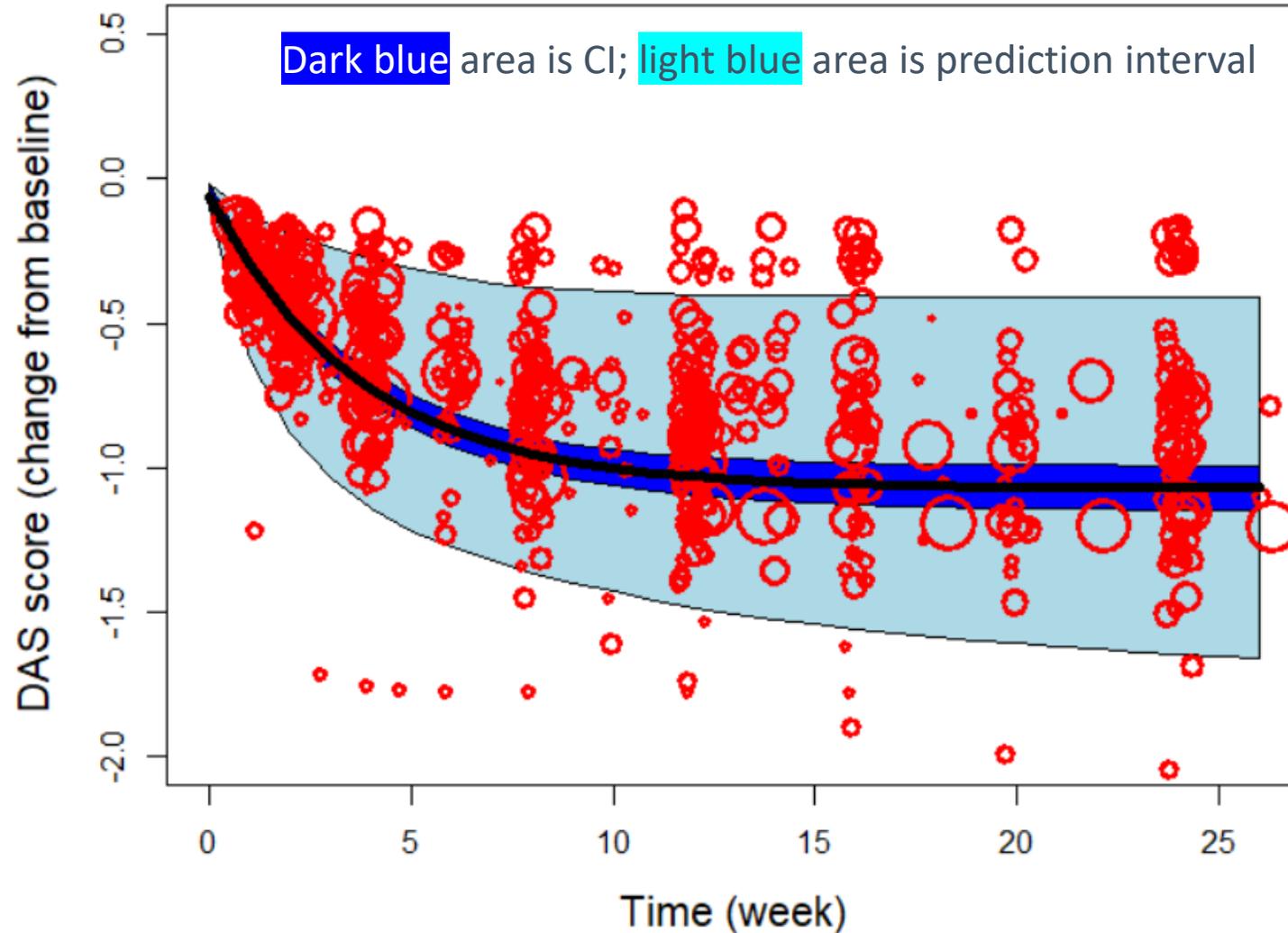
ACR70: 11.5 [3.8 - 19]

DAS LDA: 28.8 [20 - 38]

DAS rem: 19.6 [12 - 27]



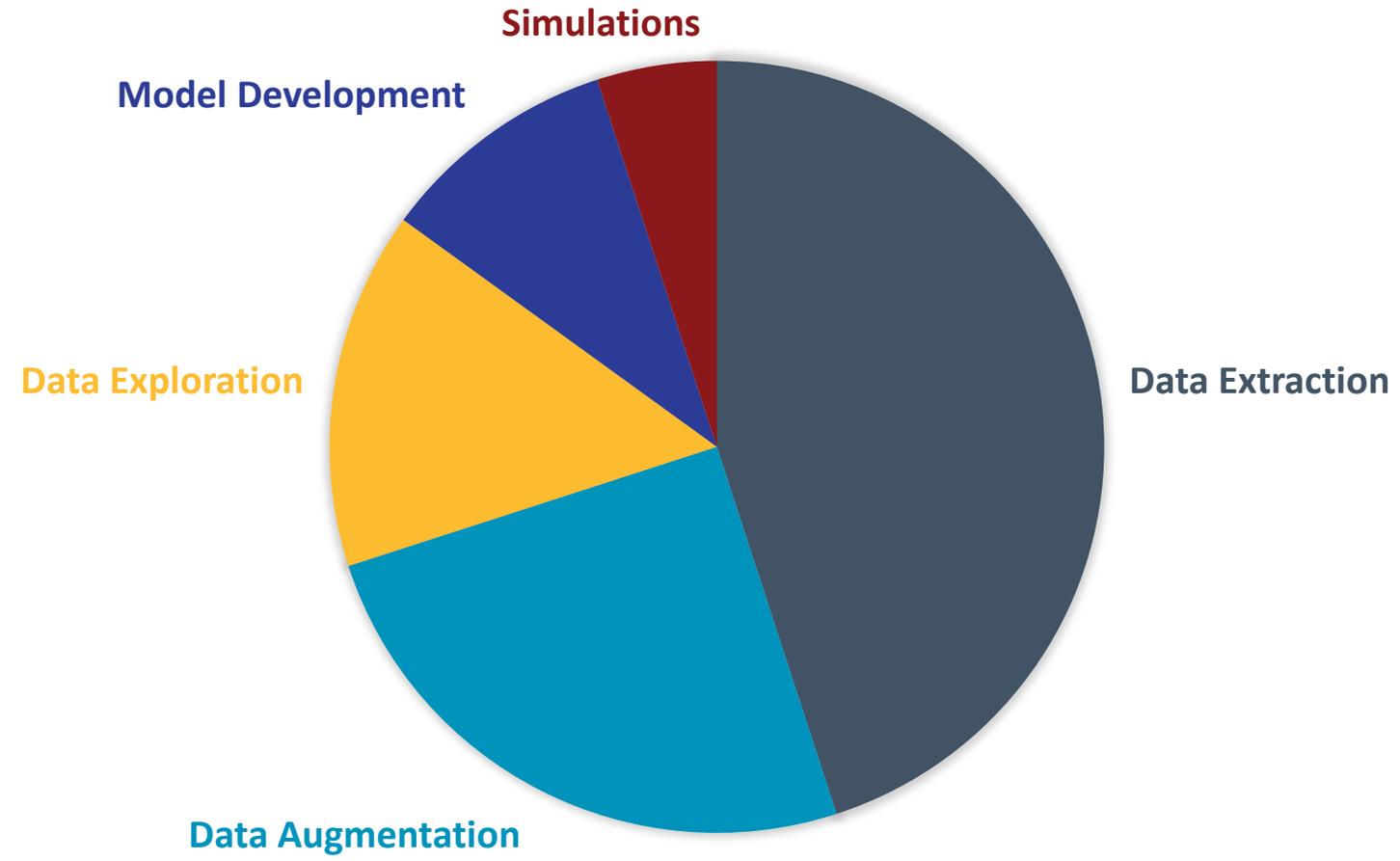
Synthetic Control Arm for Placebo in RA (DAS score)?



- Observed (red dots) vs. estimated (black line) time course of placebo DAS score change from baseline for RA patients.
- Whereas there is a very large amount of prior data for the placebo response from 179 studies (153 in RA; 26 in PsA)
- Whereas covariates such as baseline DAS score, DAS definition (ESR/CRP), Asian/non-Asian, and RA vs. psoriatic RA explain a significant amount of between trial heterogeneity in placebo response
- There is a large amount of unexplained heterogeneity indicated by PI
- Such that the historic placebo control is worth ~19 patients in a future study

To maximize the utility of MBMA, develop and maintain comparator models in advance of key decision points

TIME EXPENDITURE IN MBMA (APPROXIMATE)



MBMA provides a quantitative framework to leverage valuable external data into development and regulatory decisions

- Comparative safety and efficacy
 - There is a need to evaluate new treatment options against other existing or emerging treatment options (indirectly) for go/no-go decisions, dose selection, trial strategy
- Endpoint-to-Endpoint relationships
 - Biomarker to clinical endpoint predictions
 - Bridging across indications
- Create synthetic control arms
 - Adjusted for known and accounting for unknown factors that impact heterogeneity
- Leveraging existing information
 - Similar shape of dose response relationships of drugs within class
 - Similar impact of disease severity on treatment effect
- Optimize Trial design
 - Impact of trial design features on placebo, treatment effect, and variability

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All questions are welcome!

Model-Based Meta-Analysis (MBMA):

Optimizing Drug Development with Public Data and Predictive Models

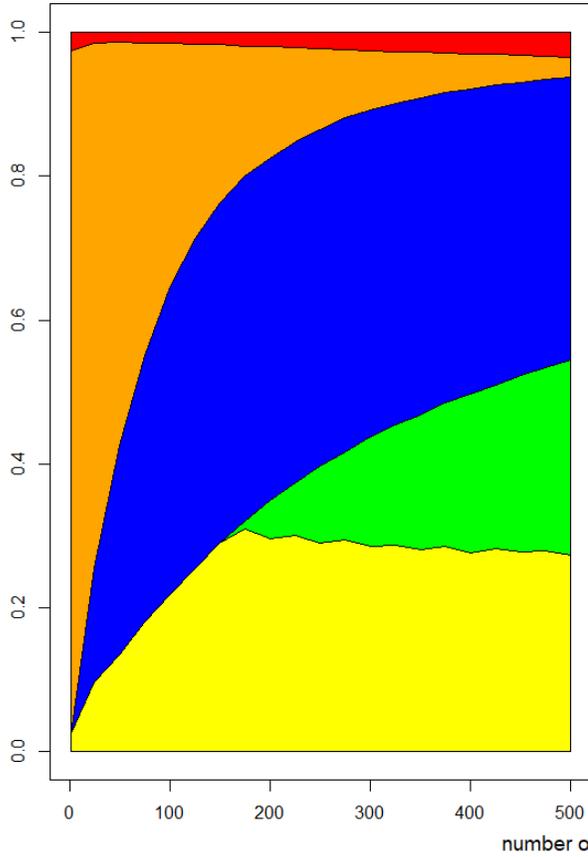
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Reading Materials

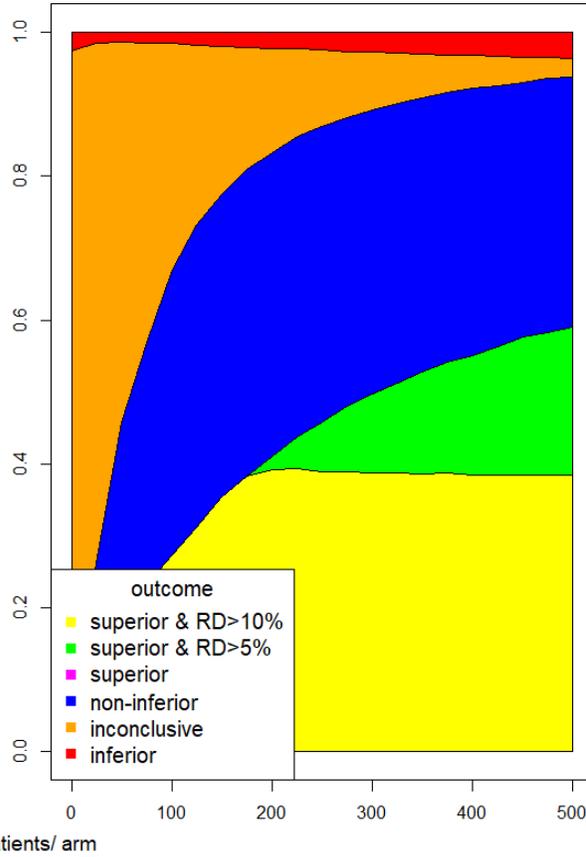
- Lalonde et al. Model-based drug development. Clin. Pharmacol. Ther. 82: 21-32 (2007).
- Milligan et al. Model-Based Drug Development: a rational approach to efficiently accelerate drug development. Clin Pharmacol Ther. 93(6): 502-14 (2013).
- Visser et al. Implementation of Quantitative and Systems Pharmacology in Large Pharma. CPT Pharmacometrics Syst. Pharmacol. 3, e142 (2014)
- Mandema et al. Model-based meta-analysis for comparative efficacy and safety: application in drug development and beyond. Clin Pharmacol Ther. 90(6): 766-9 (2011).
- Mandema et al. Therapeutic benefit of eletriptan compared to sumatriptan for the acute relief of migraine pain – results of a model-based meta-analysis that accounts for encapsulation. Cephalalgia 25: 715-25 (2005).
- Mandema et al. A dose-response meta-analysis for quantifying relative efficacy of biologics in rheumatoid arthritis. Clin Pharmacol Ther. 90: 828-35 (2011).
- Mandema et al. Therapeutic index of anticoagulants for prevention of venous thromboembolism following orthopedic surgery: a dose-response meta-analysis. Clin Pharmacol Ther. **90**: 820-7 (2011).
- Mandema et al. Time Course of Bone Mineral Density Changes with Denosumab Compared with Other Drugs in Postmenopausal Osteoporosis: A Dose-response Based Meta-analysis. J Clin Endocrinol Metab. 99: 3746-55 (2014).
- Checchio T et al. Quantitative Evaluations of Time-Course and Treatment Effects of Systemic Agents for Psoriasis: A Model Based Meta Analysis. Clin Pharmacol Ther (2017).

What is the optimal endpoint or comparator for phase III?

ACR20: baricitinib 4 mg/day vs. adalimumab



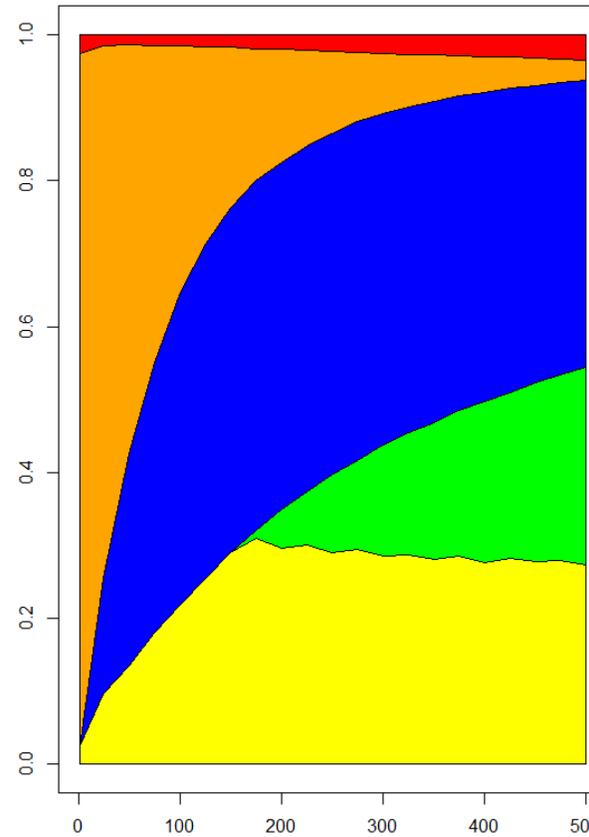
ACR50: baricitinib 4 mg/day vs. adalimumab



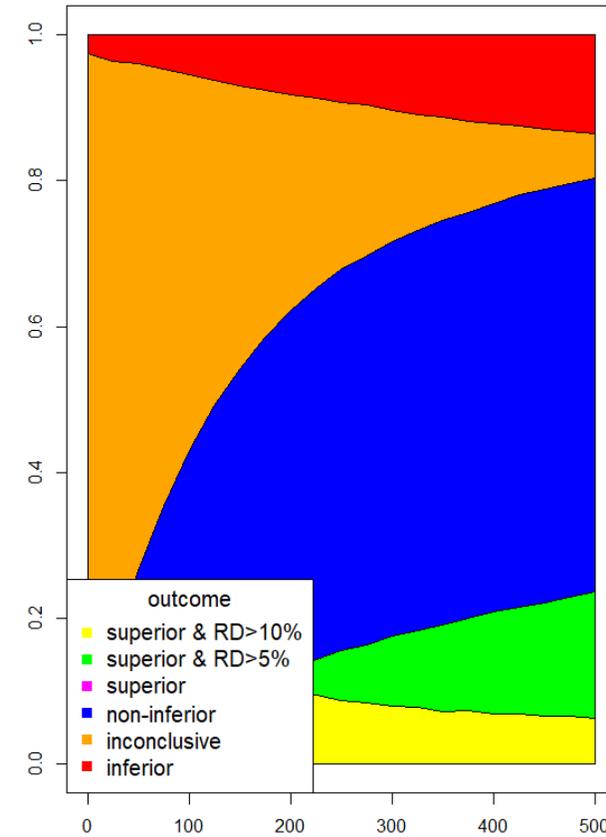
ACR20 or ACR50?

Adalimumab or certolizumab?

baricitinib 4 mg/day vs. adalimumab



baricitinib 4 mg/day vs. certolizumab



Probability of trial outcome conditional on effect size and probability of baricitinib to achieve effect size

