# BAYESIAN FRAMEWORKS FOR RARE DISEASE CLINICAL DEVELOPMENT PROGRAMS

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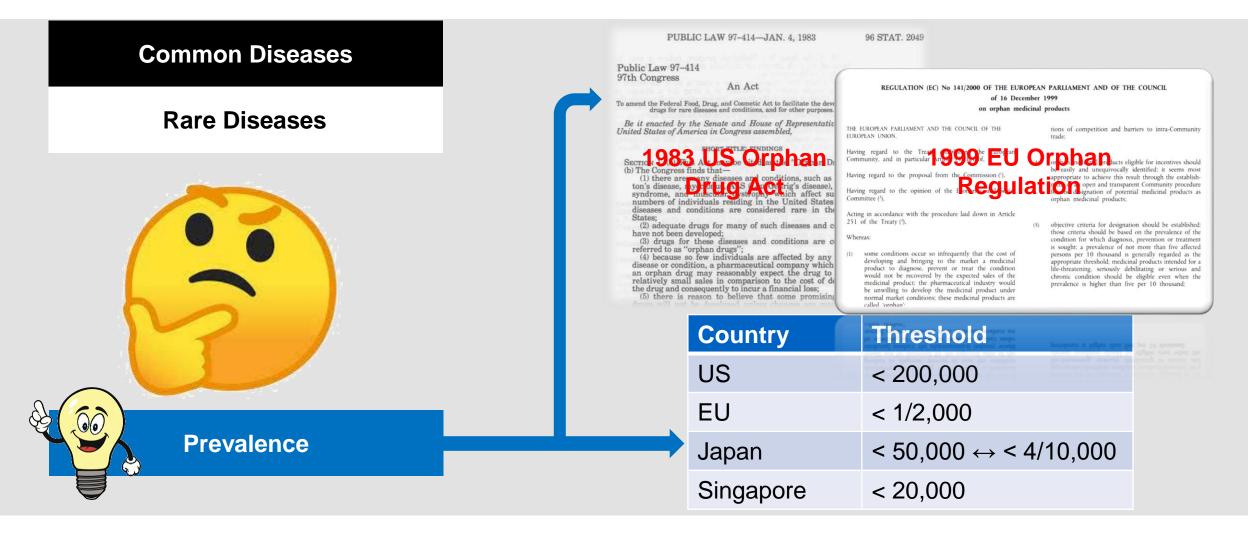


## OUTLINE

- Introduction
- Natural history studies
- Long-term safety evaluation and usage of Real-World Data in rare diseases
- Bayesian approaches in rare disease
  - Clinical trial designs
  - Bayesian approaches
- Case study
- Conclusions and future directions



#### **REGULATORY BACKGROUND OF RARE DISEASES**





#### **RARE DISEASES FACTS**

How many rare diseases?

- Orphanet: 6,000-7,000
- <u>– WHO: 5,000-8,000</u>
  - How many people have rare diseases?
    - US: 25-30 million
    - Worldwide: 300-400 million
      - What kind of diseases?
        - 80% genetic and chronic
          - How old are those patients?

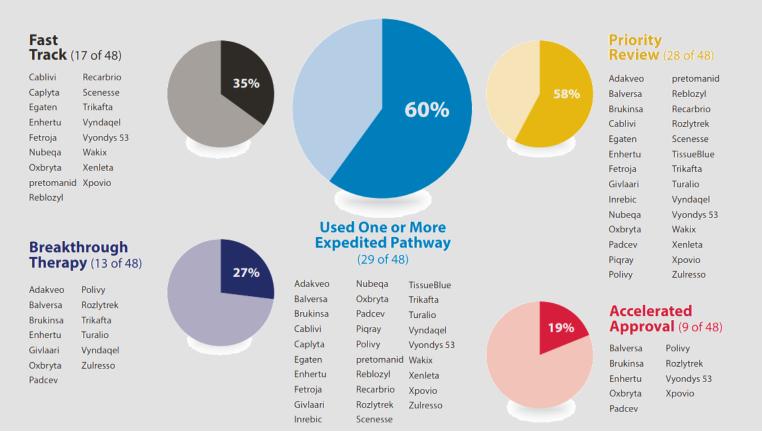
– 70% pediatric



## US FDA CDER NEW DRUG APPROVALS (2019)

#### 2019's Novel Drug Approvals

**Expedited Review Pathway Usage** 

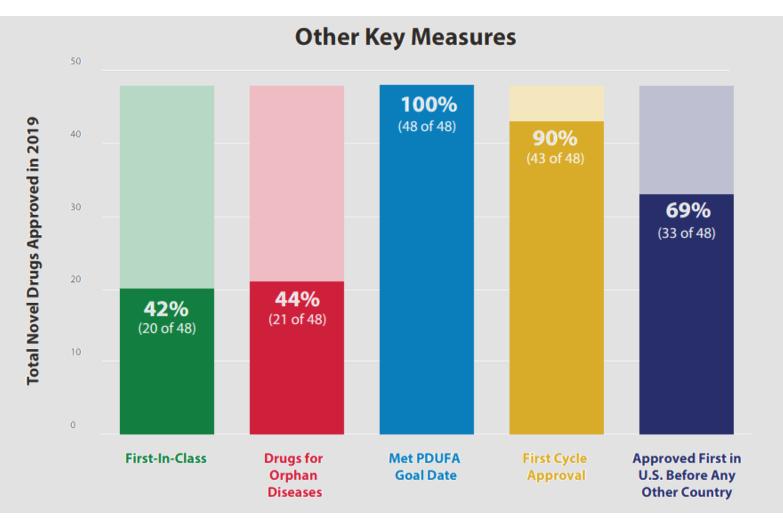


- Usage of expedited review pathway has been the majority of novel drug approvals
- Priority review consistently comprises the largest percentage

Source: CDER New Drug Therapy Approvals 2019 (https://www.fda.gov/media/134493/download)



#### US FDA CDER NEW DRUG APPROVALS (2019) – CONT.



30-40% are identified as firstin-class in the past few years

•

44% of approval for orphan disease is lower than 58% in 2018, but higher than 33% and 41% in 2017 and 2016, respectively

Source: Reports & Budgets | CDER – Reports and budgets from CDER offices and divisions (<u>https://www.fda.gov/about-fda/center-drug-evaluation-and-research-cder/reports-budgets-cder</u>)



## **NATURAL HISTORY STUDIES**

- Natural History: Course of development of a disease or condition without treatment
- FDA Guidance: Design and conduct natural history studies at the earliest stages of drug development

#### • Objectives of Natural History Studies:

- Define the disease population
- Understand and implement critical elements in clinical trial design
- Select clinical endpoints and develop sensitive and specific outcome measures
- Identify new or validate existing biomarkers
- Example: Myozyme/Lumizyme (alglucosidase alfa) Pompe disease

Source: Guidance for Industry – Rare Diseases: Common Issues in Drug Development (<u>https://www.fda.gov/media/119757/download</u>) CDER/CBER *Draft Revision 1 Jan 2019* 



## LONG-TERM SAFETY EVALUATION AND USAGE OF REAL-WORLD DATA IN RARE DISEASES

#### Post-Market Surveillance:

- Further confirm efficacy and its durability (including studies in minority populations)
- Detect safety signals
- Observe real-life patient usage of the products

#### • Rare Disease Program:

- Expedited approval pathway are often adopted
- Surrogate endpoints are used for initial approval
- Real-World Data vs. Natural History



## **BAYESIAN CLINICAL TRIAL DESIGNS**

- Clinical Development
  Program:
  - Pre-clinical studies
  - Clinical studies
  - Post-market studies
- Trial Aspects:
  - Dose selection
  - Go/No-go decisions
  - Subgroup identification

Adaptation strategies

## Platform Trial

- Multiple investigational treatments
- A shared control group
- Other Design Options
  - Cross-over
  - Single-arm



## **BAYESIAN APPROACHES LEVERAGING HISTORICAL DATA**

#### • Two-Step:

 $p(\theta \mid D_0) \propto L(\theta \mid D_0) p_0(\theta) \longrightarrow p(\theta \mid D) \propto L(\theta \mid D) p(\theta \mid D_0)$ 

#### • Power Prior:

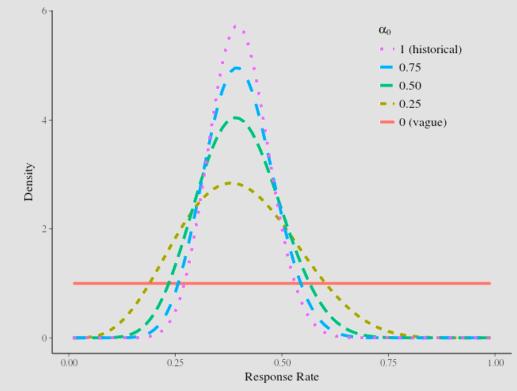
 $p_P(\theta \mid D_0) \propto L (\theta \mid D_0)^{\alpha_0} p_0(\theta)$ 

Joint Power Prior:

 $p_{\text{JPP}}(\theta, \alpha_0 \mid D_0) \propto L(\theta \mid D_0)^{\alpha_0} p_0(\theta) p(\alpha_0)$ 

• Modified Power Prior:

 $p_{\rm MPP}(\theta, \alpha_0 \mid D_0) \propto \frac{L(\theta \mid D_0)^{\alpha_0} p_0(\theta)}{\int L(\theta \mid D_0)^{\alpha_0} p_0(\theta) d\theta} p(\alpha_0)$ 





#### **ROBUST MIXTURE PRIOR**

 $p_{\text{RMP}}(\theta, w \mid D_0) \propto (1 - w) p(\theta \mid D_0) + w p_r(\theta).$   $\downarrow$   $Pr(M_r \mid D) = \frac{p(D \mid M_r)w}{p(D \mid M_r)w + p(D \mid M_h)(1 - w)}$ 

 $M_h = p(\theta/D_0)$ : Historical data prior  $M_r = p_r(\theta)$ : Robust prior (usually weakly informative)

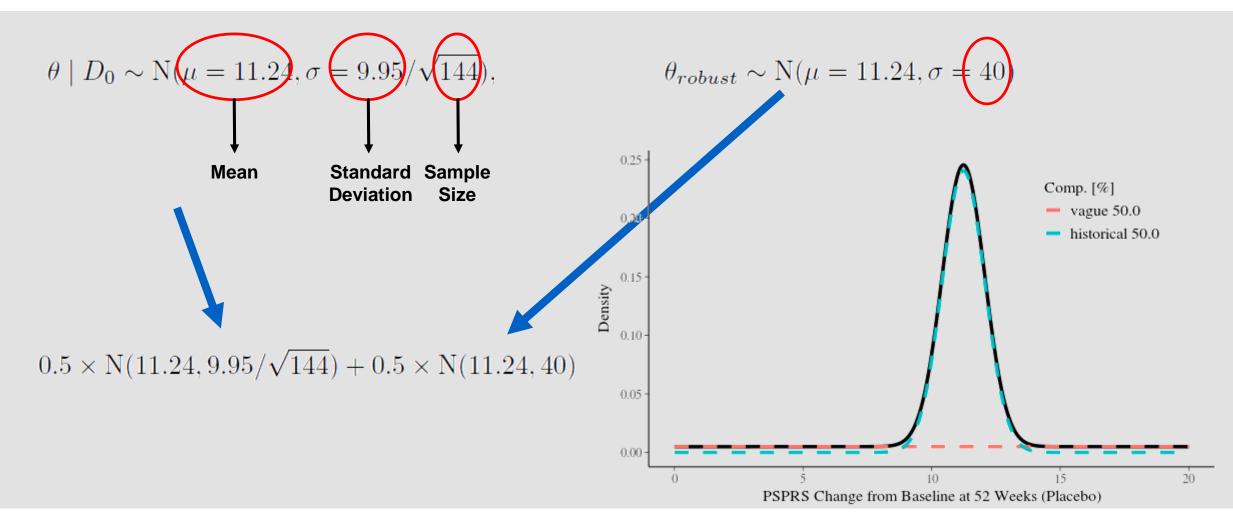


## **CASE STUDY**

- Progressive Supranuclear Palsy (PSP): a rare neurodegenerative disorder characterized by the accumulation of aggregates of tau protein in the brain
- Prevalence (Orphanet): 1/16,600\* 328.2 mil = 19,771 < 200,000
- **Historical Placebo Data:** Two double-blind, randomized, placebo-control trials (neither met 52-week primary outcome)
  - Phase II/III: Davunetide (30 mg) vs. Placebo
  - Phase II: Tideglusib (600 and 800 mg) vs. Placebo
- Primary Endpoint: PSP-Rating Scale (PSPRS) @ Week 52



#### **ROBUST MIXTURE PRIOR CONSTRUCTION**



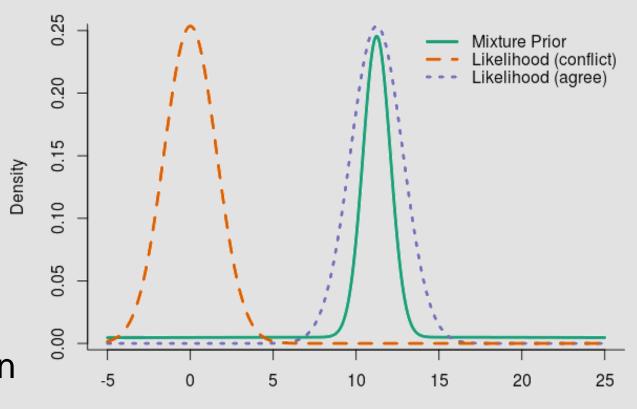


#### **DESIGNING A NEW TRIAL**

- N=40 Placebo Patients
- Prior-Data Conflict

Centered at a mean PSPRS change from baseline at 52 weeks of 0

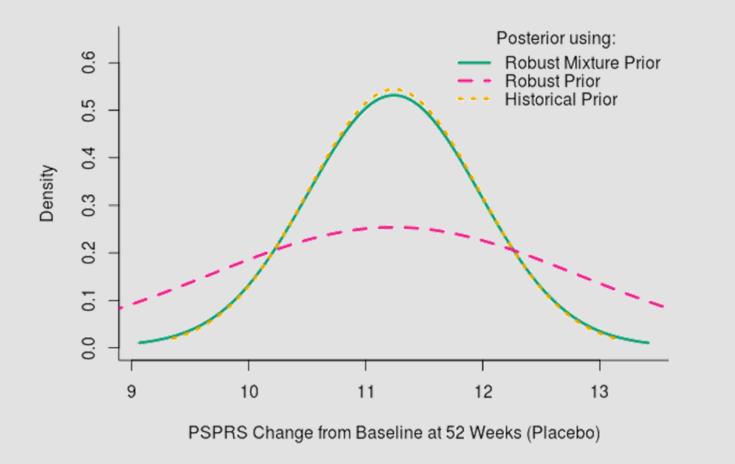
No Prior-Data Conflict
 Centered at the historical mean of 11.24



PSPRS Change from Baseline at 52 Weeks (Placebo)



## **POSTERIOR RESULTS (NO PRIOR-DATA CONFLICT)**

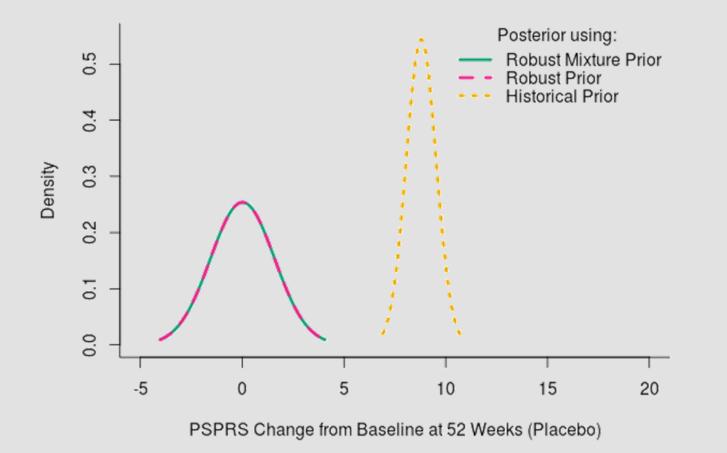


Using the robust mixture is nearly equivalent to using the historical prior alone, thus getting the maximal benefit from the historical information

 Using a vague prior, the posterior has much greater uncertainty due to the small sample size



## **POSTERIOR RESULTS (PRIOR-DATA CONFLICT)**



- Using the robust mixture is equivalent to using the robust prior alone
- Using the historical prior, the posterior is actually closer to the historical data due to the small sample size compared to the historical data



## **CONCLUSIONS AND FUTURE DIRECTIONS**

- Many current rare disease clinical programs either rely on traditional trial designs with few variations, or are purely based on clinical judgment
- The Scottish Medicines Consortium (SMC) has introduced an ultra-orphan definition associated with a new approach to decision-making on such medicines in 2018

## US FDA

- Requested additional budget for potential ultra-orphan incentives
- Complex Innovative Trial Design Pilot Program
- Consider robust Bayesian approaches



## ACKNOWLEDGEMENT

#### Co-authors:

- Forrest Williamson, Eli Lilly and Company
- Bradley P. Carlin, Counterpoint Statistical Consulting, LLC



Chapter 12: Bayesian Frameworks for Rare Disease Clinical Development Programs See also

- Chapter 6: Use of Historical Data
  Beat Neuenschwander and Heinz Schmidli
- Chapter 13: Bayesian Hierarchical Models for Data Extrapolation and Analysis in Pediatric Disease Clinical Trials *Cynthia Basu and Bradley P. Carlin*

