



COLLABORATIVE  
**TRAJECTORY  
ANALYSIS  
PROJECT**

## **Collaborative Learning from Patient Data in Rare Disease**

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# Collaborative Trajectory Analysis Project - cTAP

- Pre-Competitive coalition
- Pan-stakeholder
- Global
- Multi-registry, multiple trial
- Collaborative analytics



- “Smarter” Trials
- Current focus on Duchenne muscular dystrophy

# Disclosures

- The funds for cTAP are contributed by member drug companies and patient foundations
  - Susan Ward receives collaboration management fees
  - The Analysis Group, at which James Signorovitch is a partner, receives fees for analytic services
- I am not a statistician!

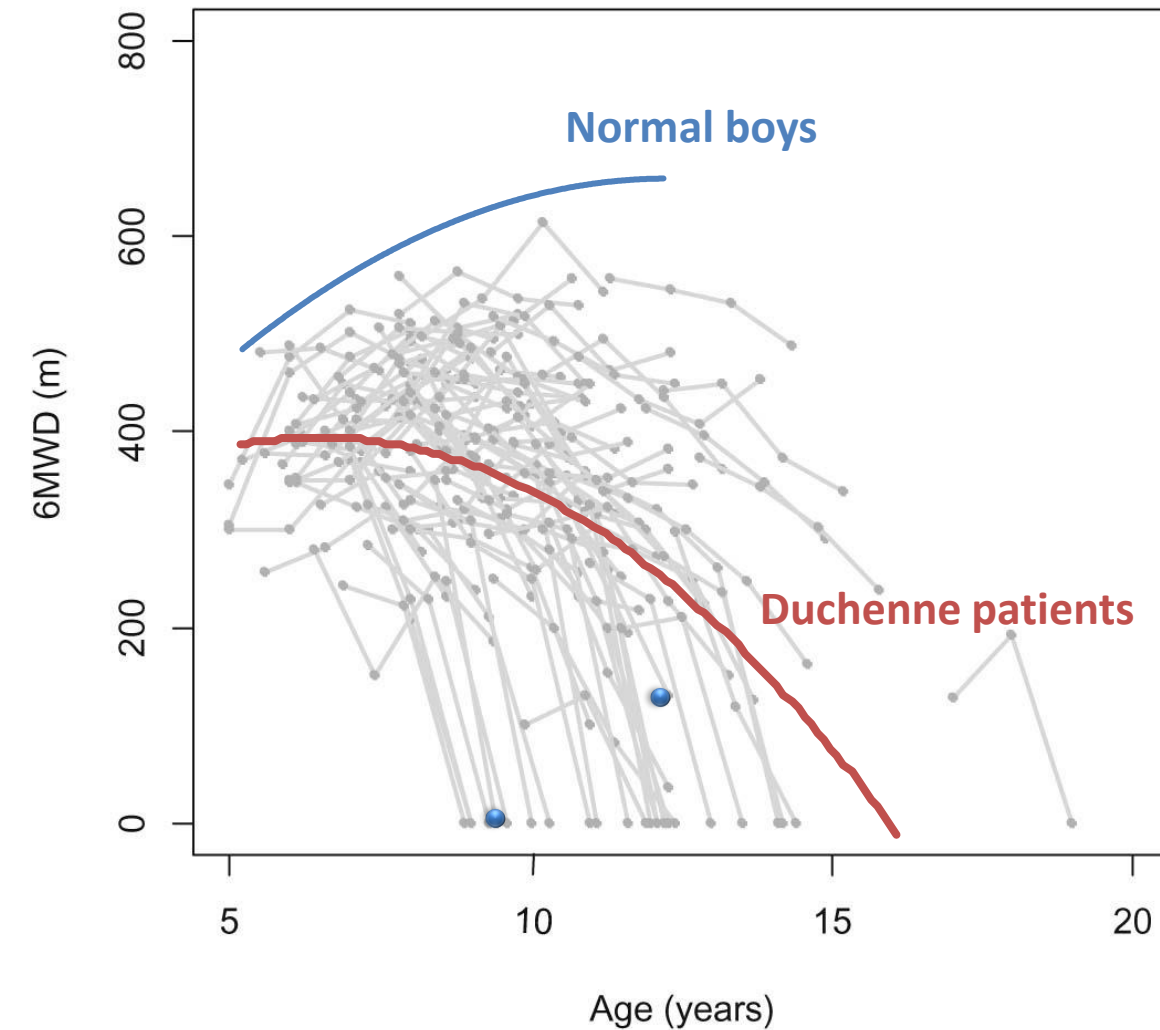
# Duchenne muscular dystrophy

*“A thousand little deaths”*



*Prescient capture of  
natural history*

# Ambulatory Function in Duchenne – complicated by maturation



- Symptom-prompted diagnosis of DMD occurs between 3-5 yrs of age
- Average trajectory of ambulatory function in DMD is progressive decline from ~ 7 yrs
- Normal boys approach peak ambulatory performance at ~10 yrs of age
- Heterogeneity in observed functional decline in DMD is likely a balance between maturation and disease progression

# Impetus – iterative clinical failures

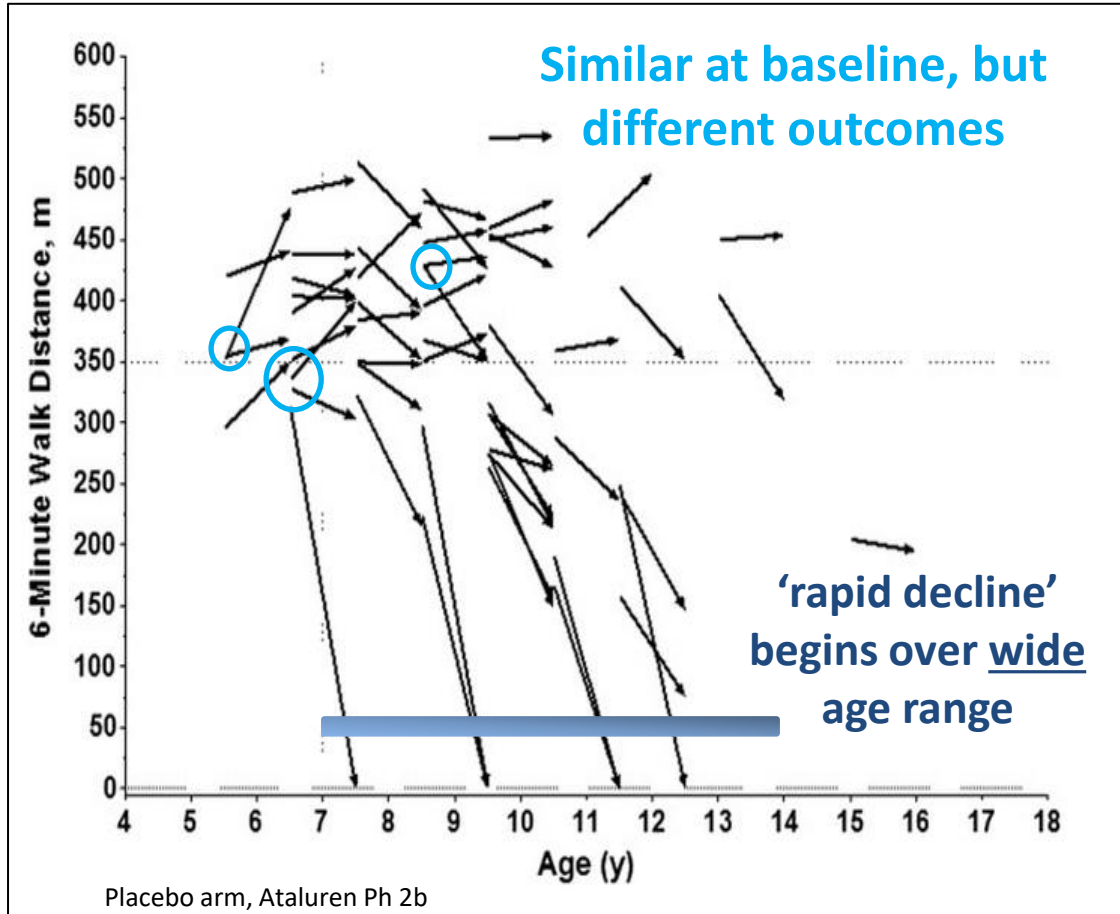
Placebo Arm Results, Ambulant at Baseline

PIVOTAL TRIAL	6MWD		n	Met endpoint?
	1yr change	SD		
DEMAND III	-53	81	61	no
PTC 007	-44	88	57	no
TADALAFIL	-51	100	116	no
PTC 020	-61	99	115	no

*higher than **anticipated** variance*

- *Why was high variance not anticipated?*
- *Do we have failed Drugs? or failed trials?*

# Dominant driver of *failed trials* in Duchenne is ..... heterogeneity of disease progression



- Similar profile of heterogeneity also seen in natural history studies
- Data held by leading clinicians in large clinicals and in curated clinical registries
- Limited data-sharing by clinical centers, ZERO sharing of clinical registry data

*“We need to design trials that test efficacy of a drug, not how much we don’t know about natural history”*

Exec Dir, Clinical Development, Pharma cTAP member  
2015

## *Explain - and account for - heterogeneity in disease progression*

- Collaborate to learn from patient data
- Bridge gaps in analytic approaches in DMD
- Create tools and insights for drug development
- Share broadly with the entire DMD community
- Deliver near-term impact for trials

### ***‘Smarter’ Trials***

- *Fully powered*
- *Smaller*
- *Leverage NH*

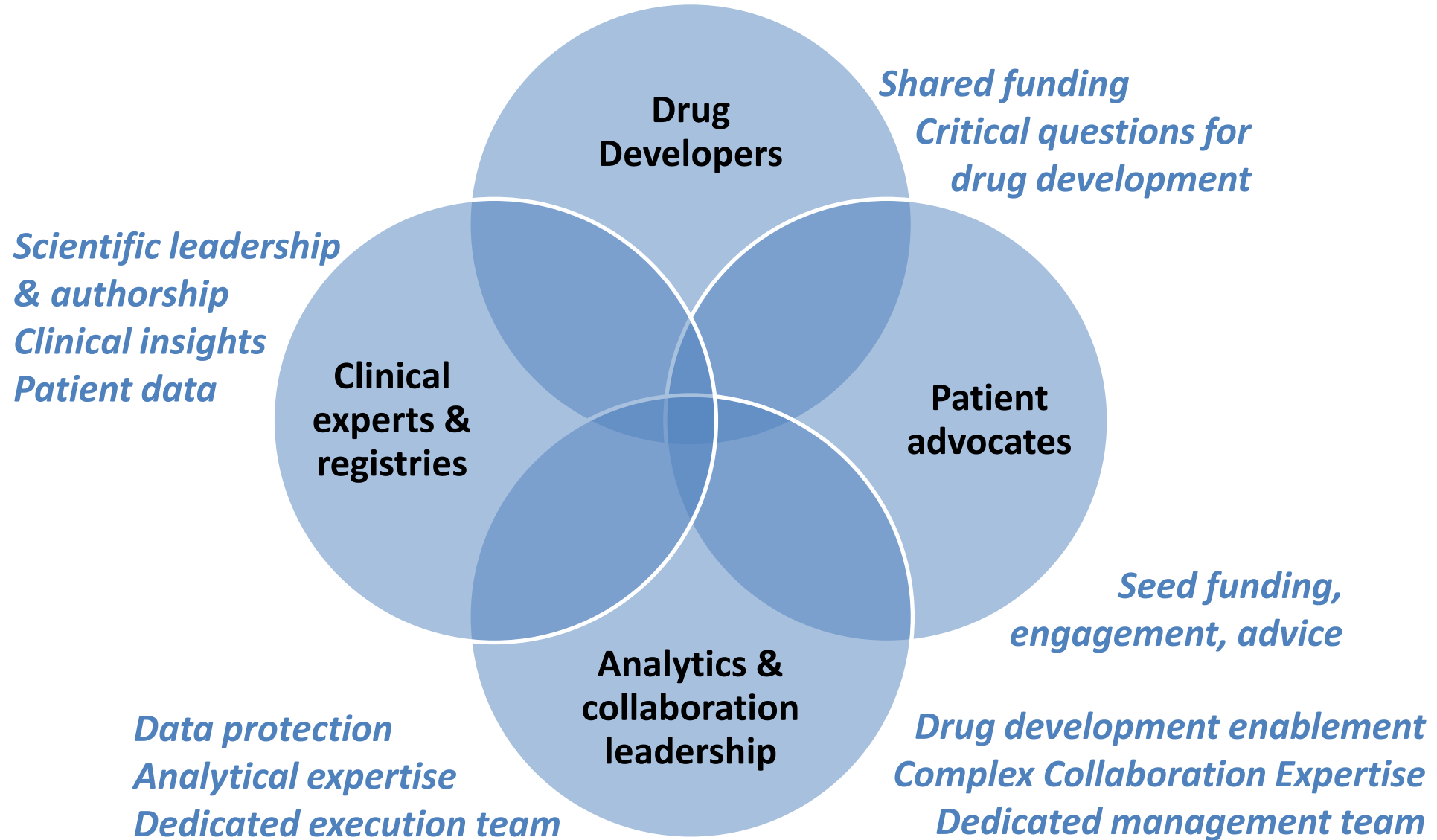
### ***‘Better’ Evidence***

- *Real World data*
- *Tie near-term to downstream*

***Therapies to Patients sooner***



# How We Collaborate



# cTAP Members and Collaborators

## Clinical experts and registries

Eugenio Mercuri



Nathalie Goemans



Francesco Muntoni



Brenda Wong



Hank Meyer



Craig McDonald



Krista Vandenberg



## Therapy Developers



## Patient Groups



## Collaboration Lead

Susan J. Ward, PhD



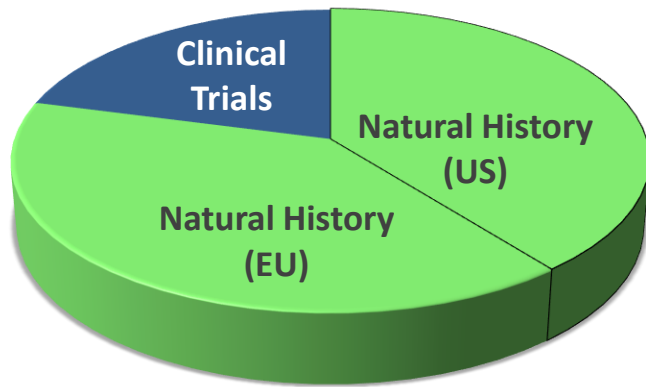
## Data Science Lead

James Signorovitch, PhD



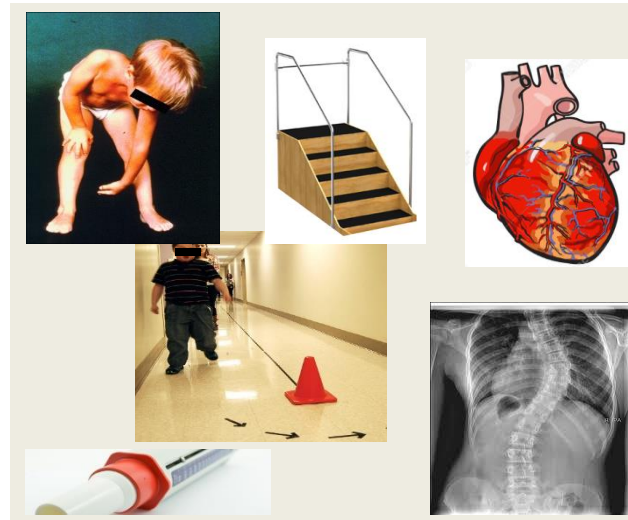
# Patient data accessed by cTAP

>2,300  
boys



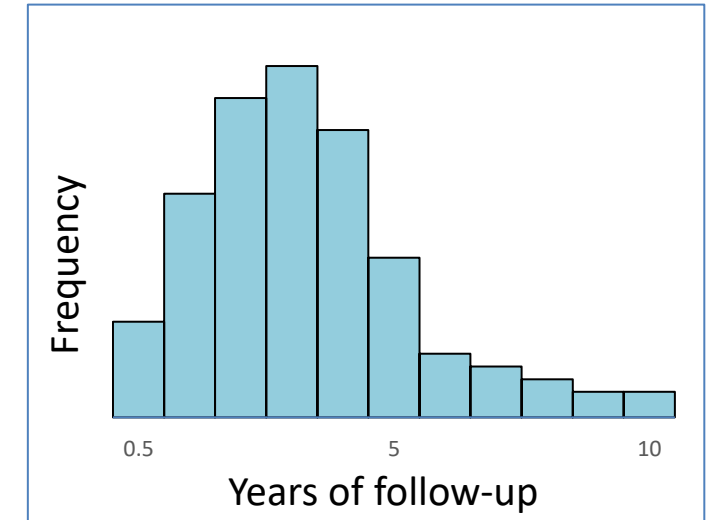
**LARGE &  
REPRESENTATIVE**

>15,000  
clinic visits



**RELEVANT &  
COMPREHENSIVE**

>1,000  
patient-years



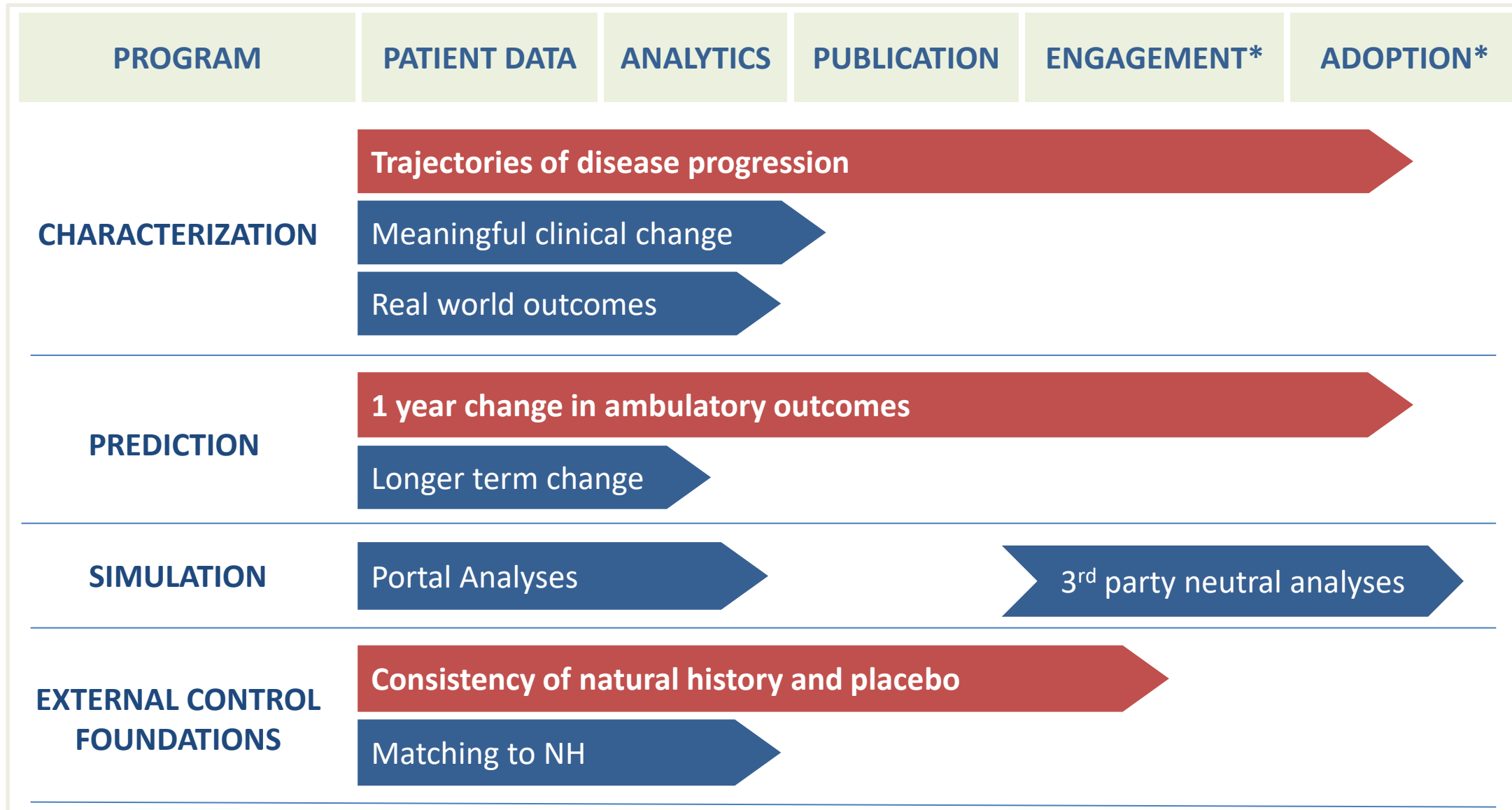
**DISEASE PROGRESSION**

# How we operate scientifically

- Governed by a Joint Steering Committee
  - Prioritize goals based on most critical needs for drug development and evaluation
  - Collaborate on research plans, interpretation, publications
  - Analytic results – not raw data -- are shared within cTAP
  - Analyses conducted per SAP by Analysis Group/other
  - Findings validated (or not) across data sources
- *Neutral 3<sup>rd</sup> party*
    - *Equitable for drug companies*
    - *Objective process (no cherries)*
    - *Equal benefit, equals costs*
  - *Impact -Focused , Nimble*
    - *Rigor, High Quality, objective*
    - *Shared knowledge, shared problem-solving*

**Builds confidence and trust**

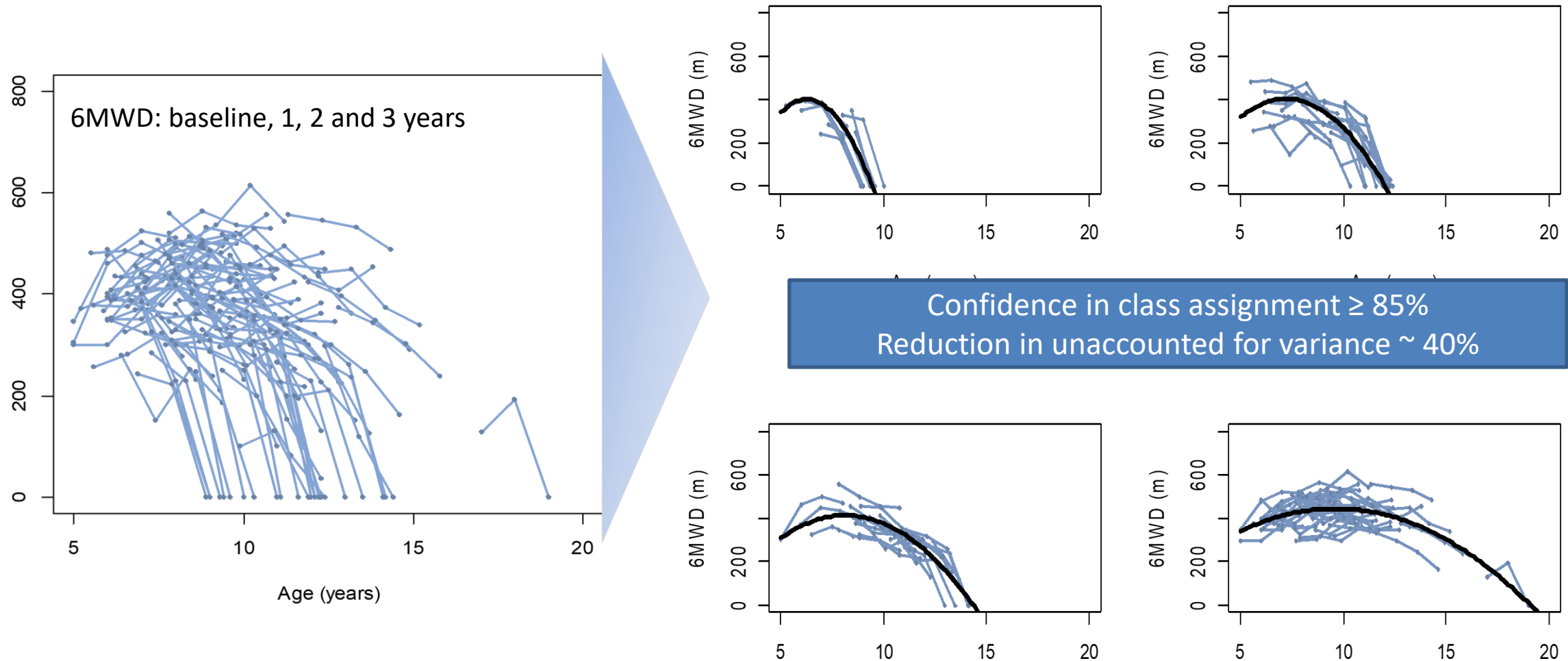
# cTAP Pipeline (leading outcome measure)



\*With primary target group – academic, regulators, payors/health authorities – dependent on program

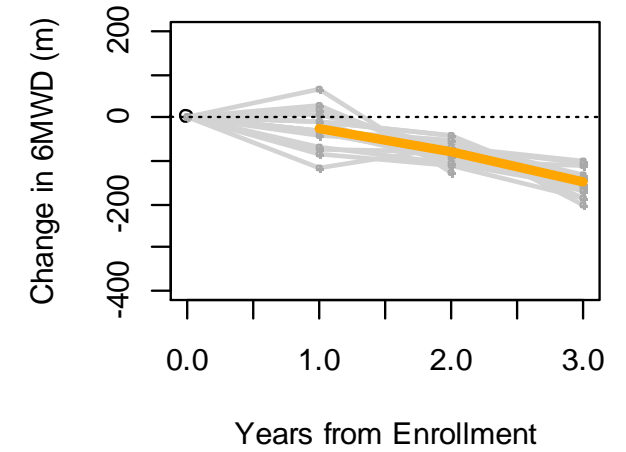
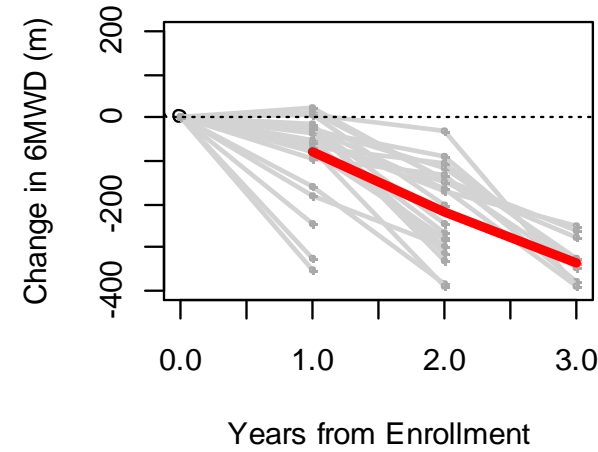
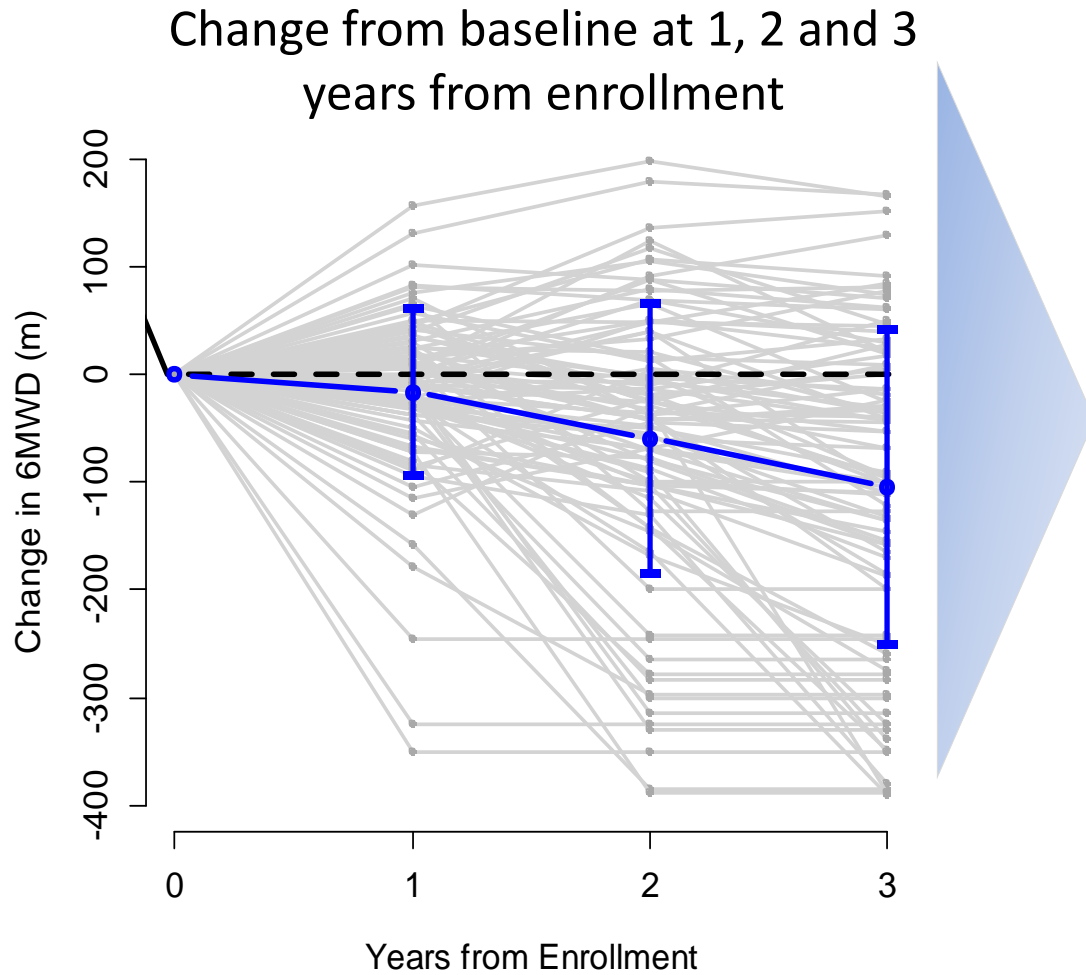
# Latent Class Growth modeling of Longitudinal Trajectories of Disease Progression in DMD

# Clustering of Longitudinal Trajectories of DMD natural history

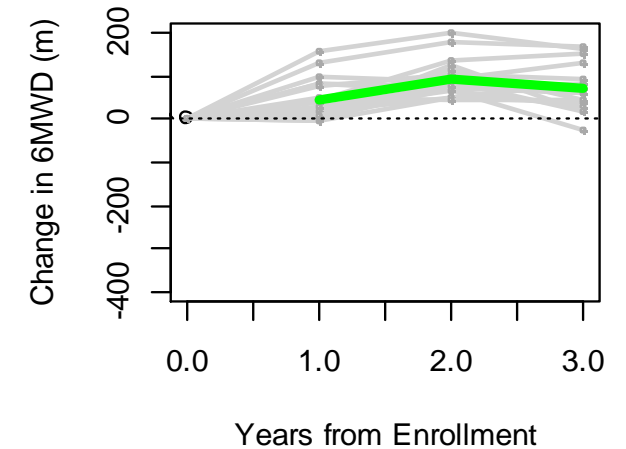
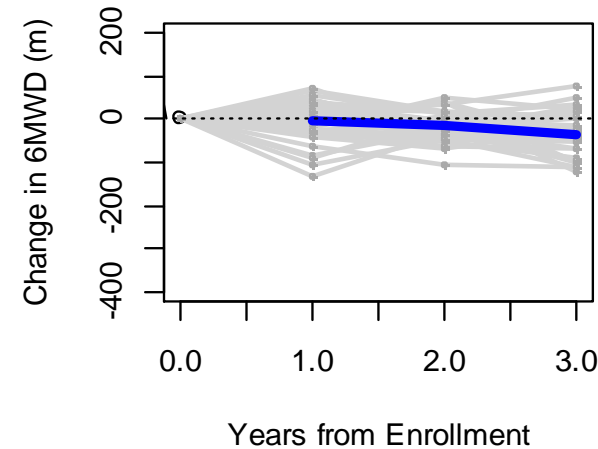


**DMD progresses at different rates in different groups of patients**

# Clustering of Longitudinal Trajectories of time from enrollment



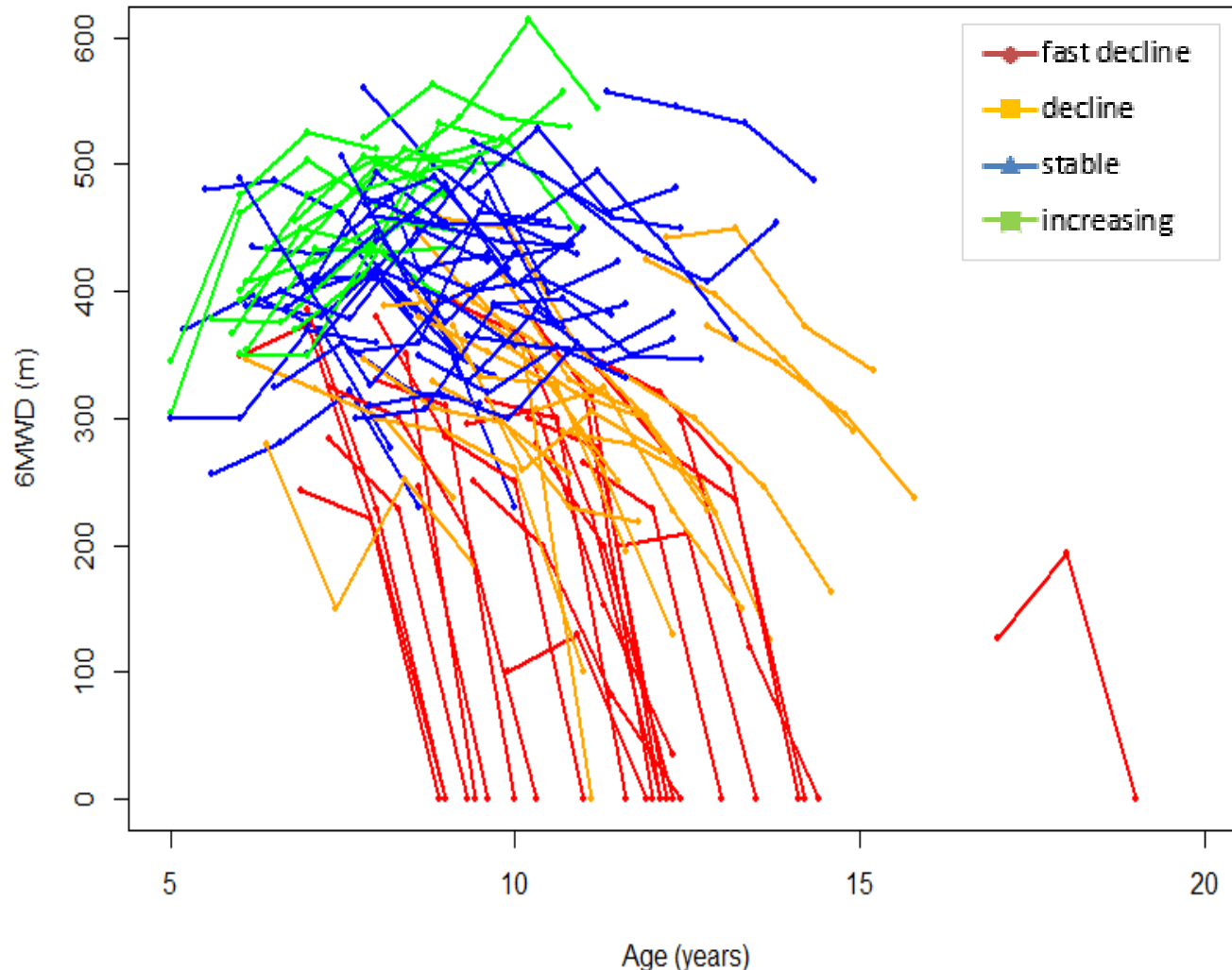
Confidence in class assignment  $\geq 89\%$   
Reduction in unaccounted for variance  $\sim 40\%$



**Different Patients may be in different Phases of Disease progression**



# What we've learned



## Clustering of Disease Progression

- suggests underlying structure
- is consistent across databases
- is seen with other (ambulatory) outcome measures
- is concordant across outcomes

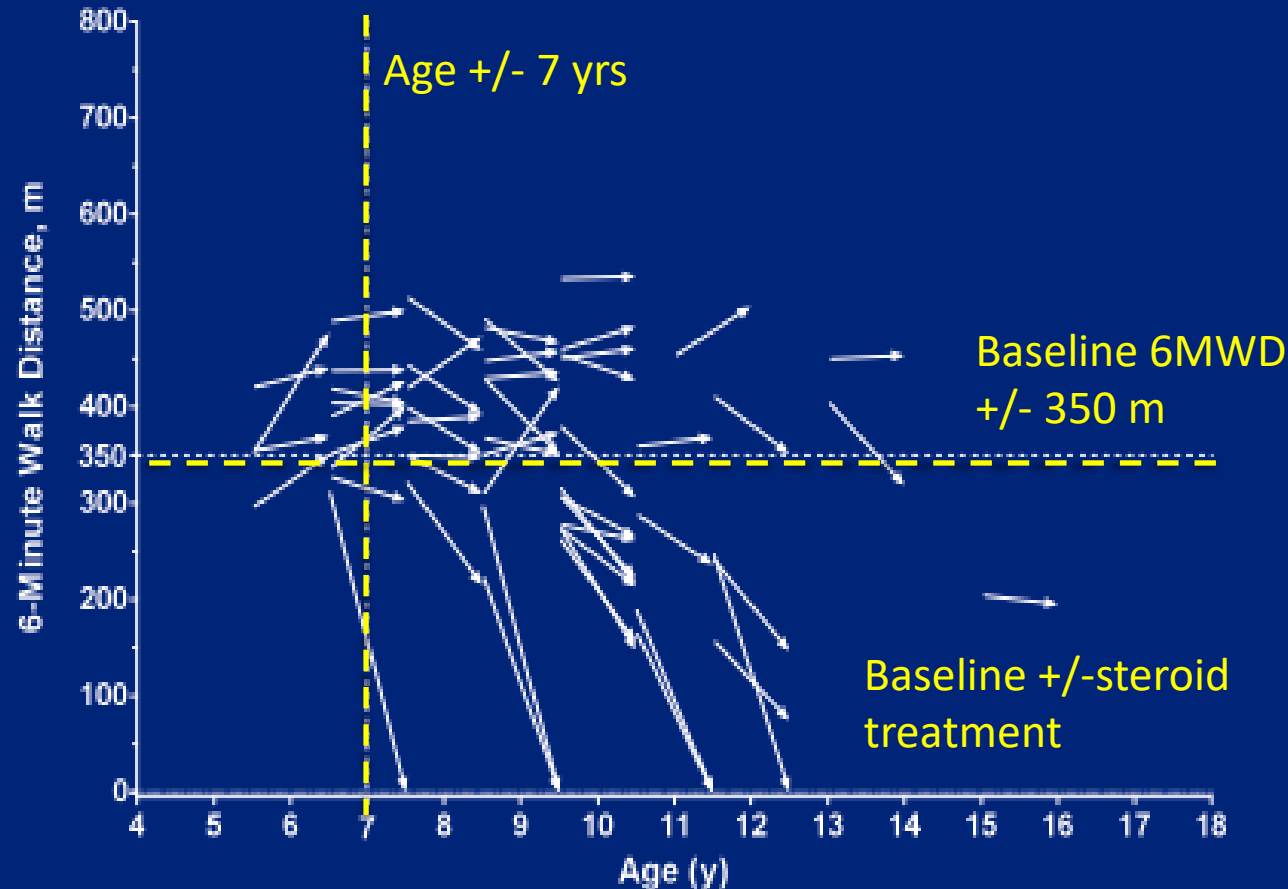
## “Value” of characterisation studies

- Resonates with non-statisticians
- “Makes sense”
  - maturation+disease
- Framework for associations

# Prognostic Factors

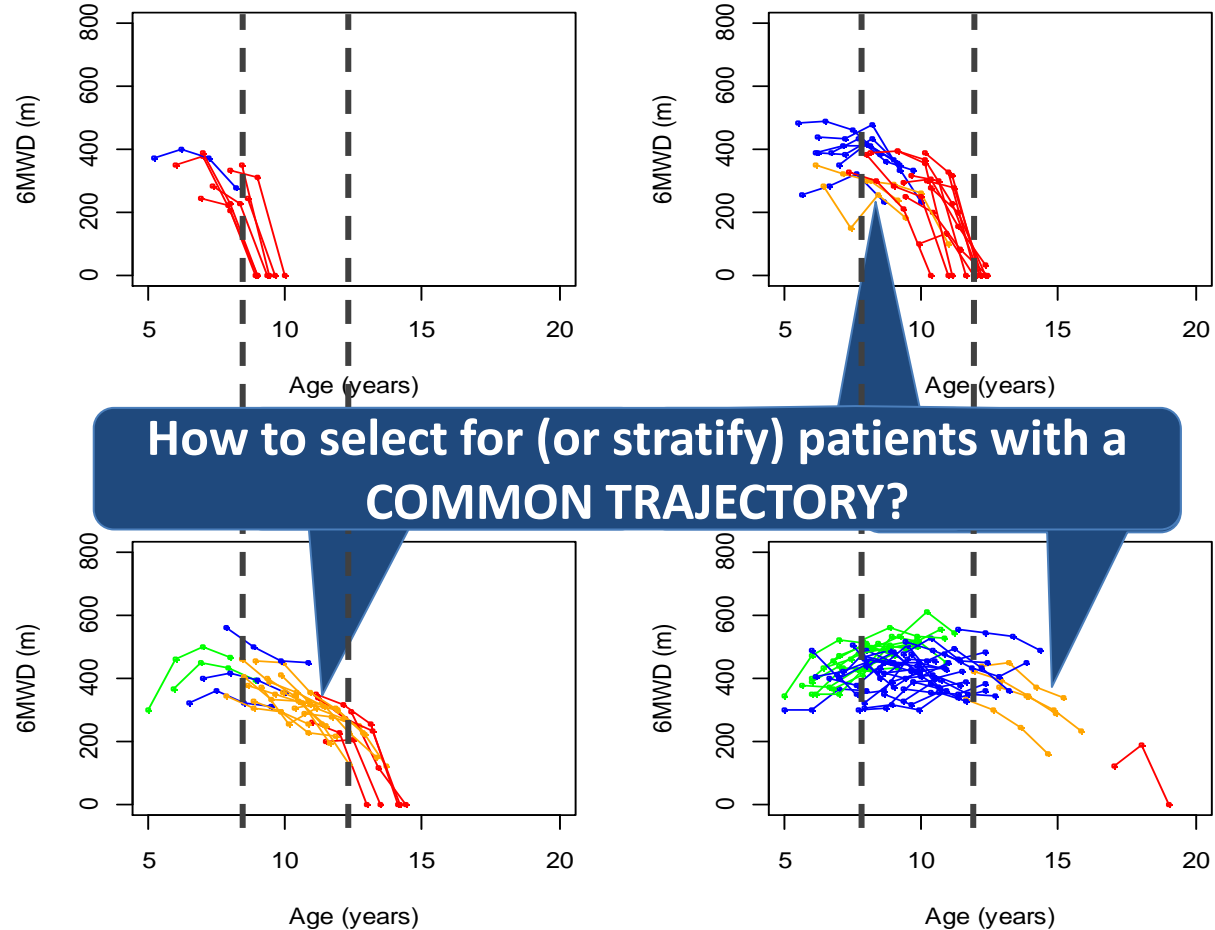
# Prognostic factors - Benchmark in 2015

Ataluren placebo – n= 57



- Developed post-hoc
- Determined by eye, not by statistics
- Included in MMRM analyses of trials
- Adopted widely to craft *inclusion criteria*

# The challenge of longitudinal heterogeneity in trial design



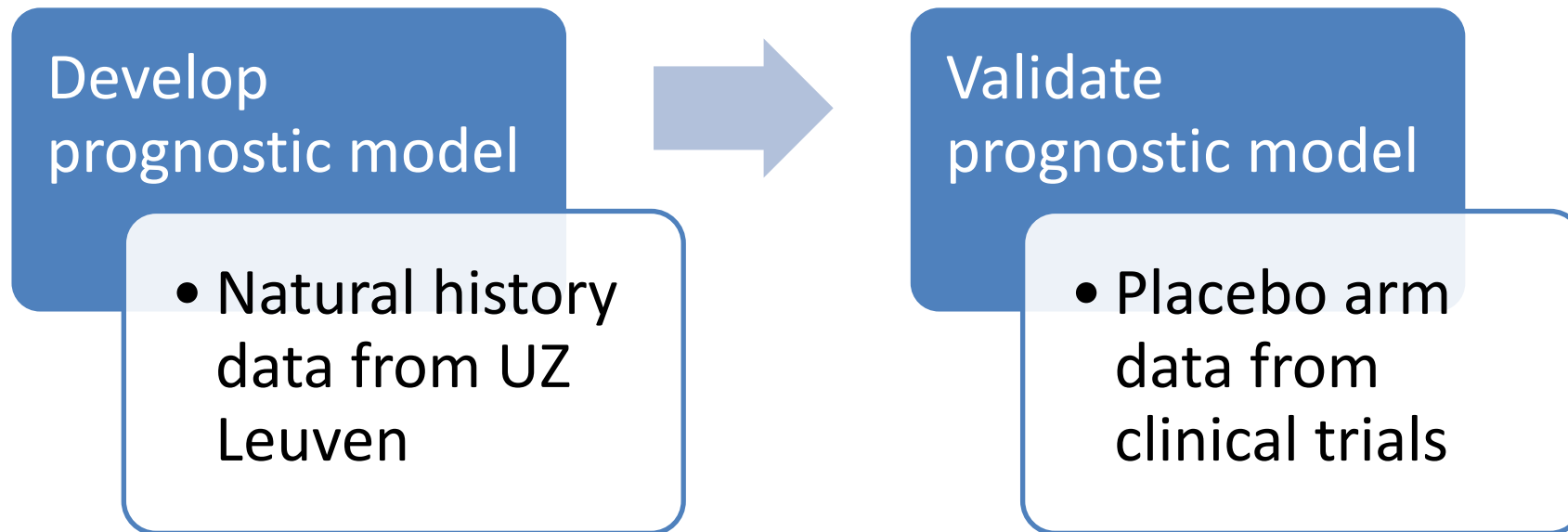
inclusion/exclusion  
criteria can narrow eligibility  
without reducing  
heterogeneity of sample

*Premise: More accurate pre-defined stratification => improve power*

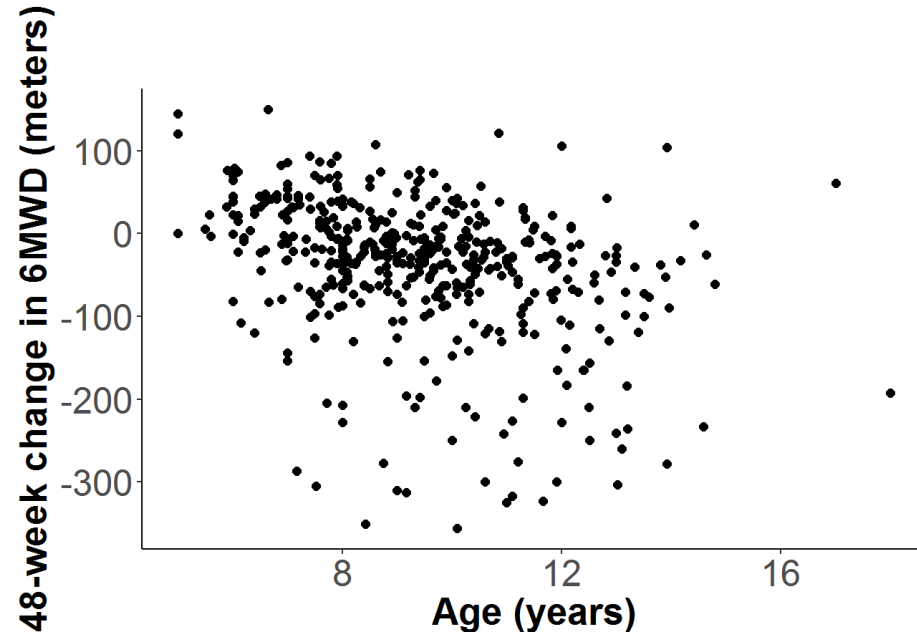
# Objectives

- How informative are the conventional baseline factors (age, 6MWD and steroid use) for predicting 1-year change in 6MWD?
- Can prognostic accuracy be improved?
- What are the most important prognostic factors?

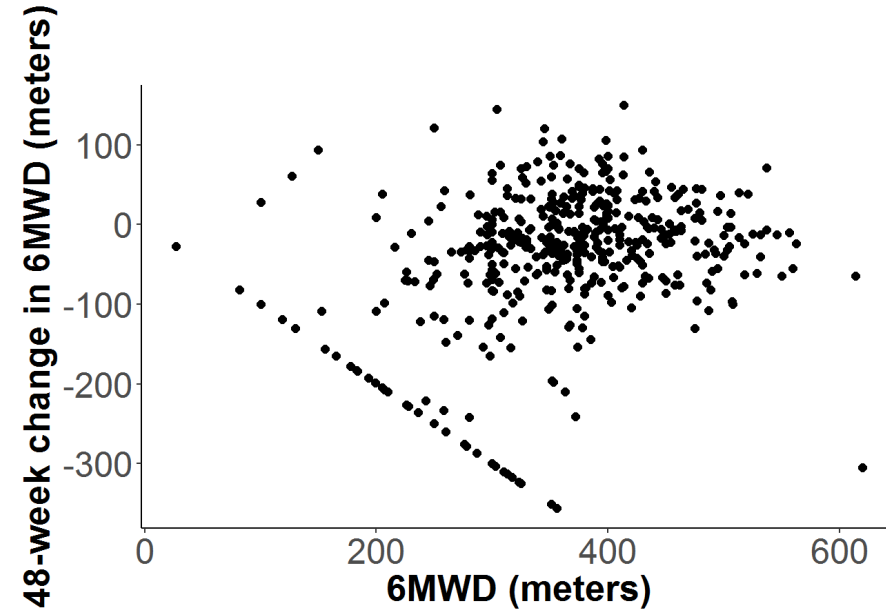
# Study Design



# Age and baseline 6MWD are each only weakly correlated with $\Delta 6MWD$



Correlation Coefficient = -0.35



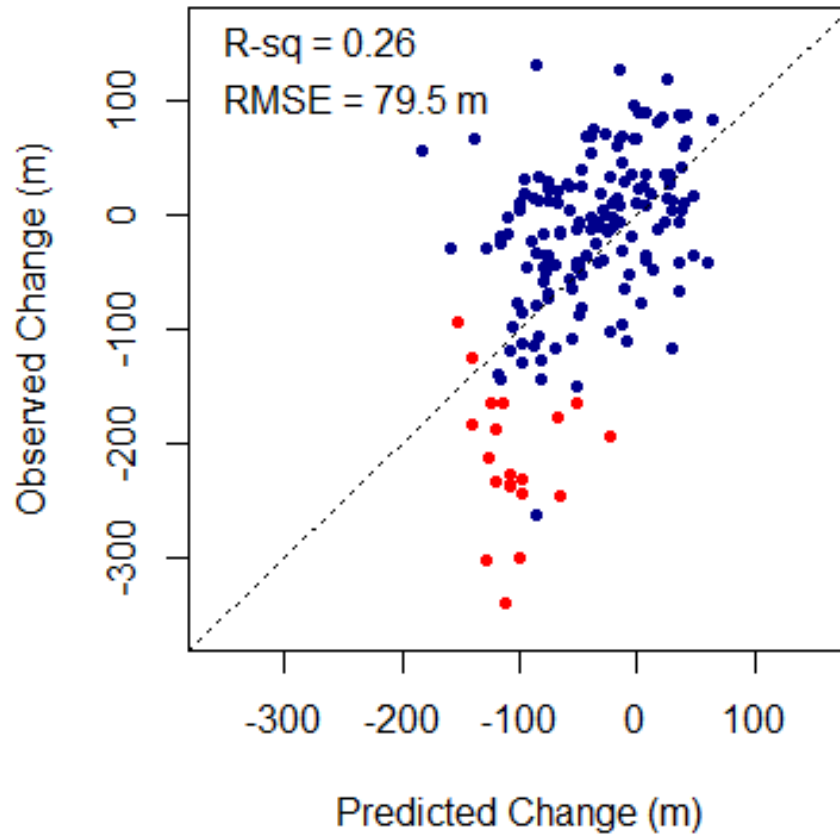
Correlation Coefficient = 0.25

*Data Pooled from the Italian Group, UZ Leuven and Lilly Placebo arm.*

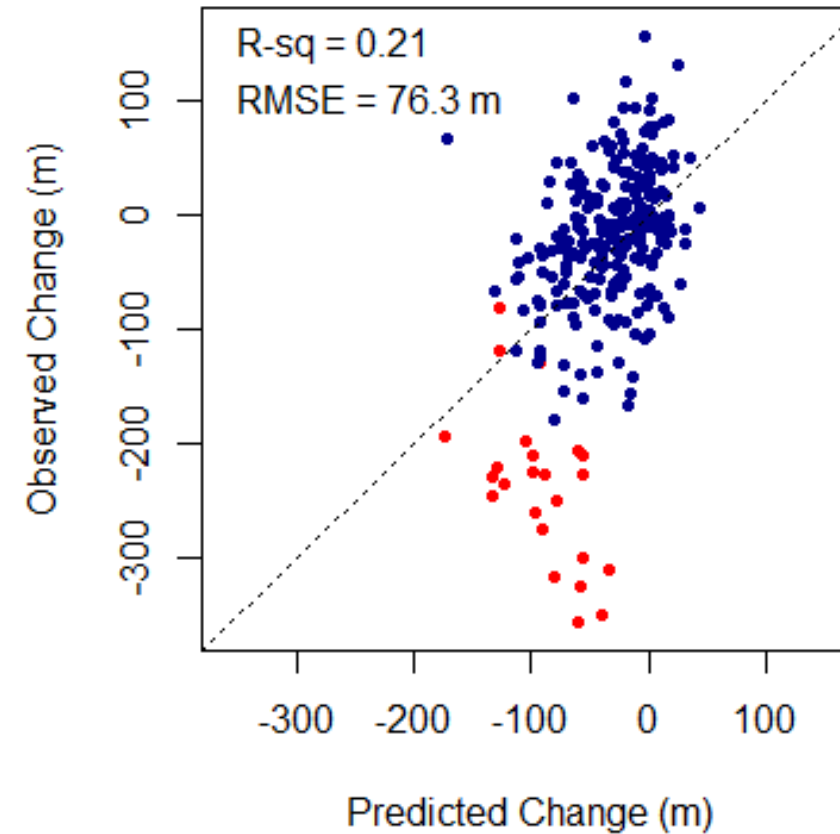
cTAP\_ASA.FDA\_091318

# Conventional Prognostic Factors account for only one quarter of observed variance in 1 yr $\Delta$ 6MWD

Leuven: age + 6MWD + steroids

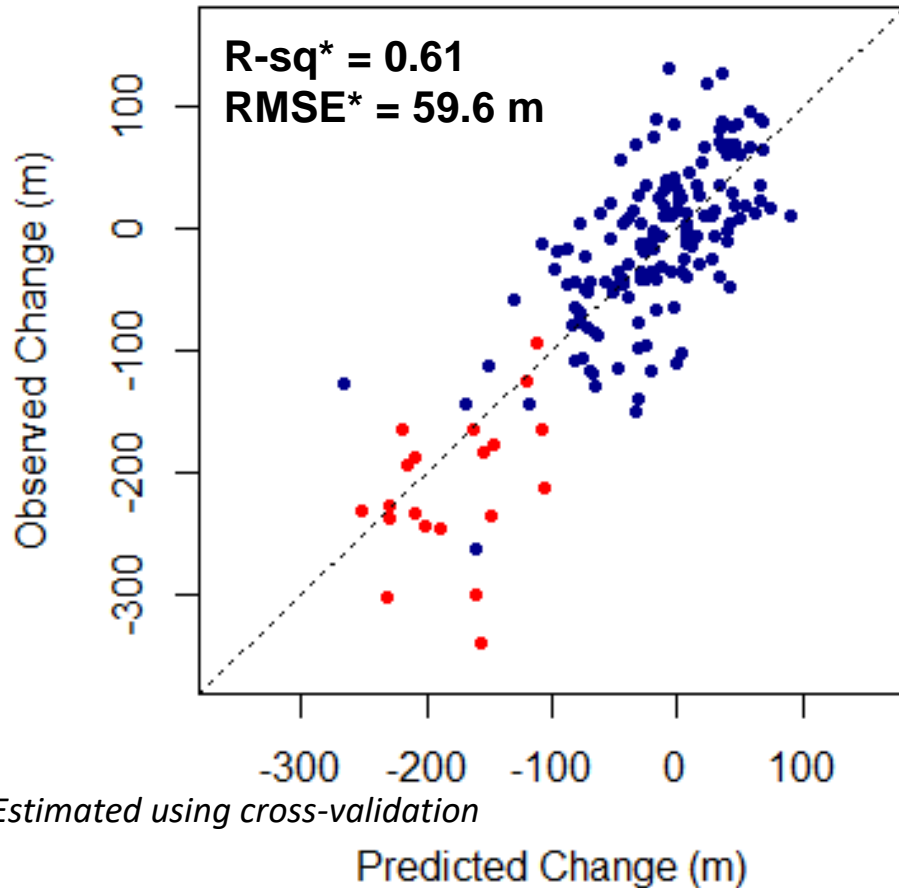


Telethon: age + 6MWD + steroids

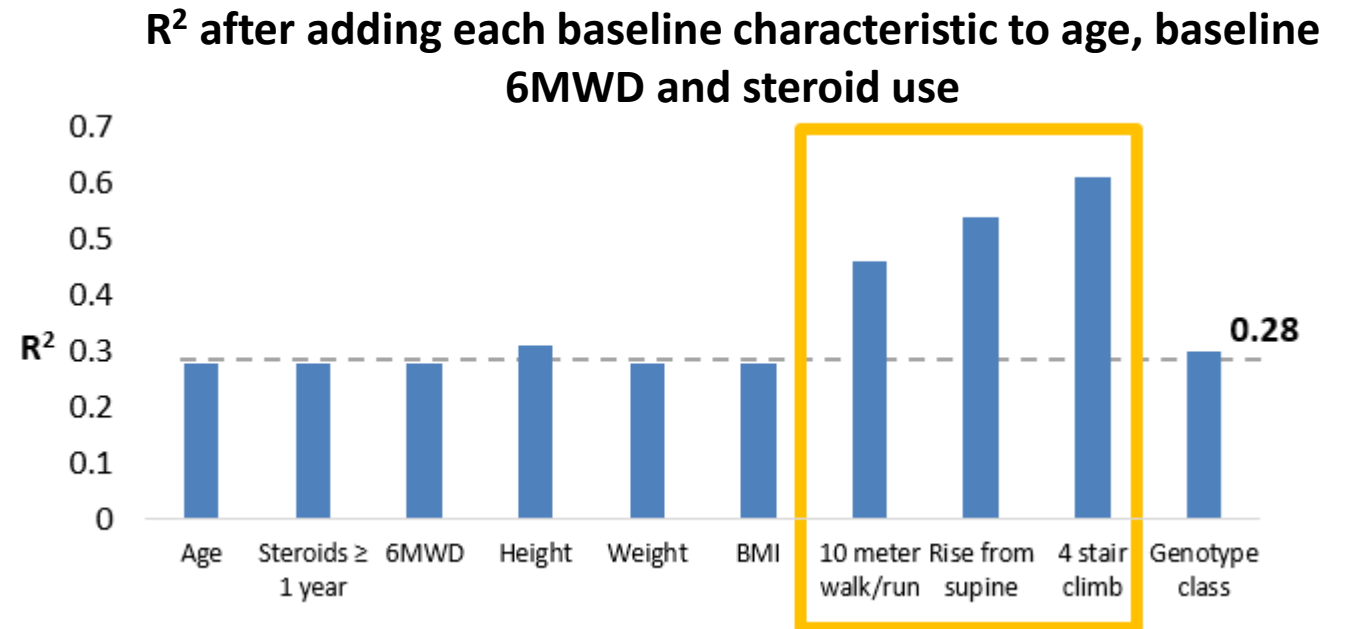




# Composite prognostic model more than doubles prognostic accuracy, reduces unaccounted for variance



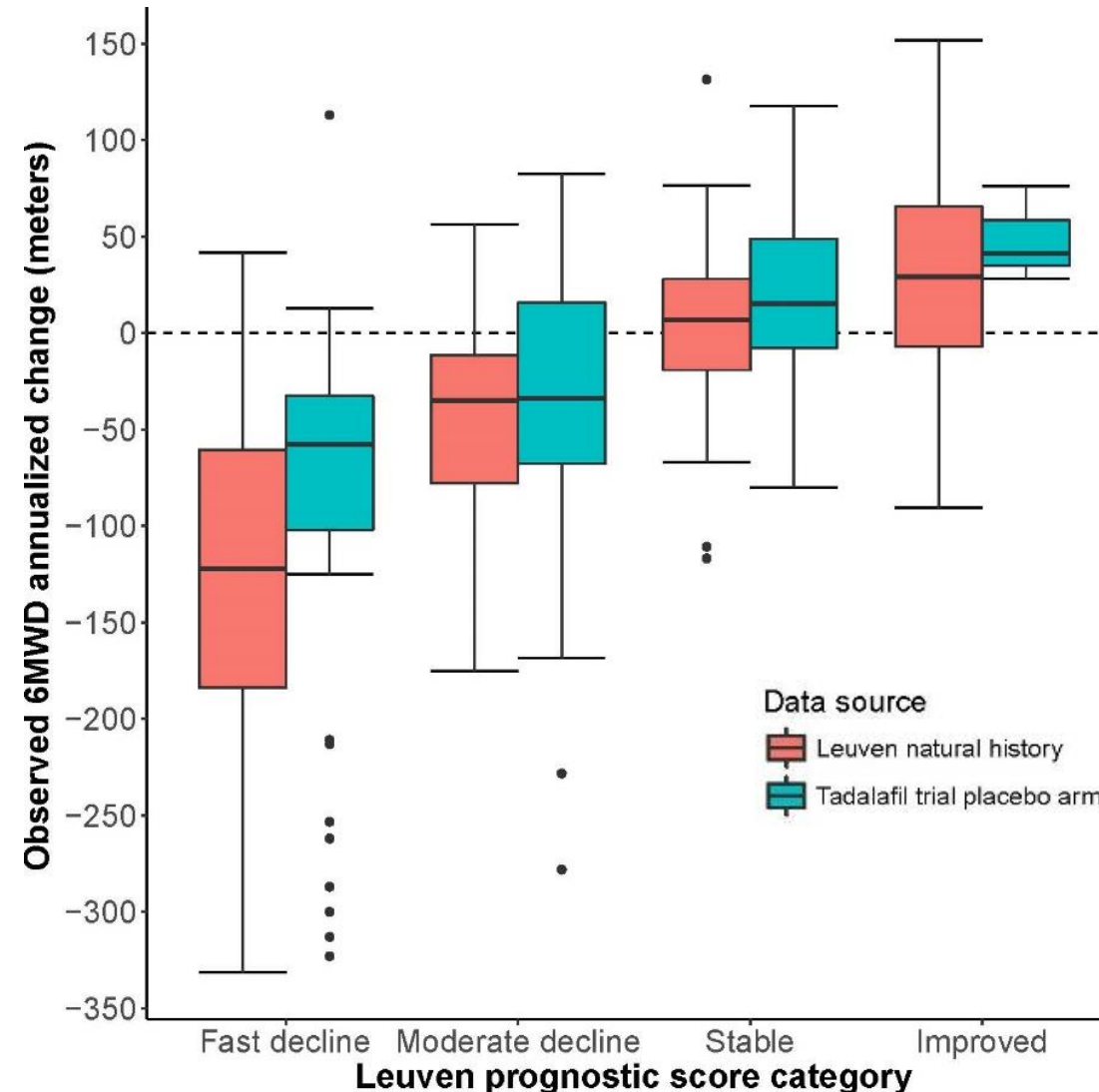
\*Estimated using cross-validation



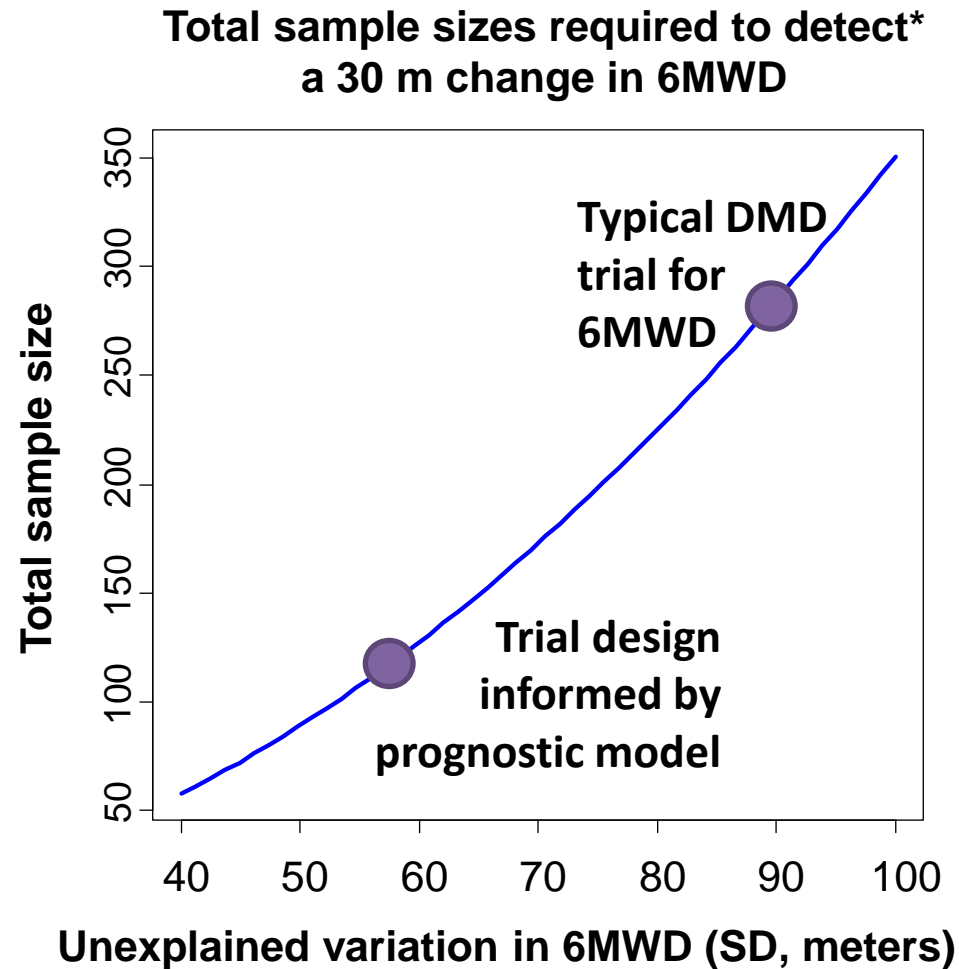
Goemans N, vanden Hauwe M, Signorovitch J, Swallow E, Song J, CollaborativeTrajectory Analysis Project (cTAP) (2016). Individualized Prediction of Changes in 6-MinuteWalk Distance for Patients with Duchenne Muscular Dystrophy. PLoS ONE 11(10): e0164684. doi:10.1371/journal.pone.0164684

# Validation in placebo arm data

- Placebo arm data from the tadalafil phase 3 trial
- No statistically significant differences in  $\Delta 6\text{MWD}$  between data sources within any baseline prognostic category (all  $p > 0.05$ ).



# Impact of selection/enrichment guided by a prognostic score



- explain and reduce variability in outcomes
- enrich for modifiable trajectories
- greater power to detect drug effects
- smaller trials with 100s fewer patients

*\*with 80% power and equal allocation to two groups*

# Conclusions

## ***In this analysis of prognostic factors for 1-year change in 6MWD***

- Baseline 6MWD, age and steroid use were not strong prognostic factors; together they explained only ~25% of variation
- A composite model that combined multiple measures of ambulatory function more than doubled explained variation to 60%
- This model performed well when applied to placebo arm data from a clinical trial
- Composite prognostic scores should prove superior in defining inclusion/exclusion criteria and/or stratification factors

# What we've learned

- Prognostic modeling also more than doubles prognostic power for 1-year change in additional ambulatory outcome measures
- consensus prognostic model for 6MWD
- Developing approaches for establish prognostic factors and model for 18 month, 2year and 3year follow-up

# Natural History as an external control?

# Natural History External Controls

- Reducing trial size is a high priority for everyone
- Smaller (or no) placebo arm in trials is a high priority for patients
- Gene therapy trials
- However, comparisons of functional outcomes such as 6MWD between drug trials and NH controls could be biased by differences in patient motivation, supportive care or assessment procedures
- This concern has been raised by regulators (e.g., April 2016 AdComm briefing documents for eteplirsen)

# Objectives

- Systematically identify and compare 48-week changes in 6MWD among natural history data sources and clinical trials placebo arms in DMD
- Do we see consistency across natural history data sources?
- Is there any evidence of systematic bias, in either direction, between placebo arms and natural history?



# Clinical trials reporting 6MWD 48-week changes on placebo arms

Placebo Arm	Number of patients	Inclusion/Exclusion Criteria			
		Steroid use (m)	Age (yrs)	6MWD (m)	Rise from supine (s)
<b>Tadalafil Phase 3</b>	116	≥6	7-14	200-400	-
<b>Ataluren Phase 2b</b>	57	≥6	≥5	≥75	-
<b>Ataluren Phase 3</b>	115	≥6	7-16	≥150	-
<b>Drisapersen Phase 2*</b>	34	≥6	≥5	≥75	≤7
<b>Drisapersen Phase 3</b>	61	≥6	≥5	≥75	-

Notes:

- Identified via systematic review of PubMed, clinicaltrials.gov and FDA briefing documents; baseline characteristics and 48-week changes in 6MWD extracted by two reviewers working independently

*\*Pooled two phase 2 trials*

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# Natural history studies reporting 6MWD

Source	Peer-reviewed publication(s) reporting longitudinal changes in 6MWD	cTAP access
<b>Italian Group (Telethon)</b>	Mercuri et al. 2016 Mazzone et al. 2011, 2013, 2016 Pane et al. 2014a, 2014b, 2014c	Shared
<b>UZ Leuven</b>	Goemans et al. 2013, 2016, 2017	Shared
<b>CINRG</b>	McDonald et al. 2013a, 2013b Henricson et al. 2012, 2013	Parallel analyses via collaboration and shared SAP
<b>Imaging DMD</b>	Willcocks et al. 2016	In discussion for data sharing

## Notes:

- Identified via systematic review of PubMed for Duchenne and ('six minute walk' OR 6MWD OR 6MWT)
- Required > 30 DMD patients with serial assessments of 6MWD
- Unpublished sources of serial 6MWD assessments include the AFM and Biomarin natural history studies; cTAP is in discussions to obtain collaborative access to both of these sources

# Study Design

- Identified non-overlapping periods of ~48-week follow-up in NH (9-13 months)
- Subjected each interval to the inclusion/exclusion criteria used in the clinical trials
- Compared mean 48-week changes in 6MWD between trial placebo arms and harmonized (matched on I/E), and between sources of NH
- Accounted for use of repeated measures via generalized estimating equations

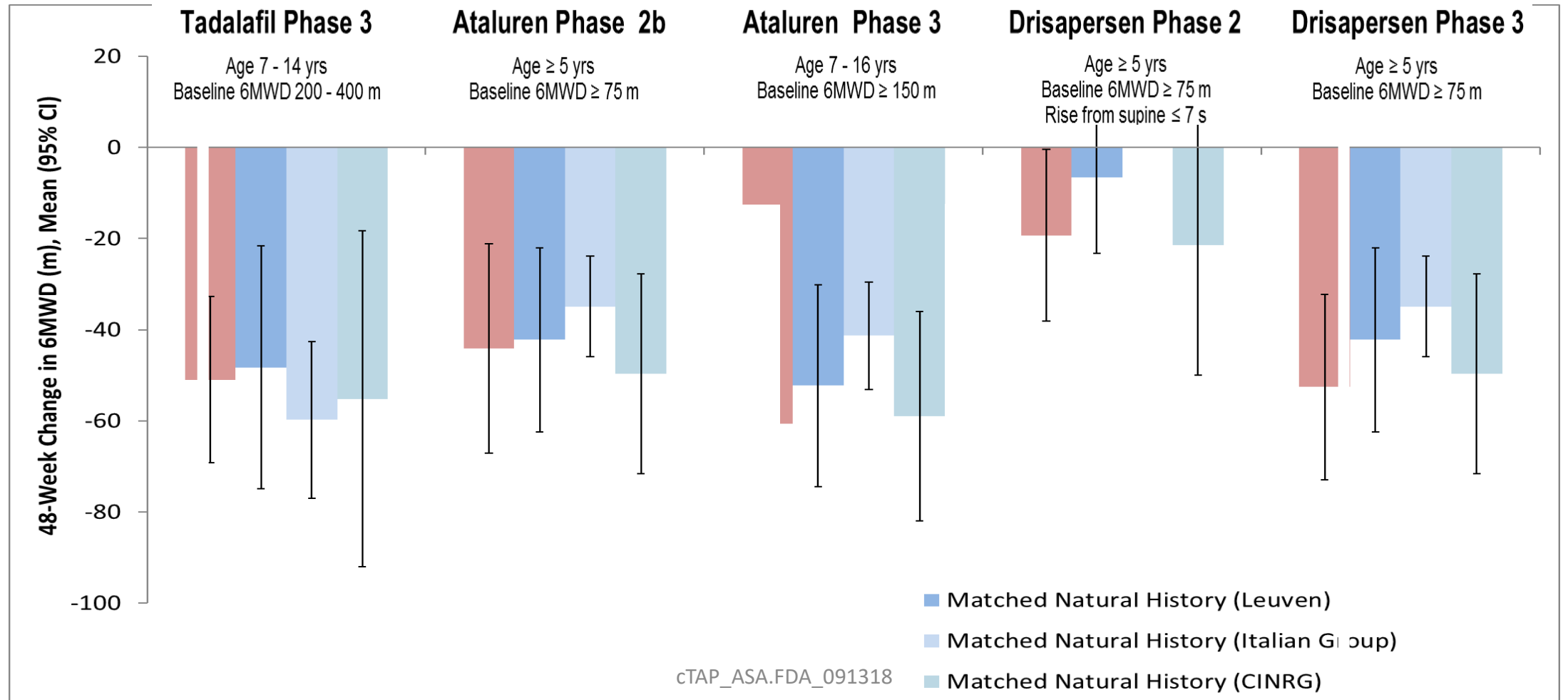
*Natural history patients were matched to trial inclusion criteria*

*Repeated measures of 6MWD were used: all non-overlapping pairs of assessments separated by approximately 48 weeks*

*Statistical analyses accounted for within-patient correlation*

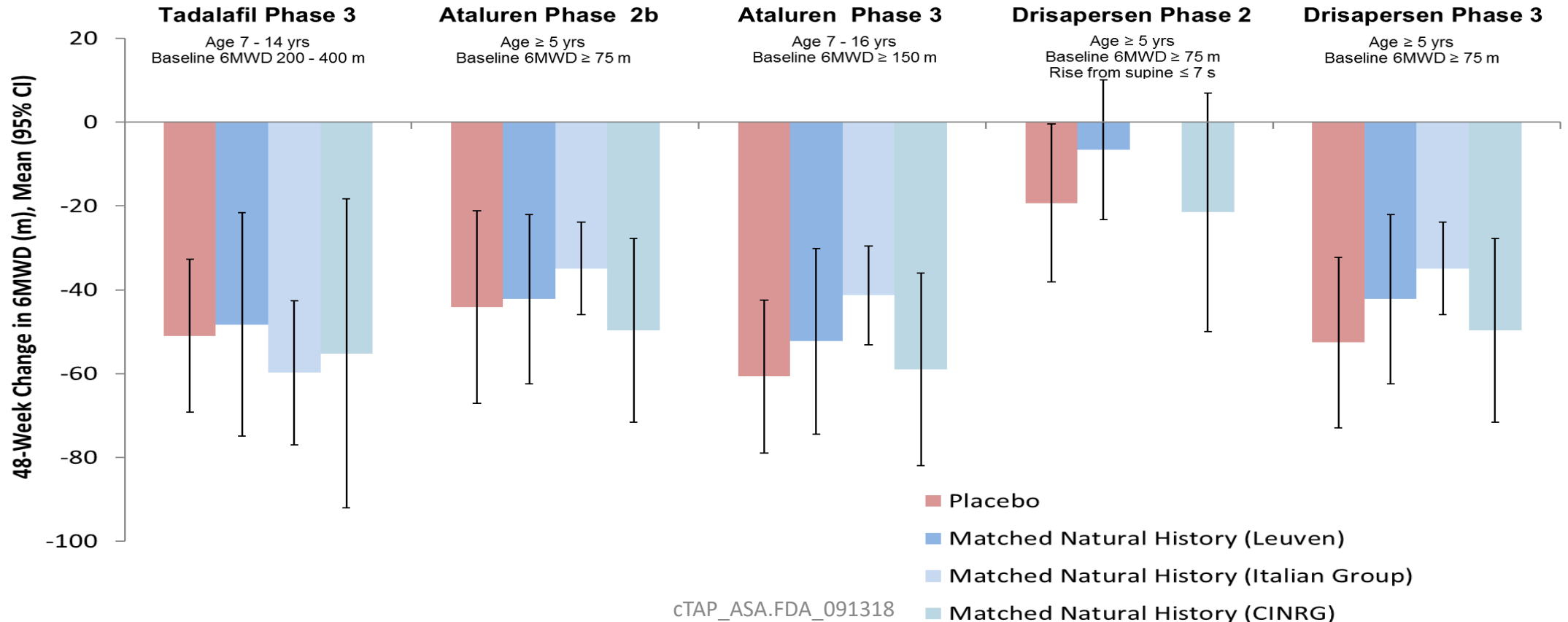
# Comparison of 48-week changes in 6MWD

- No statistically significant differences in mean  $\Delta$ 6MWD among natural history data sources
- All were all within  $\pm 18$  meters of each other after matching I/E criteria



# Comparison of 48-week changes in 6MWD

- No statistically significant differences in mean  $\Delta$ 6MWD between placebo and matched natural history
- Tendency is for placebo to have greater declines in 1 yr change 6MWD



# What we've learned

- No evidence for bias between NH and placebo for 1 yr change in 6MWD in this study
- Comprehensive assessment of sensitivity, and hidden sources of bias
- Study extended to now include 5 natural history sources and 3 placebo arms
- Currently replicating for additional outcome measures
- Assessing boundaries for when drug effect is too small for NH comparisons to be valid

# Summary perspectives

- **Natural history is crucial in rare disease drug development**
  - To learn from, to enrich trials
  - but only if researchers (and regulators) can access it
  - Patients have more power over access than they realize (yet)
- **Collaborative Problem-solving can be >>> the sum of its parts**
  - Heterogeneity has structure (relates to disease, not outcomes)
  - Methodology not necessarily fancy (doubled prognostic power)
  - Important questions might languish (consistency of outcomes across dbs)
- **Making it work**
  - Align incentives (crack the toughest nut first)
  - Independent, objective 3<sup>rd</sup> party can lower barriers, diffuse issues
  - Build Trust: be competent, be transparent, and transparently fair

# What might future trials look like?

## DRIVERS

- Stronger patient voice
- Wider adoption of data-sharing
- Faster learning through collaborative problem solving
- Robust foundation for natural history/real world evidence as controls
- Independent, SAP-driven matched comparisons for specific trials



- Smarter stratification and inclusion/exclusion criteria (independent of design)
- Smaller trials, fewer patients in placebo, enrichment with NH
- Objective, fact-based selection of controls for gene therapy trials
- Greater regulatory confidence, faster path to patient access

*Therapies to patients sooner*





COLLABORATIVE  
**TRAJECTORY  
ANALYSIS  
PROJECT**

## Collaborative Learning from Patient Data in Rare Disease

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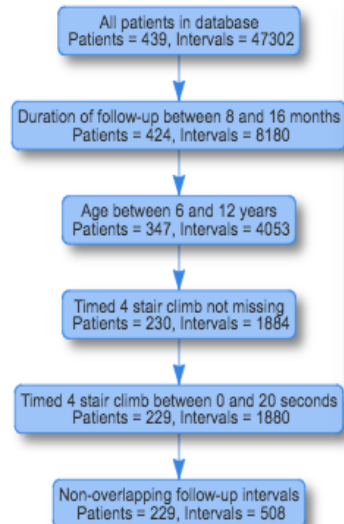
617-448-2617

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com](mailto:James.signorovitch@analysisgroup.com)

# Portal

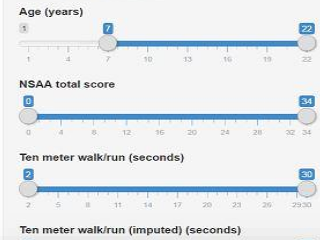
## Discovery Portal Report: CCHMC

Sample selection Baseline characteristics Outcomes Dictionary



## Discovery Portal: North Star UK

### Inclusion criteria

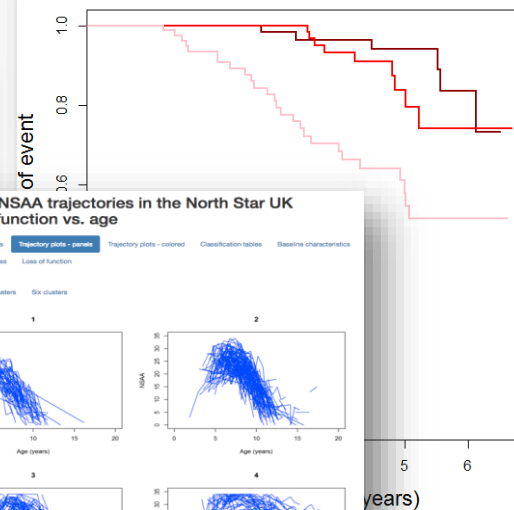


Sample selection Baseline characteristics Outcomes Correlations Metadata Definitions Settings

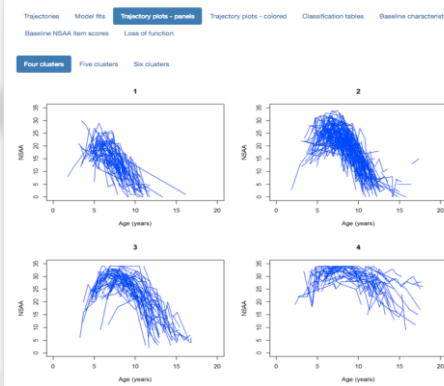
More details

Variable	All intervals (n=551)	Steroid type deflazacort (n=22)	Steroid type Missing or not started (n=173)	Steroid type prednisolone (n=356)
Age, years	9.53 ± 2.02	9.96 ± 2.54	9.31 ± 1.91	9.62 ± 2.03
NSAA total score	21.72 ± 8	19.09 ± 8.08	21.82 ± 8.24	21.84 ± 7.88
Ten meter walk/run, seconds	6.99 ± 3.06	7.63 ± 2.45	6.83 ± 2.74	7.03 ± 3.24
Ten meter walk/run (imputed), seconds	7.24 ± 3.86	7.63 ± 2.45	7.47 ± 4.67	7.1 ± 3.5
Ten meter walk/run velocity, meters/second	1.64 ± 0.56	1.44 ± 0.44	1.69 ± 0.62	1.63 ± 0.54
Ten meter walk/run velocity (imputed), meters/second	1.62 ± 0.58	1.44 ± 0.44	1.66 ± 0.66	1.62 ± 0.54
NSAA walk item score	0	0	0	0
	6 (1.09)	0 (0)	1 (0.58)	5 (1.4)
	242 (43.92)	12 (54.55)	70 (40.46)	160 (44.94)
	290 (52.63)	10 (45.45)	96 (55.49)	184 (51.69)
	13 (2.36)	0 (0)	6 (3.47)	7 (1.97)
	7.49 ± 5.89	10.08 ± 7.09	7.45 ± 6.25	7.35 ± 5.6
	8.8 ± 7.79	13.06 ± 9.78	8.65 ± 7.92	8.59 ± 7.51
	0.2 ± 0.12	0.15 ± 0.09	0.21 ± 0.13	0.2 ± 0.12
	0.19 ± 0.12	0.13 ± 0.09	0.2 ± 0.13	0.19 ± 0.12
	37.54 ± 23.84	34.27 ± 19.8	35.34 ± 25.74	38.64 ± 23.04

## Time to FMS > 4



## Clustering NSAA trajectories in the North Star UK database: function vs. age



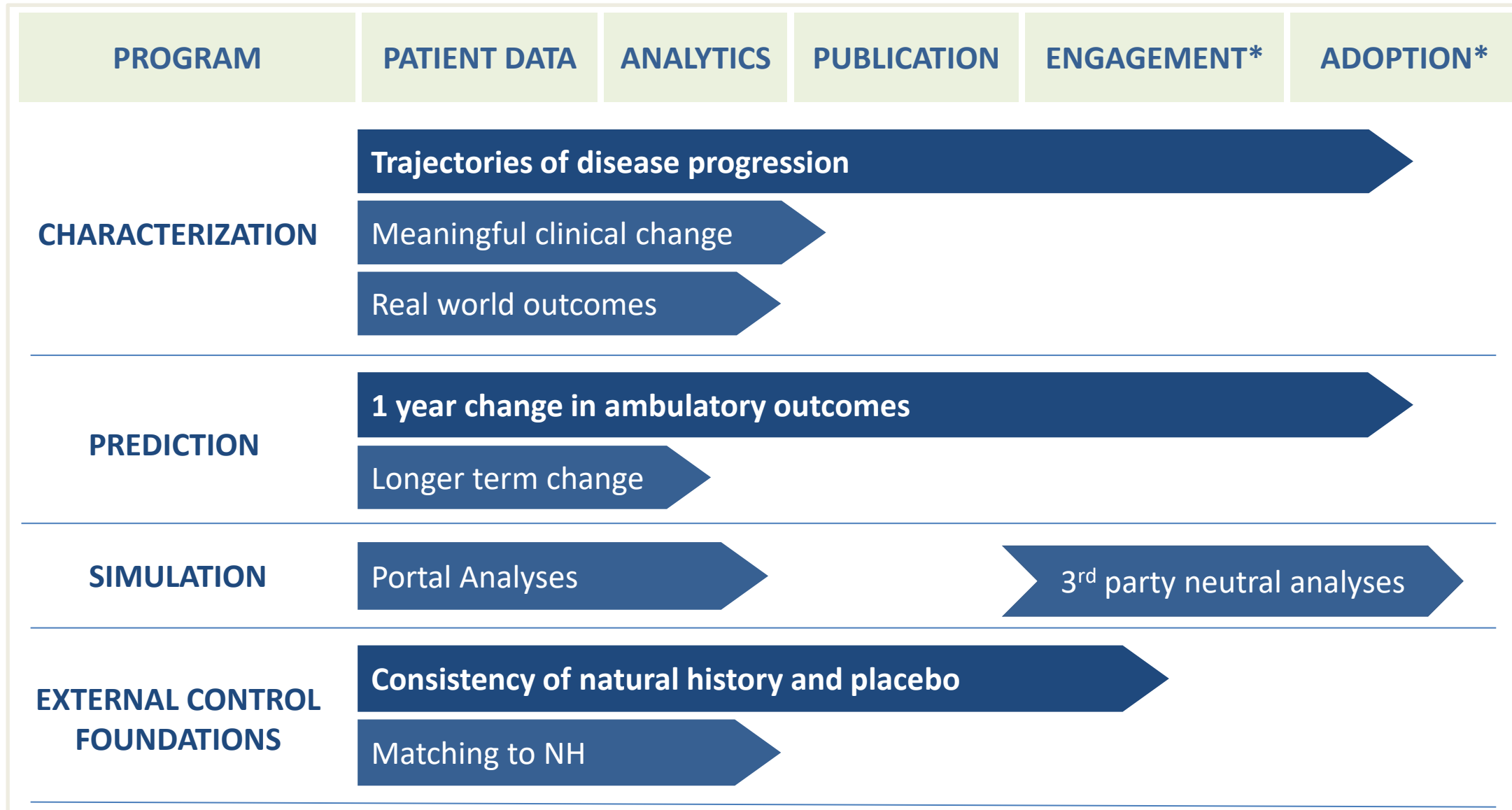
*Harmonized data*

*Dynamic analytics*

*Rapidly test hypotheses*

*Timely replication and extension*

# cTAP Pipeline (leading outcome measure)



\*With primary target group – academic, regulators, payors/health authorities – dependent on program

# Common mining questions from cTAP members

- Counts: Do you have sufficient data to....
- Design: Given my trials I/Es, what might we anticipate as 1 yr change in my primary outcome measure? How consistent is that estimate across data sources?
- Enrolment: How can I accelerate/improve efficiency of enrolment - without compromising the planned power of my study?
- Interpretation: Using the baseline data of patients in our trial, what are the outcomes for a matched cohort?