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Use of item-based non-linear mixed effects model to improve confidence in Phase III clinical trial decision-making

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Failures in clinical drug development

Phase II & III



Early termination of “poor” assets

	Reasons for failure	
	Phase II	Phase III
Efficacy	48%	55%
Safety	25%	14%
Strategy	21%	14%
Others	6%	17%

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Efficacy failures:

- x Insufficient knowledge about target population
- x Insufficient sample size
- x Large uncertainty in drug efficacy estimate
- x Insufficiently powerful analysis methods

Chronic Obstructive Pulmonary Disease

(COPD)

- Outcomes of interest in efficacy trials:
 - ✓ Airway obstruction (Forced Expiratory Volume in 1 second – FEV₁)
 - ✓ Dyspnea (Transition Dyspnea Index - TDI)
 - ✓ Health status (St. George Respiratory Questionnaire - SGRQ)
 - ✓ Exacerbations
 - ✓ **Patient reported Outcomes (PROs):**
 - Reported directly by the patient
 - Increasingly used in drug development

How do we “measure” disease?

PRO in COPD

- **EXACT** – 14 questions related to COPD symptoms
- **E-RS:COPD** – 11 questions specifically related to respiratory symptoms

Item number	Item-level construct	Score	Symptom construct
7	Breathless today	0 - 4	Breathlessness
8	Breathless with activity	0 - 3	
9	Short of breath – personal care	0 - 4	
10	Short of breath – indoor activity	0 - 3	
11	Short of breath – outdoor activity	0 - 3	
2	Cough frequency	0 - 4	Cough and sputum
3	Mucus quantity	0 - 3	
4	Difficulty with mucus	0 - 4	
1	Congestion	0 - 4	Chest symptoms
5	Discomfort	0 - 4	
6	Tightness	0 - 4	
12	Tired or weak	0 - 4	Additional attributes
13	Sleep disturbance	0 - 4	
14	Scared or worried	0 - 3	

EXACT total scores uses all 14 items with logit scoring transformed to a 0 to 100 interval-level scale; E-RS:COPD scores are based on summation to yield ordinal-level scales with a total score ranging from 0 to 40



How are such data typically analysed?

Mixed Model Repeated Measures (MMRM)

- Total score analysis
- Non-ignorable missing data requires a sensitivity analysis
- Time is handled as discrete variable

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Sum of scores



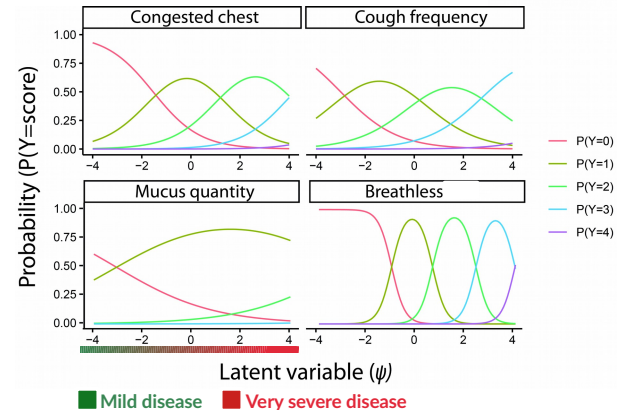
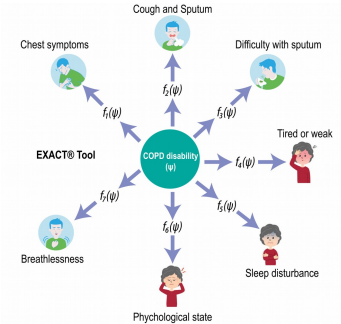
- Mathematical models expressing the **probability** of the particular response to a **scale item** as a function of an **underlying trait** or **latent variable**
- All questionnaire information is used
- Handles non-ignorable missingness
- Increased power to detect drug effect

Alternative analysis

Item-based response model (IRM)

$$P(y_{ij} \geq k) = \frac{e^{(a_j(\psi_i - b_{j,k}))}}{1 + e^{(a_j(\psi_i - b_{j,k}))}}$$

$$P(y_{ij} = k) = P(y_{ij} \geq k) - P(y_{ij} \geq k + 1)$$



Phase II

Observational Study > [AAPS J. 2019 Apr 26;21\(4\):60. doi: 10.1208/s12248-019-0319-9.](#)

A Novel Method for Analysing Frequent Observations from Questionnaires in Order to Model Patient-Reported Outcomes: Application to EXACT® Daily Diary Data from COPD Patients

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Phase III?



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Phase III?

Clinical Trial > [Adv Ther. 2018 Jan;35\(1\):56-71. doi: 10.1007/s12325-017-0650-4.](#)
Epub 2018 Jan 8.

Once-Daily Triple Therapy in Patients with COPD: Patient-Reported Symptoms and Quality of Life

Maggie Tabberer¹, David A Lomas², Ruby Birk³, Noushin Brealey³, Chang-Qing Zhu⁴, Steve Pascoe⁵, Nicholas Locantore⁵, David A Lipson^{5,6}

Phase III, randomized, double-blind, double dummy, parallel-group, multicenter study to evaluate once daily *fluticasone furoate/umeclidinium/vilanterol* (FF/UMEC/VI) (100 ug/62.5 ug/25 ug) inhalation powder versus twice daily *budesonide/formoterol* (BUD/FOR) (400 ug/12 ug) in patients with **COPD**



Aim & Methods

To illustrate how a **new methodology** to analyse **PRO data** based on a longitudinal **IRM** improves **confidence** in a **Phase III clinical trial** endpoint compared with a **MMRM** analysis.

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Step 1^a - IRM

- **Item characteristic functions**
- ✓ Graded response model (a logistic transformation to model each item)
- ✓ Independent occasion approach

Step 2^a - Longitudinal model - modeling Ψ_i :

- Use the estimates of Ψ from step 1 as DV^b
- Different parameters between treatment arms
- Pre-specified Weibull model
- Smoking status and geographical regions on $\psi_{i,t=0}$

a: Analysis performed using NONMEM v.7.4.4 with an Intel FORTRAN compiler and PsN v.5.1.0

b: Schindler et al. Pharm Res 2018;35:122

Ψ_i : Individual latent variable

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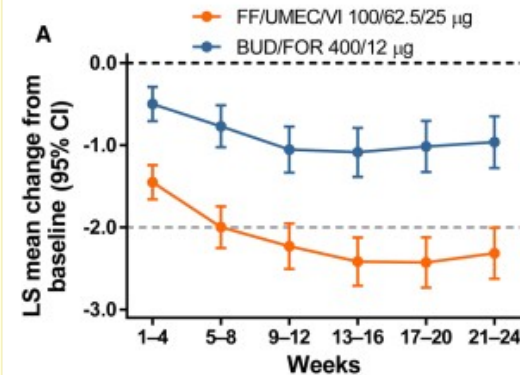
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Step 3^a - Total score (RS-total)^c simulations and CI calculation for each treatment arm

- Perform simulations including parameter uncertainty ($\$PRIOR$)
- Calculate 95% CI - Mean difference in total score between baseline and four week intervals:
 - ✓ Week 0 - 4, week 5 - 8, week 9 - 12 ...

Published MMRM results



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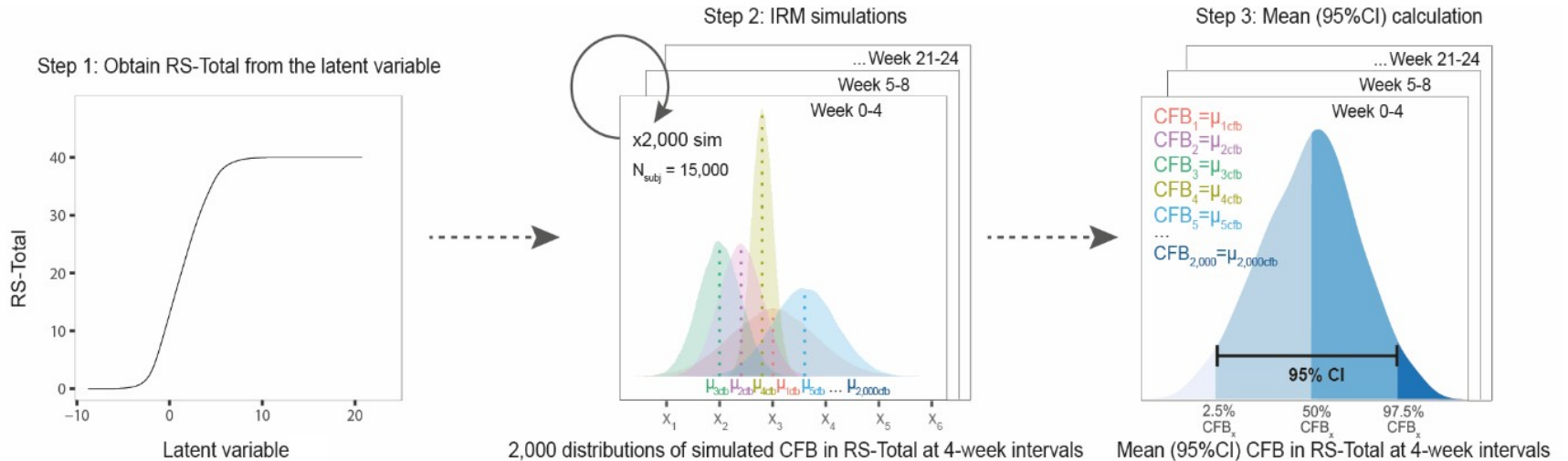
Ψ_i : Individual latent variable

c: Total score from the E-RS:COPD questionnaire



Simulations

Including parameter uncertainty



Simulations included uncertainty for each parameter value (\$PRIOR)

NWPRI:

- Multivariate normal distribution for THETAs
- Inverse Wishart distribution for OMEGAs

Sample size (N)

$$N = \left(\frac{CI_{MMRM}}{CI_{IRM}} \right)^2$$



Baseline characteristics	FF/UMEC/VI (n=907)	BUD/FOR (n=894)
Age (years)	64.2 (8.56)	63.6 (8.73)
FVC (L)	2.84 (0.80)	2.87 (0.79)
FEV ₁ (L)	1.25 (0.46)	1.24 (0.45)
Male (n)	675 (74%)	658 (74%)
Smoker (n)	396 (44%)	392 (44%)
COPD GOLD disease status	Moderate: 298 (33%) Severe: 501 (55%) Very severe: 107 (12%)	Mild: 1 (0.1%) Moderate: 290 (32%) Severe: 477 (53%) Very severe: 124 (14%)
RS- total score ^a	12.2 (5.85)	12.9 (5.96)

Values are mean (SD) or n (%); a: scores were calculated as the mean value during baseline period defined as from day -14 to day -1; FVC: forced vital capacity ; FEV₁: forced expiratory volume in one second; GOLD: global initiative for chronic obstructive lung disease; RS-total score ranged from 0 to 40; In this analysis, nine patients were excluded from the intention to treat population of the FULFIL clinical trial (NCT02345161) for the following reasons: absence of E-RS:COPD score data for the whole study period (4 patients), dispensing errors (4 patients), and missing recorded time (1 patient).



Parameter estimates

Longitudinal model

Parameter	FF/UMEC/VI Value (RSE)	BUD/FOR Value (RSE)
Baseline ($\Psi_{i,t=0}$) ^a	0.33 (0.23)	0.29 (0.28)
Time of response (T_R)	0.08 (0.03)	0.08 (0.03)
Maximum response (R_{MAX})	-0.31 (0.11)	-0.16 (0.25)
γ	9.27 (0.13)	16.9 (0.75)
Offset	-0.27 (0.08)	-0.08 (0.30)
ω^2 Baseline	1.09 (0.03)	1.47 (0.03)
ω^2 Time of response	0.45 (0.03)	0.46 (0.03)
ω^2 Maximum response	0.89 (0.04)	0.98 (0.06)
ω^2 Offset	0.37 (0.05)	0.42 (0.05)
RUV	0.32 (0.02)	

$$\Psi_i(t) = \Psi_{i,t=0} + R_{MAXi} \cdot \left(1 - e\left(-\left(\frac{\ln(2)}{T_{Ri}} \cdot t\right)^\gamma\right)\right) + Offset_i$$

- Weibull describes a late onset effect
- Offset near-instant or potentially symptomatic drug effect

a: Effects of smoking status and geographical regions on $\Psi_{i,t=0}$ were included in the model as these covariates were considered in the MMRM analysis



Parameter estimates

Longitudinal model

Covariates	Value (RSE)
Smoking effect	0.13 (0.62)
Region 1	-0.64 (0.13)
Region 3	-0.61 (0.13)
Region 4	-0.20 (0.40)
Region 5	-0.41 (0.27)
Region 6	-0.98 (0.12)

Region 1 (21%): Germany, Greece, Italy

Region 2 (24%): Russian Federation, Ukraine

Region 3 (21%): Bulgaria, Hungary, Romania, Slovakia

Region 4 (18%): Czech Republic, Estonia, Poland

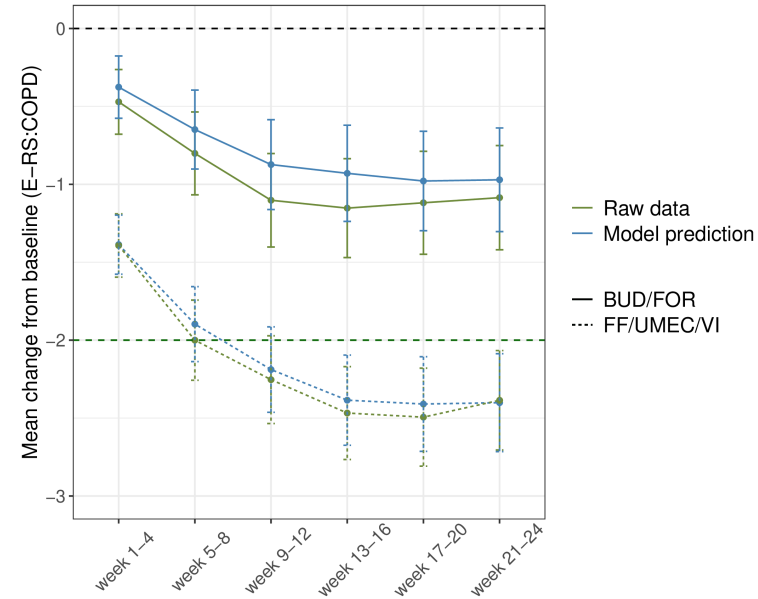
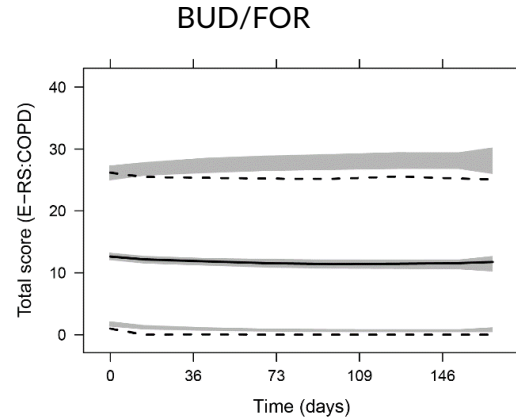
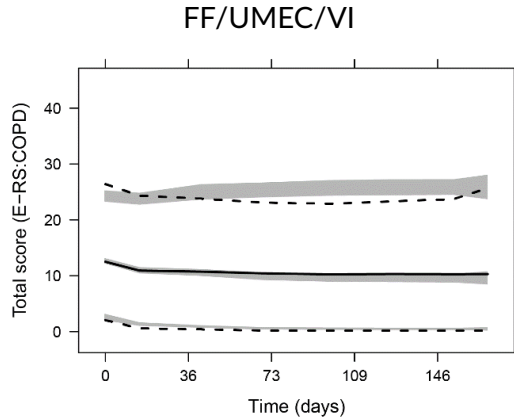
Region 5 (6%): China, Republic of Korea

Region 6 (10%): Mexico

	Correlation	Value (RSE)
FF/UMEC/VI	ω^2 Maximum response~ ω^2 Offset	11% (0.24)
	ω^2 Baseline~ ω^2 Offset	-12% (0.20)
BUD/FOR	ω^2 Maximum response~ ω^2 baseline	-13% (0.17)
	ω^2 Baseline~ ω^2 Offset	-10% (0.27)



Model predictive performance

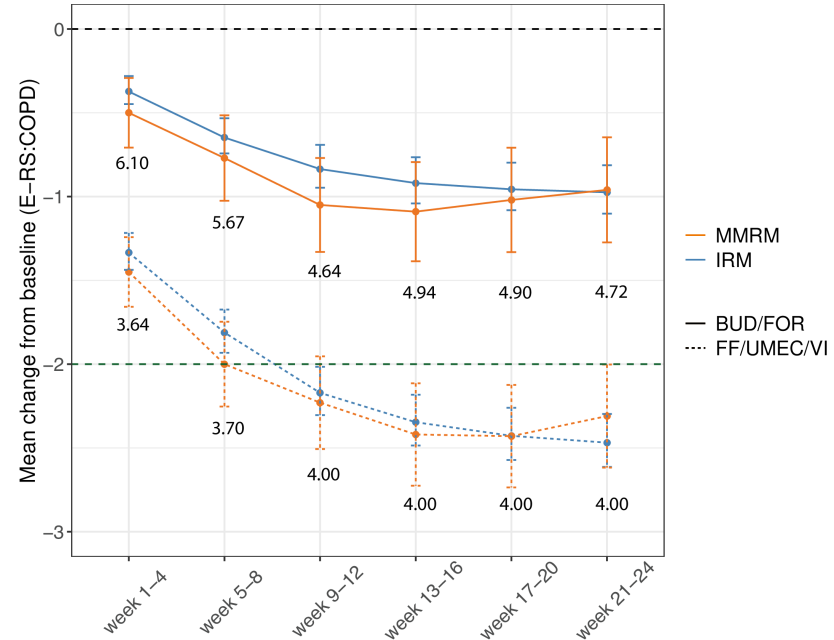


Lines are the 2.5th, 50th and 97.5th percentile of the observed data, and grey areas are the corresponding 95% confidence interval from model simulations (N=500)



Precision in outcome measures

- ✓ The IRM improved precision of the estimated drug effect compared to MMRM
- ✓ A **4.0 (FF/UMEC/VI)** and **4.72 (BUD/FOR)** times larger sample size for the MMRM analysis to achieve the precision obtained with IRM at week 21 - 24



Discussion & conclusion

- Advantage of using a non-linear mixed effect model-based analysis of Phase III clinical item-response level data, over MMRM for the same clinical endpoint in both methods (IRM and MMRM)
- Proposed methodology is applicable to any item-level data
- IRM may improve decision-making in drug development:
 - ✓ Prediction of outcomes from short term to long term study
 - ✓ Smaller studies sizes and more informative analysis

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