# A framework for prediction of progression-free survival based on modelling of sub-endpoints



Sreenath M. Krishnan,<sup>1</sup> Lena E. Friberg,<sup>1,2</sup> Joakim Nyberg,<sup>1</sup> Céline Sarr,<sup>1\*</sup> Ronald Niebecker,<sup>3</sup> Shaonan Wang,<sup>3</sup> Alejandro Perez Pitarch<sup>3</sup>

<sup>1</sup>Pharmetheus AB, Uppsala, Sweden; <sup>2</sup>Department of Pharmacy, Uppsala University, Uppsala, Sweden; <sup>3</sup>Boehringer Ingelheim Pharma GmbH & Co.KG, Biberach, Germany

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- The assessment of PFS in solid tumours is based on RECIST, which evaluates certain sub-endpoints:
- Change in the sum Appearance of Progression of of longest new lesion (NL) non-target lesion diameter of target (NTR) lesions (SLD)
- TTE models have been used to describe PFS. However, TTE models may not be discriminant, especially in small trials
- An alternative method that jointly models PFS as target lesion progression and non-target progression has been

\*Corresponding author email address: <a href="mailto:celine.sarr.ext@boehringer-ingelheim.com">celine.sarr.ext@boehringer-ingelheim.com</a>

#### **Methods**

- Data from Phase III trials were used to develop the modelling framework (Figure 1; Table 1)
- For longitudinal SLD data, two TGI models were considered; one assumed acquired resistance<sup>2</sup> and the other assumed the presence of resistant and sensitive populations at treatment initiation<sup>3</sup>
- Binary NTR and NL data collected at each tumour scan were modelled using a logistic regression model estimating the probability of an event at the time of the scan
- In case of estimation issues with logistic regression

## **A Methods (cont'd)**

- Death prior to progression and dropout from study due to other reasons than progression could happen at any time; hence, a TTE model was used
- In the joint modelling framework, the clinical endpoints including PFS were simulated based on the events in the sub-endpoints' models
- The model evaluations of sub-endpoints as well as the joint model were based on VPCs. NONMEM<sup>®</sup> 7.4 software was used for model development<sup>7</sup>

Table 1. Data for population model (N=2049 patients)

LUX-Lung 3	LUME-Lung 1	LUME-Ovar 1
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proposed by Yu *et al.*<sup>1</sup> Using this methodology, target lesion progression is derived from tumour growth dynamics, and risk of non-target progression is correlated to target lesion tumour size with respect to time. This approach has the potential to predict PFS based on data collected in the early stages of drug development<sup>1</sup>

# **Objectives**

 The aim of the current study was to develop a modelling and simulation framework, based on sub-endpoints, that predicts long-term typical clinical readouts such as PFS, ORR, DOR and BOR

BOR, best overall response; DOR, duration of response; ORR, objective response rate; PFS, progression-free survival; RECIST, Response Evaluation Criteria In Solid Tumors; TTE, time-to-event

models, a TTE model was considered for modelling NTR and NL

#### Figure 1. Overview of the three Phase III trials used in the population model

LUX-Lung 3 <sup>4</sup> NCT00949650 Untreated <i>EGFR</i> m+ NSCLC		LUME-Lung 1 <sup>5</sup> NCT00805194 NSCLC after failure of 1L chemotherapy		LUME-Ovar 1 <sup>6</sup> NCT01015118 Advanced ovarian cancer with no prior systemic treatment			
Afatinib	Pemetrexed /cisplatin	Nintedanib + docetaxel	Placebo + docetaxel	Nintedanib + carboplatin + paclitaxel	Placebo + carboplatin + paclitaxel		
<b>PFS, months:</b> 11.1 vs 6.93.4 vs 2.717.2 vs 16.6							
HR (95% C 0.58 (0	<b>:I):</b> ).43–0.78)	0.79 (0.0	68–0.92)	0.84 (0.	72–0.98)		
1L, first-line; CI, confidence interval; <i>EGFR</i> m+; epidermal growth factor receptor mutation-positive; HR, hazard ratio; NSCLC, non-small cell lung cancer; TGI, tumour growth inhibition							

	Afatinib (n=225)	Chemo (n=107)	Nintedanib + docetaxel (n=607)	Placebo + docetaxel (n=605)	Nintedanib + chemo (n=332)	Placebo + chemo (n=173)
Median age, years (range)	61 (28–86)	62 (31–83)	60 (29–84)	60 (30–80)	59 (23–84)	59 (25–79)
Median SLD, mm (range)	55 (10–760)	52 (10–234)	85 (10–410)	84 (10–394)	49 (10–330)	48 (10–457)
Observed NTR*, n (median TTE, wks)	58 (35)	25 (17)	144 (13)	175 (13)	181 (48)	72 (46)
Observed NL*, n (median TTE, wks)	111 (38)	44 (23)	334 (12)	356 (12)	556 (61)	288 (60)
Deaths before progression, n (median TTE, wks)	8 (13)	2 (8)	87 (10)	62 (10)	19 (15)	8 (12)
Dropouts before progression, n (median TTE, wks)	30 (146)	20 (23)	82 (12)	67 (12)	254 (120)	115 (120)

\*Data incomplete for LUME-Ovar 1 Chemo, chemotherapy; NONMEM, nonlinear mixed effects modelling; VPCs, visual predictive checks; wks, weeks

# Key findings and conclusions

- Population models for individual metrics were developed that jointly described PFS
- Despite the diversity of responses observed in the Phase III trials, the model framework could also predict other long-term clinical readouts, such as ORR, DOR and BOR



- The model framework may be used for predicting PFS before PFS data are mature
- This approach may therefore provide support for early decision making in indications for which PFS is an important clinical endpoint

of the poster and supplementary content

### III. Results

K-M, Kaplan-Meier

- Models for individual metrics are shown in the supplementary content
- VPCs based on 200 simulations showed acceptable predictive performance of the tumour size model (Figure 2)

#### Figure 2. VPC of SLD of target lesions in LUX-Lung 3



### **III** Results (cont'd)

#### Figure 3. K–M VPCs of death events in LUX-Lung 3

Solid line: observed data. Shaded area: 95% CI from 200 model simulations



## **III.** Results (cont'd)

#### Figure 4. PFS simulations based on individual metrics Observed PFS (solid line) vs 95% CI (shaded area) of the simulated PFS

#### A) LUX-Lung 3



#### Time since first dose (days)

- K–M VPCs showed acceptable predictive performance of the logistic regression models for NTR, NL and for death events (Figure 3)
- K–M VPCs showed acceptable predictive performance with regard to PFS across studies (Figure 4)
- The model framework could also predict long-term clinical readouts other than PFS, such as ORR, DOR and BOR, despite the diversity of response observed in the three Phase III trials (see supplementary content)

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