Tumour growth model to improve early PFS estimation thereby enhancing phase II go/no decisions

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BACKGROUND

- Survival is considered the most reliable endpoint in cancer clinical trials, but adequate determination of overall survival is often not feasible because of:
 - small sample sizes

Statistical Consulting

SIMULATION

- A simulation study of 1000 repetitions Measurements at discrete each assuming 6 months accrual was timepoints conducted for 7 scenarios: • Incorporating measurement errors Proof of Concept (PoCP) scenarios Accounting for patient accrual and 2 with 1000 patients each and follow-up time of 2 or 6 months **Observed SoD Model** (like HR estimated from real for scenario 1 and 2 patient data) Phase II scenarios 3-7 with 80 Estimating model parameters based on patients each and follow-up time of patient-specific measurements derived 6 months from observed model: • Fixed effects (5 parameters for all Each simulation run based on the patients) published parameterisation of [1] Random effects (3 parameters for each patient) Figure 1: Measurements until PD or censored from test or reference as input for estimating Predicted SoD Model parameters (fixed and random effects) of the Linear Mixed Model for randomly selected patients Figure 2: RECIST PD = 85.71 + 20% * 85.71 at scheduled = 102.85 visits with Fimepoint of PD: 9 months

True SoD Model

True Hazard Ratio (HR) for PFS

- **Observed Hazard Ratio for PFS** Predicted Hazard Ratio for PFS based on the predicted PD time

- short follow-up in early phase studies
- An early but reliable estimate of progression-free survival (PFS) in phase II can serve as decision criterion to
 - carry the investigational drug to the next phase III trial
 - stop further investigations of the anti-cancer compound
- High failure rates of phase III trials (~ 60%) and high costs emphasize high importance of improving the decision process for pharmaceutical companies especially in the field of oncology
- Early PFS estimations are often used to start planning pivotal phase III trials, but estimations with less than 50% of events observed do not use all available information from tumour measurements over time
- Tumour-growth models can use all longitudinal information and could potentially improve early PFS estimations

METHOD

TUMOUR-GROWTH MODEL

- Mixed-effects models to quantify non-linear individual relationships between time from randomisation and tumour burden using all available tumour measurements
- Applied model is flexible, allowing for wide range of different shaped growth curves (u-, j-, n- or linear-shaped)
- Tumour burden is measured by Sum of Diameter (SoD) of target lesions (TL) for each patient *i* over (continuous) time *t* by the



following linear mixed model:

 $SoD(t,i) = \beta_0 + \beta_1 * t + \beta_2 * \log(t+1)$ $+ bi_0 + bi_1 * t + bi_2 * \log(t+1)$ $+ \beta_3 * treatment(i) * t$ $+ \beta_4 * treatment(i) * \log(t+1)$ $+ \boldsymbol{\varepsilon}(t,i)$

with $\beta_0, \beta_1, \beta_2, \beta_3, \beta_4$ fixed effects

bi individual patient effects over time (random effects) $\varepsilon \sim N(0, \sigma^2)$ measurement errors (depending on time) and patient)

 β_3 , β_4 treatment effects over time (test and reference)

 Optimal parameterisation of tumour-growth in non-small cell lung cancer (NSCLC) was published implying time-dependent Hazard Ratios (HR) [1]:

 $\beta_0 = 88.305$ Fixed effects, i.e. $\beta_1 =$ population mean over time:

3.592 $\beta_2 =$ 2.423 1.335 $\beta_3 =$ $\beta_4 = -12.748$

Random effects:

$\left(bi_{0} \right)$		(/0\		/ 5.595	-2.155	0.679 \
bi ₁	$\sim N$		0	,	-2.155	16.842	-3.082
$\left(\frac{bi_2}{2}\right)$		$\left(\right)$	$\langle 0 \rangle$		0.679	-3.082	3.865 //

• For each patient *i* the mixed model yields a trajectory describing patient-specific tumour dynamics over time t (see Figures 1 and

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 Observed PFS estimation 	Table 1: Results of different simulation scenarios								
resulted in HR of 0.34 (95%-			Observed		Predicted				
percentiles: 0.29-0.40) (see Table 1)	Scenario	Measurement error σ^2	% events	HR Median (95% CI)	% events	HR Median (95% CI)			
 Predicted HR was estimated as 	Proof of Concept (PoCP) scenarios								
0.77 to 0.71 depending on the	1	1	37%	0.34 (0.29, 0.40)	84%	0.77 (0.69, 0.85)			
scenario (see Table 1)	2	1	62%	0.52 (0.47, 0.58)	84%	0.77 (0.69, 0.84)			
• Overall true HR was 0.78 (95%	Phase II study scenarios								
perc.: 0.69-0.85)	3	1	62%	0.52 (0.30, 0.93)	84%	0.77 (0.46, 1.24)			
• For 6, 12 and 120 months the	4	1.5	62%	0.52 (0.28, 0.91)	84%	0.76 (0.45, 1.26)			
time varying HRs were 0.41	5	2	63%	0.53 (0.29, 0.92)	85%	0.76 (0.43, 1.28)			
(0.36, 0.47), 0.60 (0.54, 0.66)	6	9	66%	0.55 (0.31, 0.93)	85%	0.75 (0.43, 1.27)			
and 0.77 (0.69, 0.85)	7	25	74%	0.61 (0.36, 1.03)	84%	0.71 (0.40, 1.32)			

CONCLUSIONS

- Tumour-growth models should be used to improve PFS estimations based on early readouts, especially in the presence of time dependent HRs, since they consistently provide far better estimates of the overall true HR by more than 10 points in all simulations.
- In general, tumour-growth models can support decision making in drug development, as supplement to classical PFS estimations and facilitate a better understanding of the mode of action of a drug [2,3].
- Not incorporating occurrence of new lesions or death limits tumour-growth models to indications and treatment lines where those are not the primary source for PD. However, model extensions

2): $\widehat{SoD}(t,i) = \widehat{\beta_0} + \widehat{\beta_1} * t + \widehat{\beta_2} * \log(t+1) + \widehat{bi_0} + \widehat{bi_1} * t + \widehat{bi_2} * \log(t+1)$ $+\widehat{\beta_3} * treatment(i) * t + \widehat{\beta_4} * treatment(i) * \log(t+1) + \widehat{\varepsilon}(t,i)$

- Forecast for timepoint of tumour progression for each patient calculated by applying Response Evaluation Criteria In Solid Tumours (RECIST 1.1) on patient-specific tumour trajectories:
 - At least 20% increase in SoD of TL, taking smallest sum on study (nadir) as reference and
 - Absolute increase of at least 5mm in SoD or
 - Appearance of one or more new lesions (not applicable for our simulations because we simulate neither the occurrence of new lesions nor death)
- These event times are used to calculate HR of PFS for two treatment arms by means of Cox regression with treatment as covariate

incorporating those processes are straightforward.

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Nothing to declare.

REFERENCES

[1] Reck M, Mellemgaard A, Novello S, Postmus PE, Gaschler-Markefski B, Kaiser R, Buchner H. Change in non-small-cell lung cancer tumor size in patients treated with nintedanib plus docetaxel: analyses from the Phase III LUME-Lung 1 study. OncoTargets and Therapy 2018; 11:4573-4582. [2] Lambin P, Leijenaar RTH, Deist TM, et al. Radiomics: the bridge between medical imaging and personalized medicine. Nature Reviews. Clinical Oncology. 2017. 14(12):749-762. [3] Limkin EJ, Sun R, Dercle L, et al. Promises and challenges for the implementation of computational medical imaging (radiomics) in oncology. Annals of Oncology 2017. 28:1191-1206.

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