

Identification of early longitudinal prognostic predictors of efficacy in lung cancer patients: A coupled tumour dynamics & C-reactive protein model

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C-reactive protein

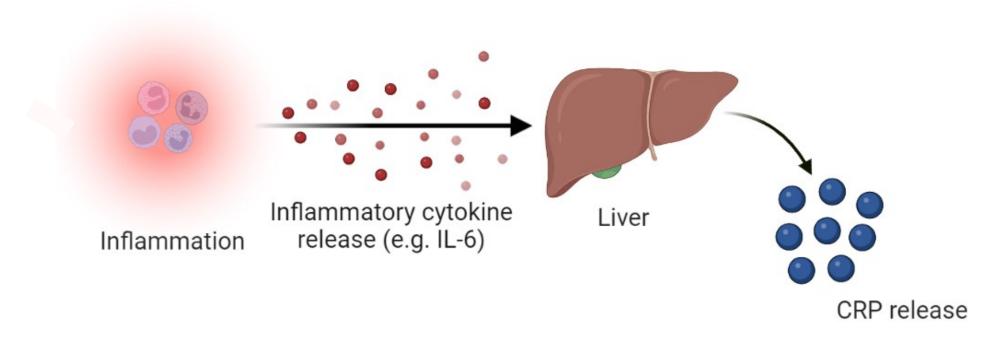


Figure created in Biorender.com

- Non-specific marker of inflammation, infection or tissue injury^{1,2}
- Metric of inflammatory response^{1,2}

¹ Ansar and Ghosh (2013) : C-reactive protein ² Pepys and Hirschfield (2003)

IL-6 : Interleukin 6





C-reactive protein

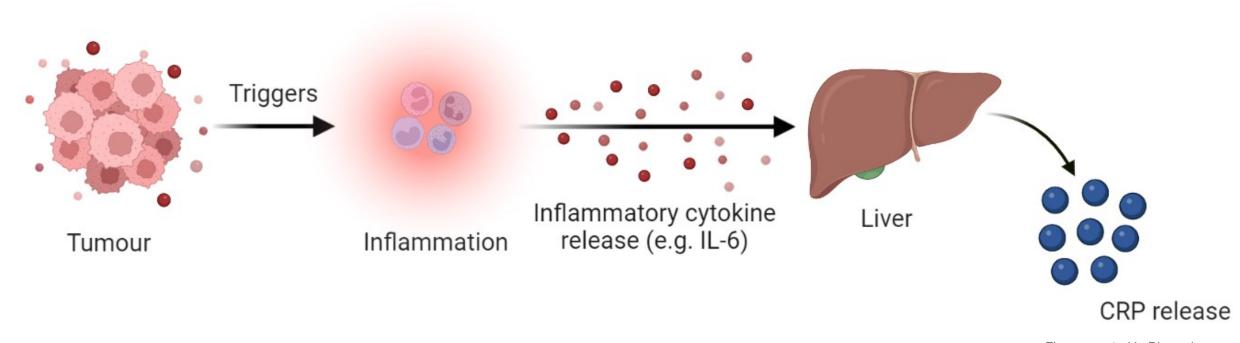


Figure created in Biorender.com

- Associated with advanced cancer stages, metastasis and poor prognosis^{1,2}
 - CRP reflects severity of tissue damage associated with cancer growth and progression

¹ Heikkilä et al. (2007) ² Hart *et al.* (2020)

: C-reactive protein : Interleukin 6

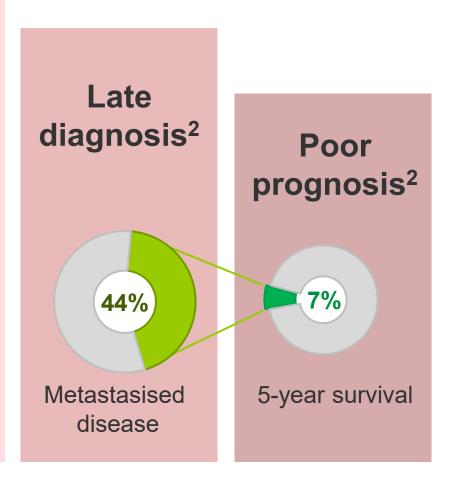




Lung cancer

High disease burden¹

- 2nd most common cancer type (2.2 M)
- 1st leading cause of cancer-related death (1.8 M)





¹ Data source: GLOBOCAN 2020 (http://gco.iarc.fr/) [Accessed 6 April 2023]

² State of lung cancer, American Lung Association 2022 report



Motivation for our research

- Need prognostic markers to:
 - Predict response to NSCLC treatment
 - Identify patients at risk of poor prognosis

Knowledge gap

Focus on baseline biomarker concentration rather than longitudinal concentration



Objective

Identify early prognostic predictors of efficacy in lung cancer patients

Focus on longitudinal data and CRP as inflammatory marker
Would modulating inflammation be a good prognostic factor?

Would monitoring inflammation reflect disease outcome?

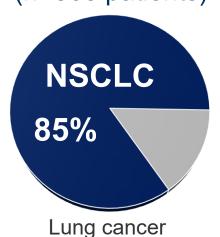


Data source: CEPAC-TDM study





Study population (n=365 patients)

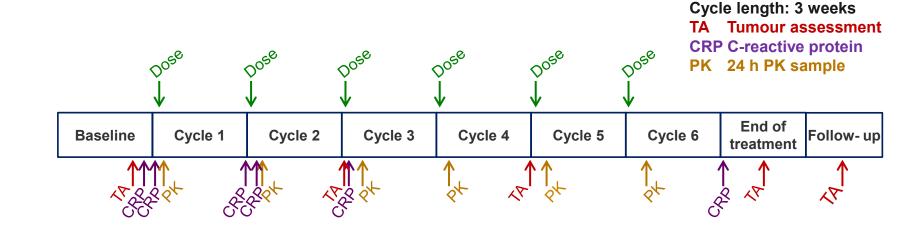


Advanced NSCLC (Stage IIIB, IV)

1st-line treatment
Paclitaxel +
carboplatin/cisplatin

M. Joerger et al. (2016)

Study design



Two paclitaxel treatment arms

Standard dosing (200 mg/m²) Individualised dosing (additional PK samples)

NSCLC: Non-small cell lung cancer

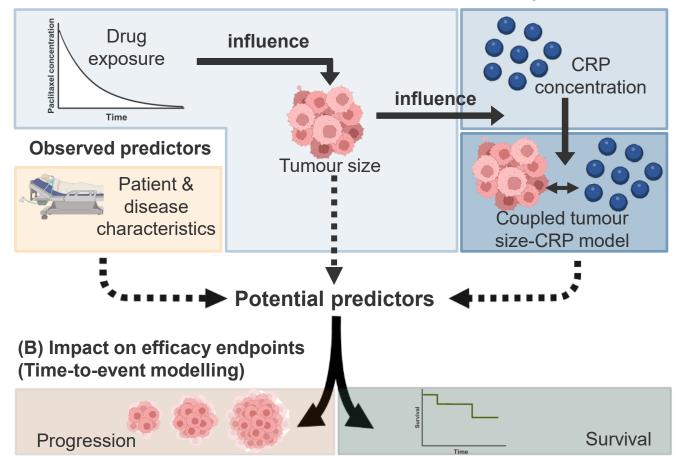
CESAR Study of PAClitaxel Therapeutic Drug Monitoring [EUDRACT 2010-023688-16]





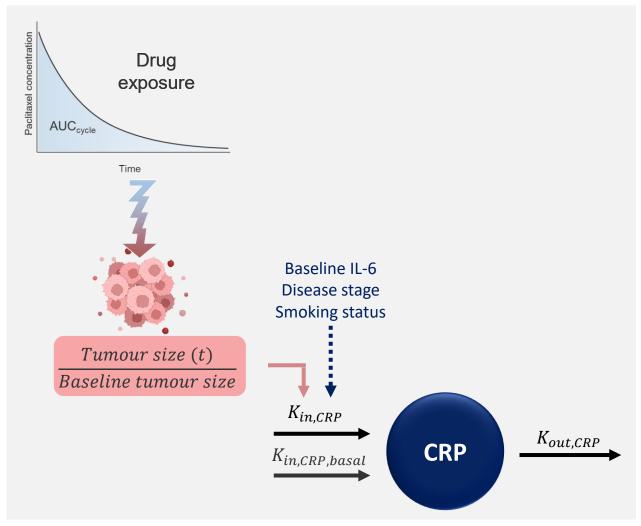
Modelling framework

(A) Mechanistic understanding of relationship between drug exposure, tumour size and CRP concentrations, and model-derived predictors



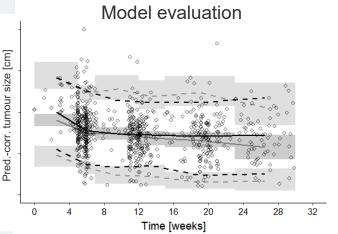






Characterisation of tumour dynamics

- Previously developed tumour growth inhibition model¹ (n=365 patients)
- Chemotherapyinduced tumour dynamics model
 - Drug exposure
 - Resistance
- Derived individual tumour sizes



 $K_{in.CRP}$: Zero-order production rate constant

 $K_{in,CRP,hasal}$: Zero-order basal production rate constant

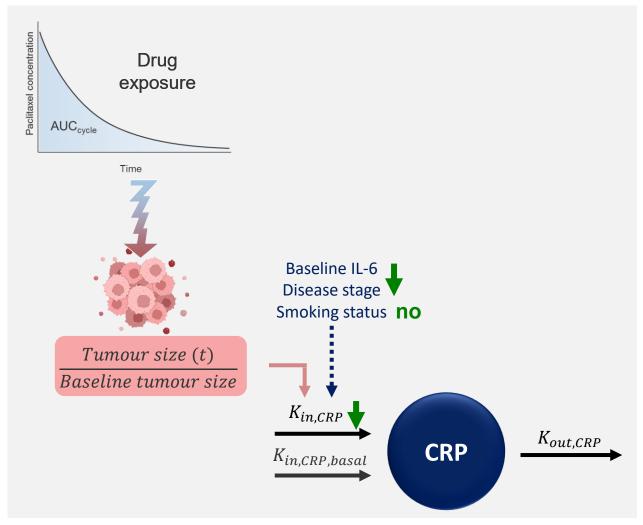
 $K_{out,CRP}$: First-order degradation rate constant

CRP: C-reactive protein IL-6: Interleukin 6



¹ Ojara *et al.* (2023)





Characterisation of tumour dynamics

- Previously developed tumour growth inhibition model¹ (n=365 patients)
- Chemotherapyinduced tumour dynamics model
 - Drug exposure
 - Resistance
- Derived individual tumour sizes

Characterisation of **CRP**

- CRP turnover model (n=257 patients)
- Identify impacting factors on CRP synthesis

: Zero-order production rate constant $K_{in.CRP}$

: Zero-order basal production rate constant $K_{in,CRP,basal}$ $K_{out,CRP}$

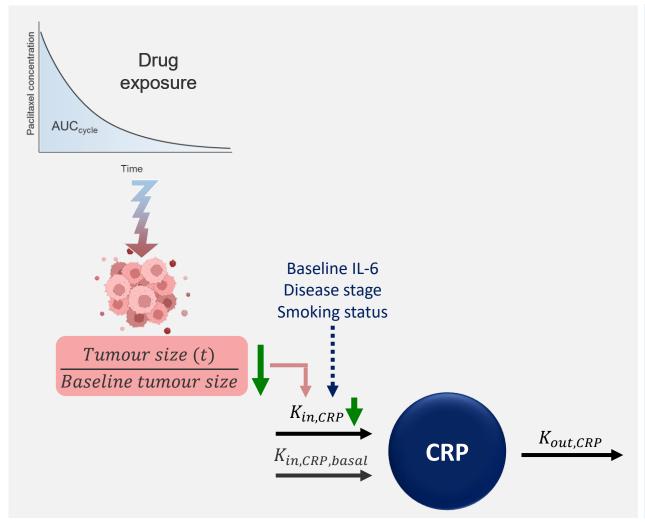
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¹ Ojara *et al.* (2023)





Characterisation of tumour dynamics

- Previously developed tumour growth inhibition model¹ (n=365 patients)
- Chemotherapyinduced tumour dynamics model
 - Drug exposure
 - Resistance
- Derived individual tumour sizes

Characterisation of CRP

- CRP turnover model (n=257 patients)
- Identify impacting factors on CRP synthesis

Linear relationship
between fold-change
in tumour size relative
to baseline and
CRP synthesis

 $K_{in.CRP}$: Zero-order production rate constant

 $K_{in,CRP,basal}$: Zero-order basal production rate constant

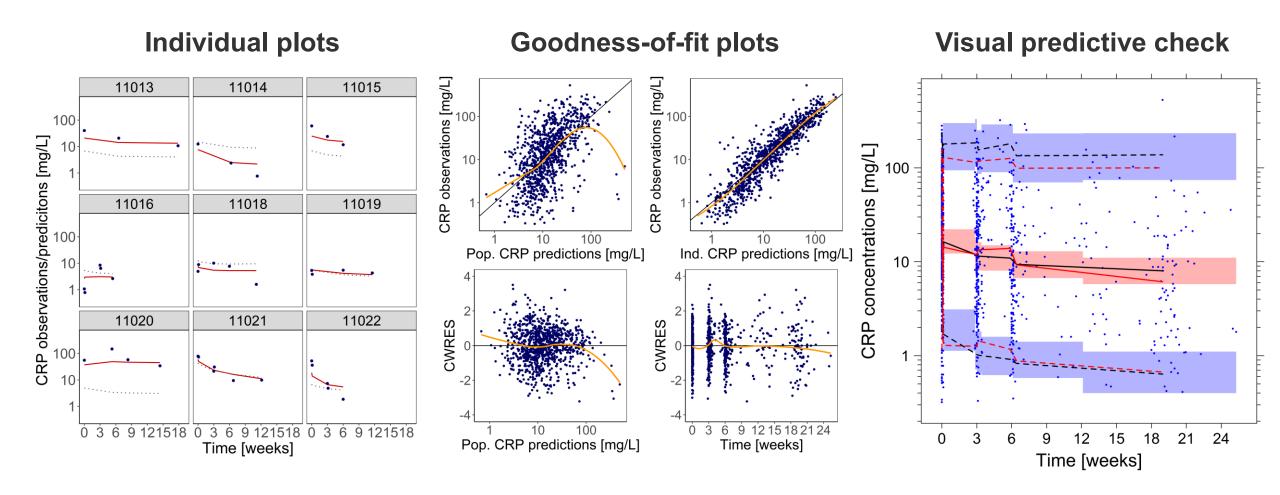
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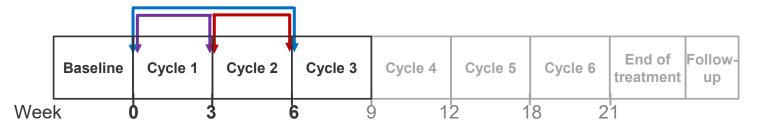
CWRES: Conditional weighted residuals





Selection of predictors

Focus on early metrics (first 3 cycles)



- Selected predictors:
 - 1. CRP-related metrics, neutrophil to lymphocyte ratio *→ markers of inflammation*



CRP-related metrics

- Observed baseline
- Estimated: cycle 1,2,3
- Relative change
 Absolute change
 Fold change
 cycle 2 from 1
 cycle 3 from 1
 cycle 3 from 2



Neutrophil to lymphocyte ratio

- Observed: cycle 1, 2
- Relative change
- Absolute change
- Fold change

CLINICAL PHARMACY

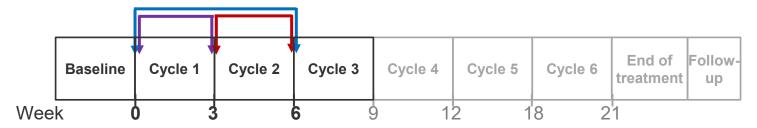
cycle 2 from 1

CRP: C-reactive protein



Selection of predictors

Focus on early metrics (first 3 cycles)



- Selected predictors:
 - 1. CRP-related metrics, neutrophil to lymphocyte ratio *→markers of inflammation*
 - 2. Tumour size-related metrics



Tumour size-related metrics

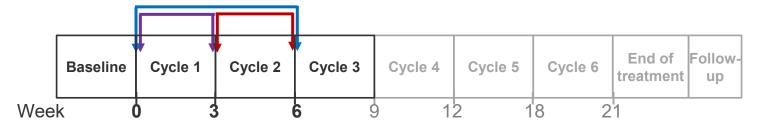
- Observed baseline
- Tumour growth rate
- Fold change in tumour size on $K_{in,CRP}$
- Estimated tumour size relative to baseline tumour size:
 - Week 8
 - Week 7





Selection of predictors

Focus on early metrics (first 3 cycles)



- Selected predictors:
 - CRP-related metrics, neutrophil to lymphocyte ratio → markers of inflammation
 - 2. Tumour size-related metrics
 - Patient/disease characteristics



Patient and disease characteristics

Baseline ECOG

Smoking status

Liver metastasis

Brain metastasis

Disease stage

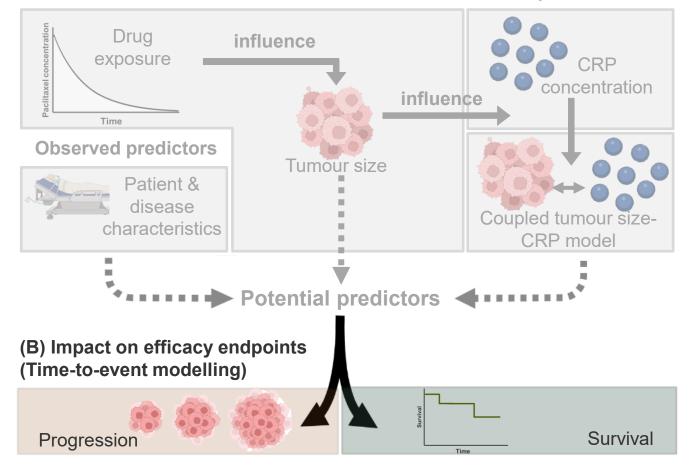
CRP: C-reactive protein

ECOG: Eastern Cooperative Oncology Group



Modelling framework

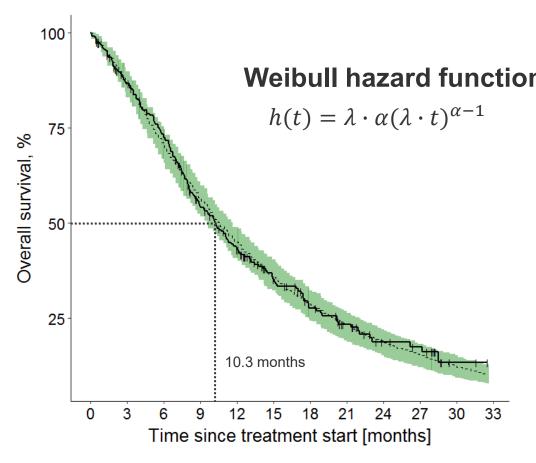
(A)Mechanistic understanding of relationship between drug exposure, tumour size and CRP concentrations, and model-derived predictors







Time-to-event modelling of overall survival and identification of significant predictors



Weibull hazard function Identified significant predictors

- · Inflammatory level at cycle 3 (CRP_{cycle3})
- **Tumour load** (Baseline tumour size)
- · Tumour shrinkage (RS8)
- · Liver lesions

Solid line : Observed survival data

Dashed line : Median model predicted profile

Dotted line : Median overall survival

Vertical lines : Censoring

Green shade: 90% confidence interval

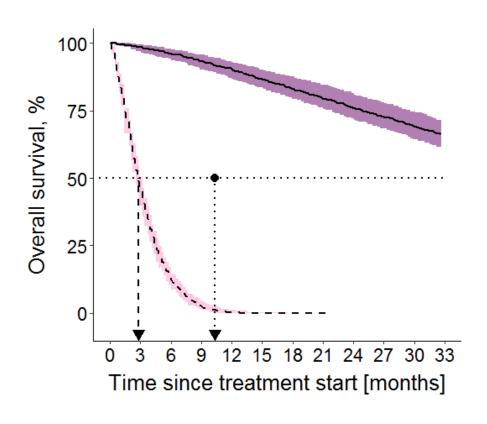
 λ : Scale parameter α : Shape parameter CRP : C-reactive protein

RS8: Tumour size at week 8 relative to baseline tumour size





Time-to-event modelling of overall survival and identification of significant predictors



Identified significant predictors

· Inflammatory level at cycle 3 (CRP_{cycle3})



· **Tumour load** (Baseline tumour size) **♦**



· Tumour shrinkage (RS8) **♦**



· Liver lesions no yes

Solid line : Observed survival data

: Median model predicted profile Dashed line : Median overall survival

Vertical lines : Censoring

Dotted line

Green shade: 90% confidence interval

Predictor: 5th percentile/no Predictor: 95th percentile/yes Observed median overall survival

: Scale parameter : Shape parameter : C-reactive protein

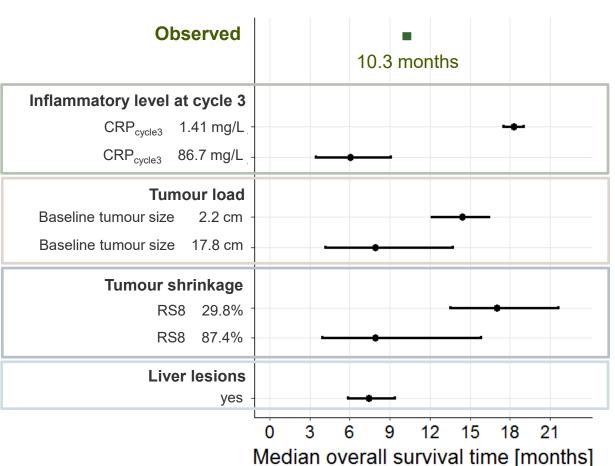
: Tumour size at week 8 relative to baseline tumour size



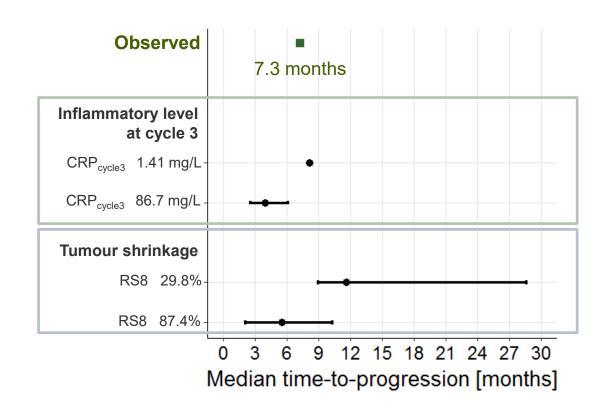


Impact of significant predictors

Overall survival



Time-to-progression



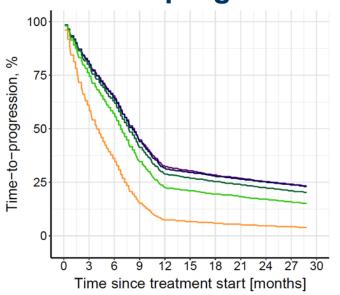
CRP: C-reactive protein

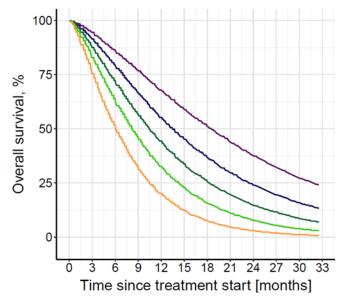
RS8: Tumour size at week 8 relative to baseline tumour size





Simulation-based impact of *inflammation* level (CRP_{cycle3}) on time-to-progression and survival





Model-based simulation

CRP _{cycle3} concentration	(percentile)
1.41 mg/L	(P _{0.05})
4.37 mg/L	$(P_{0.25})$
——— 11.1 mg/L	$(P_{0.50})$
28.8 mg/L	$(P_{0.75})$
86.7 mg/L	$(P_{0.95})$

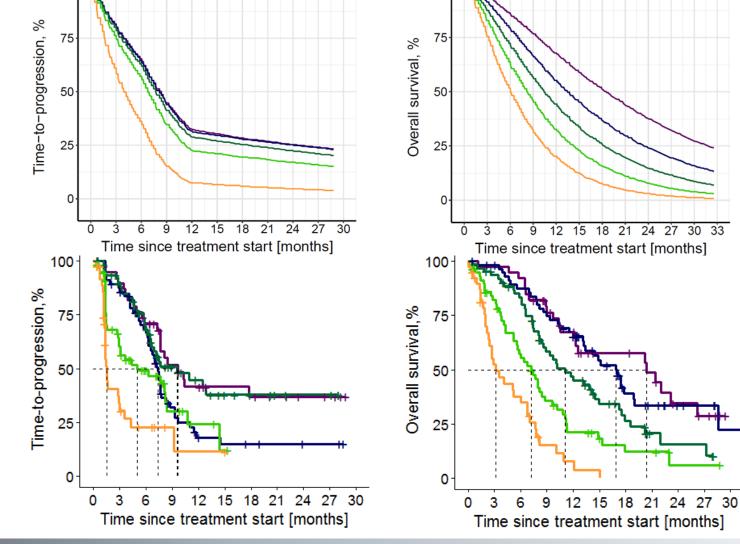
- At minimal inflammation level ~5 mg/L
 - No impact on risk of progression
- At higher inflammation level ~ >30 mg/L
 - Worse outcome
- Threshold: ~11 mg/L reached at cycle 3

- Regular pattern
- No threshold





Simulation-based impact of *inflammation* level (CRP_{cycle3}) on time-to-progression and survival against observed events



Model-based simulation

CRP _{cycle3} concentration	(percentile)
1.41 mg/L	(P _{0.05})
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——— 11.1 mg/L	$(P_{0.50})$
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86.7 mg/L	$(P_{0.95})$

Observed events

Model-estimated CRP_{cycle3} concentration

CRP: C-reactive protein



(percentile interval)

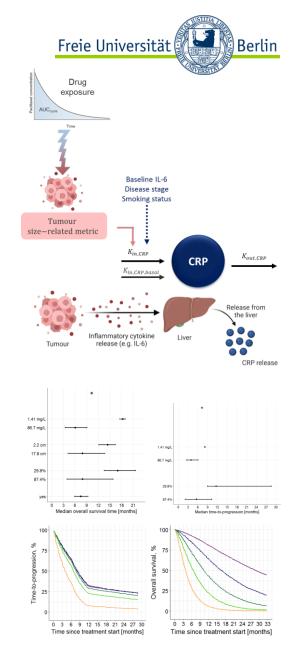
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Summary

- Developed framework incorporating relations between drug exposure, tumour dynamics and biomarker
 - Characterised circulating CRP
 - ➤ **Linked** CRP (*marker of inflammation*) and tumour size (*marker of disease aggressiveness*)
 - Allowed estimation of longitudinal predictors for all patients.
- Early longitudinal prognostic variables identified:
 - Progression and overall survival in NSCLC patients
 - ➤ Non-baseline CRP:
 - > more reflective of patient status & disease status
 - > stronger predictor compared with baseline concentration
 - ➤ Although CRP is a non-specific biomarker: holds a **strong** prognostic value as predictor of efficacy
 - Patient's inflammatory status (CRP concentration) reflects risk of progression and survival → potential prognostic marker

CRP :C-reactive protein

NSCLC :Non-small cell lung cancer





Conclusion & perspectives

- Identified patients at higher risk of early progression/short survival
 - Potential for early intervention
- Longitudinal biomarker data should be further exploited and chosen over baseline data for disease- and patient-related predictions



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