



Towards understanding anti-infliximab antibody development to predict Crohn's disease patients' underlying immunogenicity status

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Immunogenicity and Anti-Drug Antibodies
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Overview



Introduction

Anti-drug antibodies leveraged data

Modelling immunogenicity

Next steps & conclusion

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Modelling immunogenicity

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Infliximab in Crohn's Disease patients

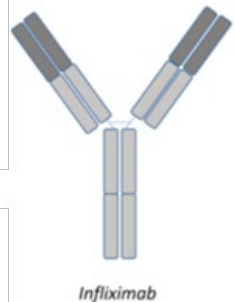


Crohn's Disease: Inflammatory bowel disease

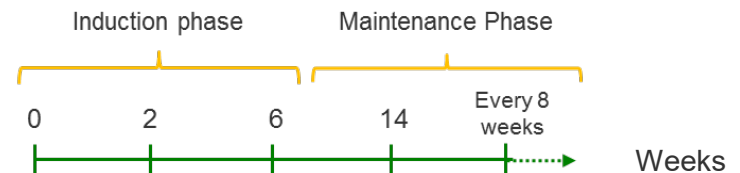
- Chronic inflammation of the gut leading to mucosal impairment

Infliximab

- Anti-tumour necrosis factor α (TNF α)
- Immunosuppressive and anti-inflammatory effect
- Indications: **Crohn's disease** and other inflammatory diseases.



- Administration of infliximab:
5 mg/kg, i.v. infusion



- Challenge of infliximab therapy: Loss of response



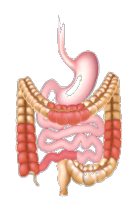
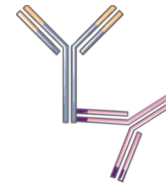
How to predict/prevent?

Loss of response (LOR)

- 50% patients fail to respond to treatment (LOR) in the first year
 - Dosing regimen intensification or change in therapy (different mAb)
- LOR related to sub-therapeutic drug concentrations

Increase CL

- High baseline TNF α
- High baseline C-reactive protein (CRP)
- Low serum albumin
- High body weight
- Sex (male)
- **Anti-drug antibodies (ADA)**



ADA: Anti-drug antibody
 CRP: C-Reactive protein
 LOR: Loss of response

mAb: Monoclonal antibody
 TNF α : Tumour necrosis factor α

Ryman J.T. (2017).
 Adegbola S.O. et al (2018)
 Keizer RJ et al. (2010)

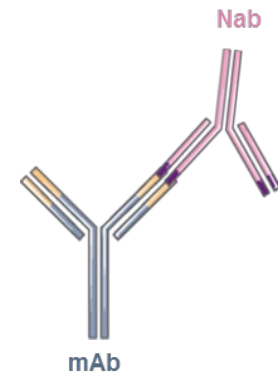
Anti-Drug Antibodies (ADA)

- Monoclonal antibodies (mAb) therapy may provoke an immunogenic reaction:
→ Formation of anti-drug antibodies (ADA) against therapeutic mAb

- Multiple species:

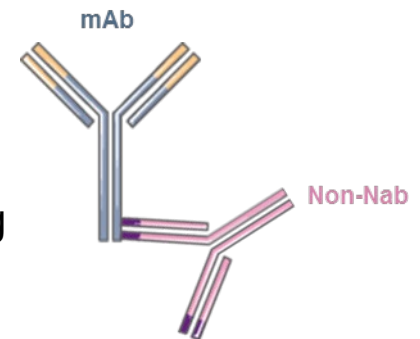
- **Neutralising antibodies (Nab):**

- Binding to complementarity determining regions
→ Prevents binding of IFX to target (TNF α)
 - Level of neutralisation dependent on titer of Nab



- **Non-neutralising antibodies (Non-Nab):**

- Do not interfere with antigen-binding capacity of mAb
 - Formation of circulating ADA-mAb immune complexes creating an additional elimination pathway
→ Increase elimination of mAb

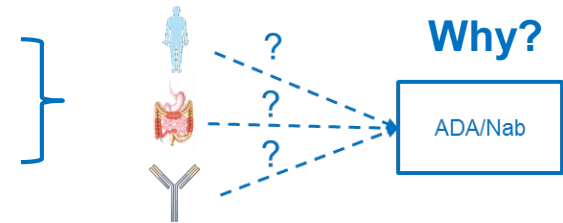


Motivation and objective

- Presence of ADA related to IFX low concentration.
→ ADA increase risk of treatment failure (LOR)
- Nab decrease efficacy by preventing IFX binding to its therapeutic target, TNF α .
→ Nab impact PK and PD : higher risk of LOR
- ADA and Nab development depends on different patients-/disease-/drug-related factors
- ADA often looked at as status, covariate effect on CL
- Nab sub-entity rarely taken into account in modelling activities

Need to know early on which patients are at risk of ADA/Nab development

→ Prevent LOR



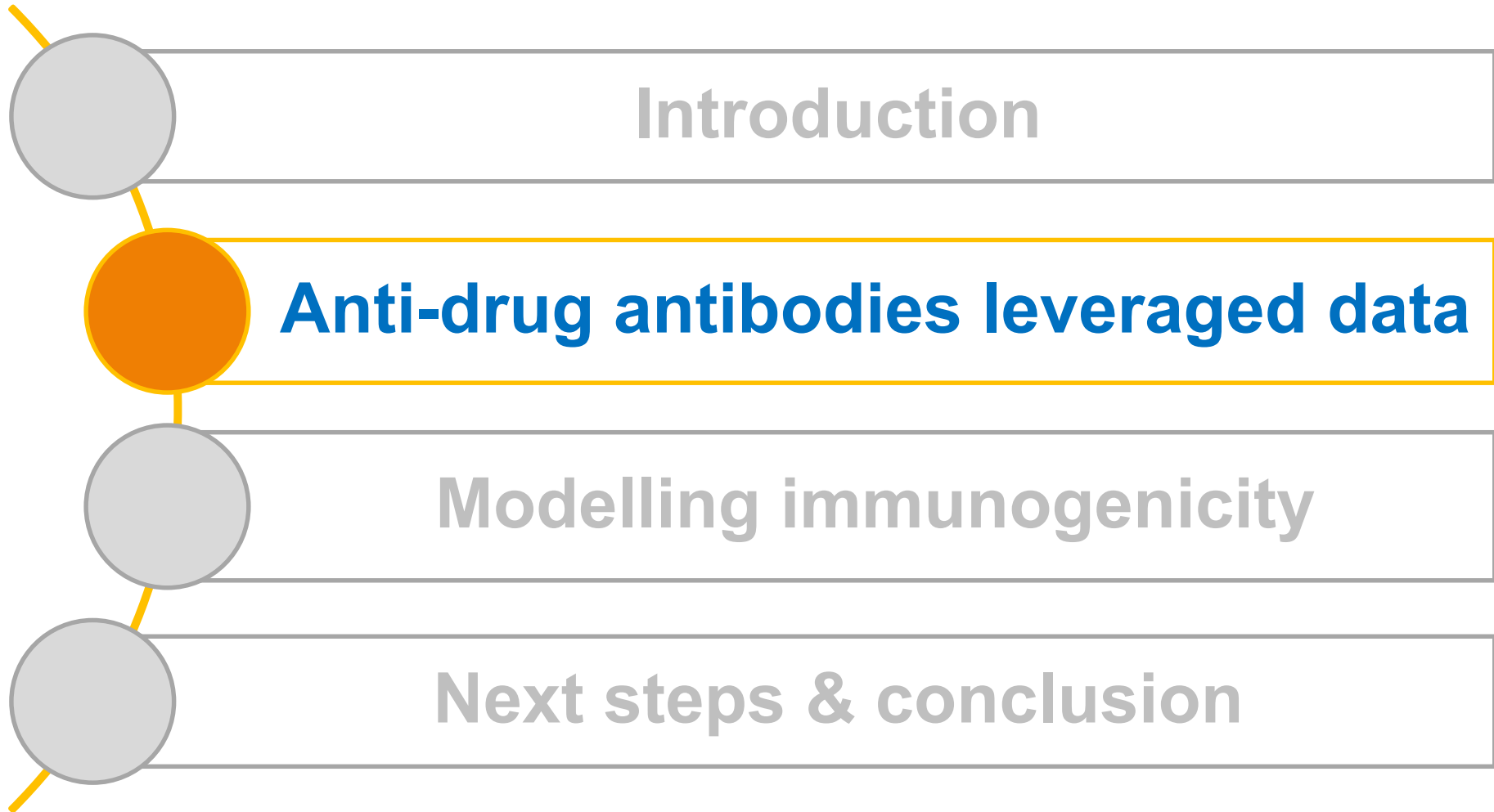
Need for a deeper understanding of ADA and Nab dynamics

→ How to predict underlying immunogenicity in CD patients receiving IFX?

ADA: Anti-drug antibodies
IFX: Infliximab
LOR: Loss of response

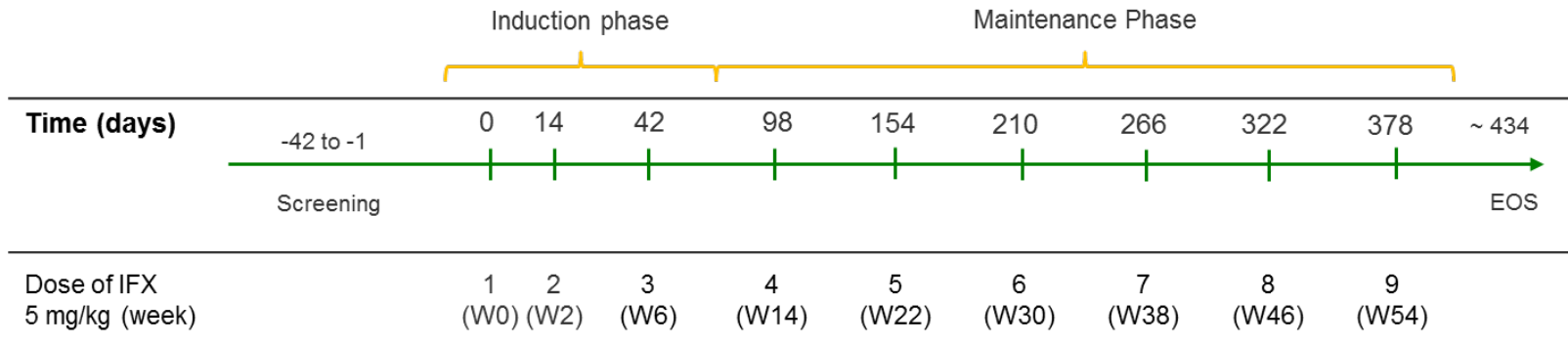
Nab: Neutralising ADA
PD: Pharmacodynamic
PK: Pharmacokinetic

Overview



Study schedule

- 220 CD patients receiving IFX therapy, all naïve to mAbs

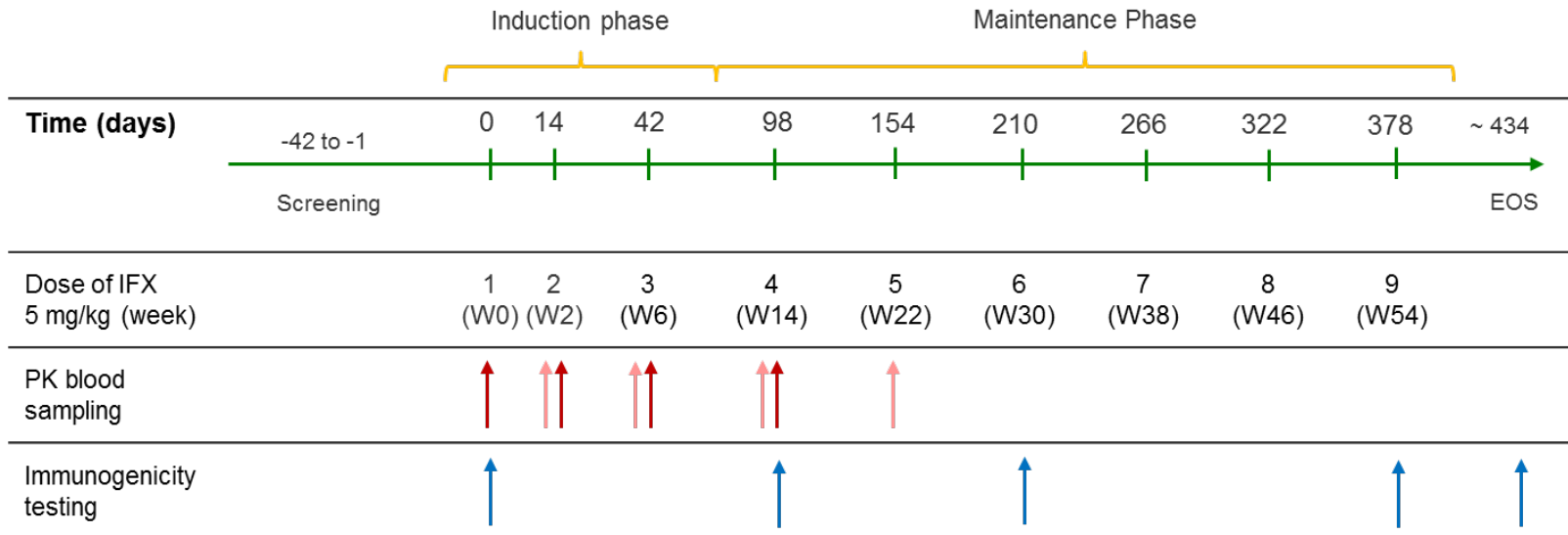


CD: Crohn's disease
 EOS: End of study
 IFX: Infliximab
 mAbs: Monoclonal antibodies



Study schedule

- 220 CD patients receiving IFX therapy, all naïve to mAbs



→ Both ADA and Nab are reported as titers

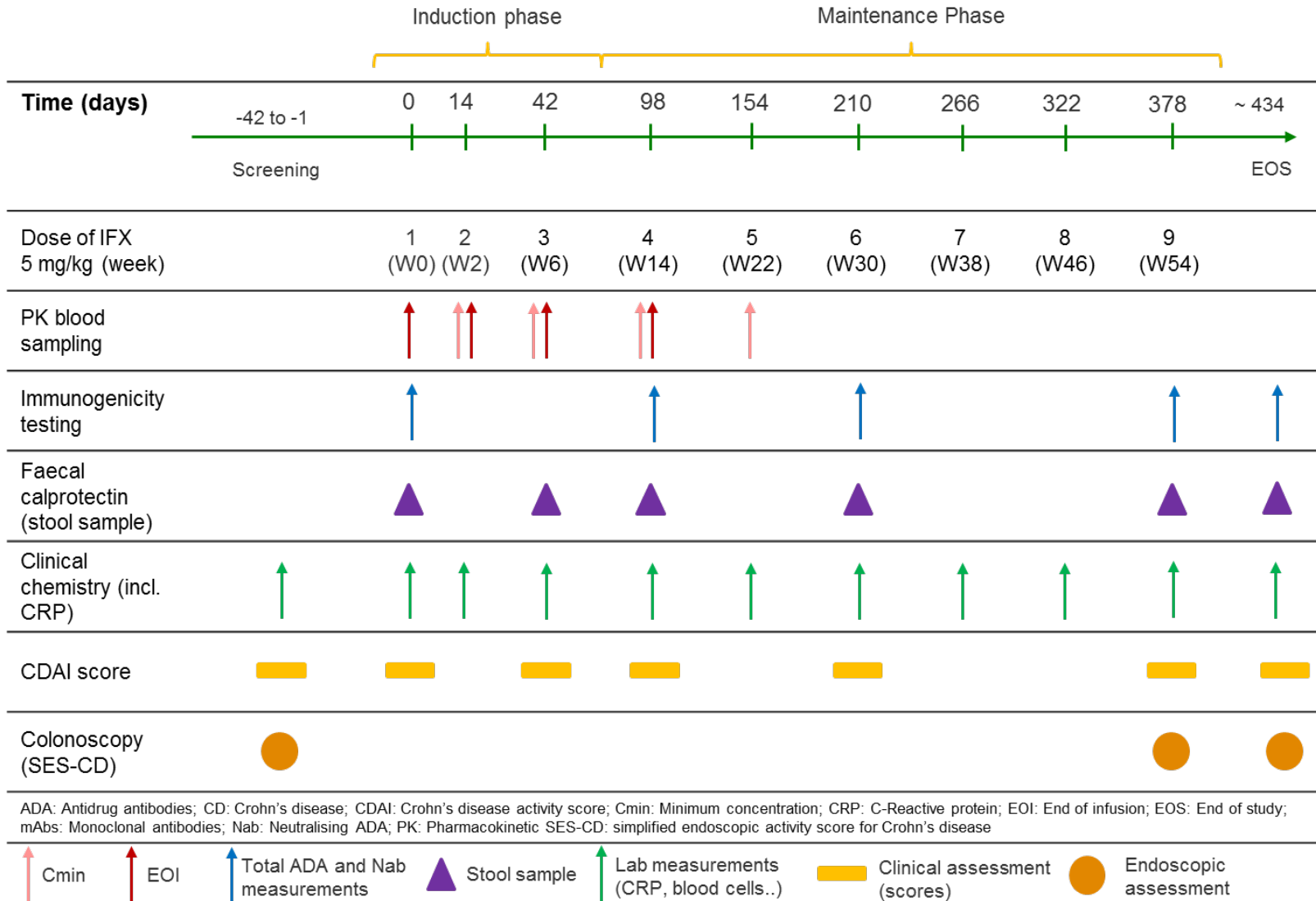
↑ Cmin ↑ EOI ↑ Total ADA and Nab measurements

CD: Crohn's disease
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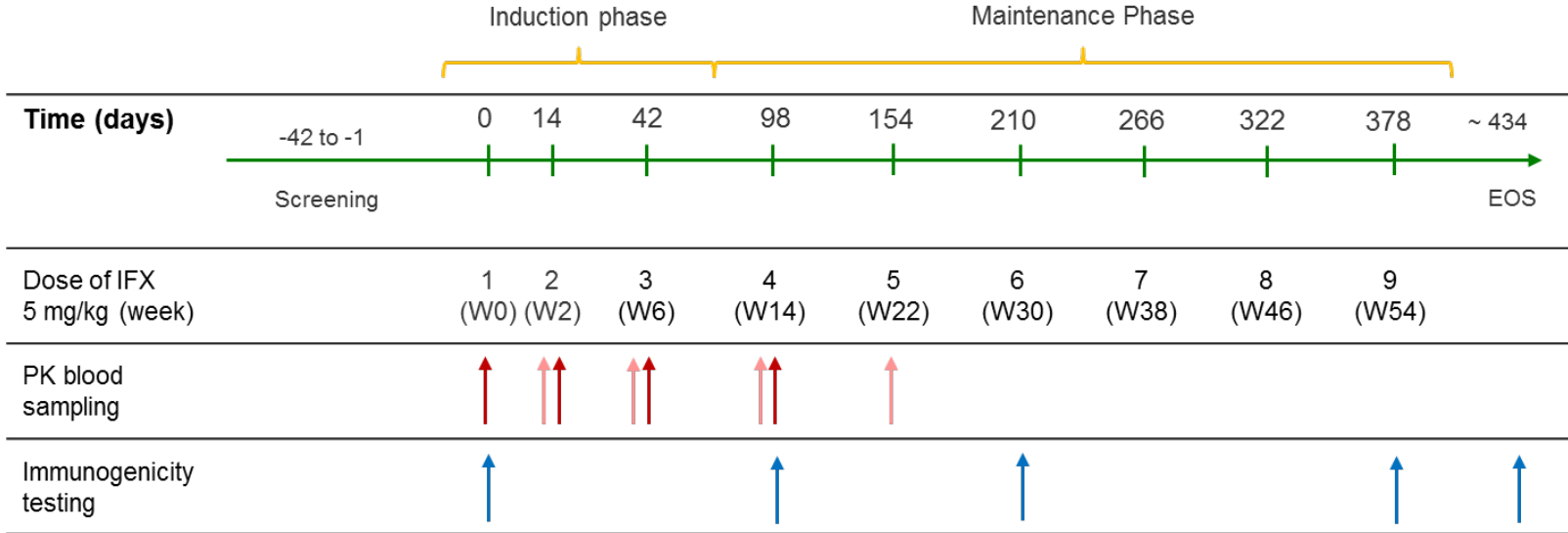
Study schedule

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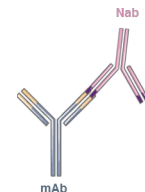
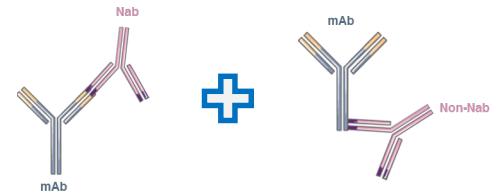
Study schedule

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ADA testing strategy

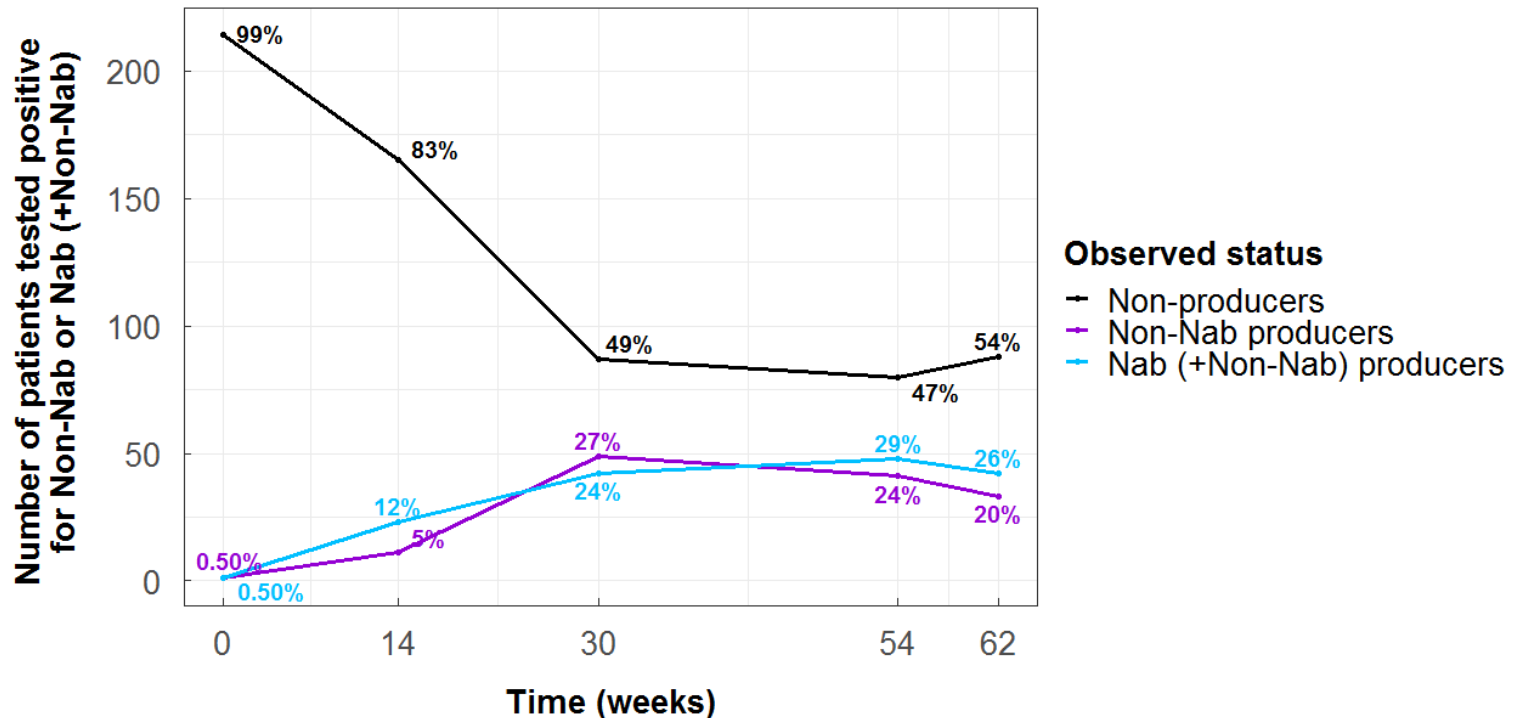
- Patients tested for ADA: No differentiation between Nab and non-Nab
 - Assay: ELISA (enzyme-linked immunosorbent drug-tolerant assay)
 - If positive: patients labelled “**ADA positive**”
 - Total ADA (non-Nab and/or Nab) measured: “**Total ADA titers**”
- Patient tested for Nab:
 - Assay: ECL (electrochemiluminescent) immunoassay
 - If positive: patient labelled “**Nab positive**”
 - Nab measured: “**Nab titers**”



Occurrence of anti-drug antibodies

- Patients with **ADA** during the study: **125/220 (57%)**
- Patients with **Nab** during the study: **78/220 (34%)**

%Nab+/ADA+: **50%** **68%** **48%** **53%** **61%**

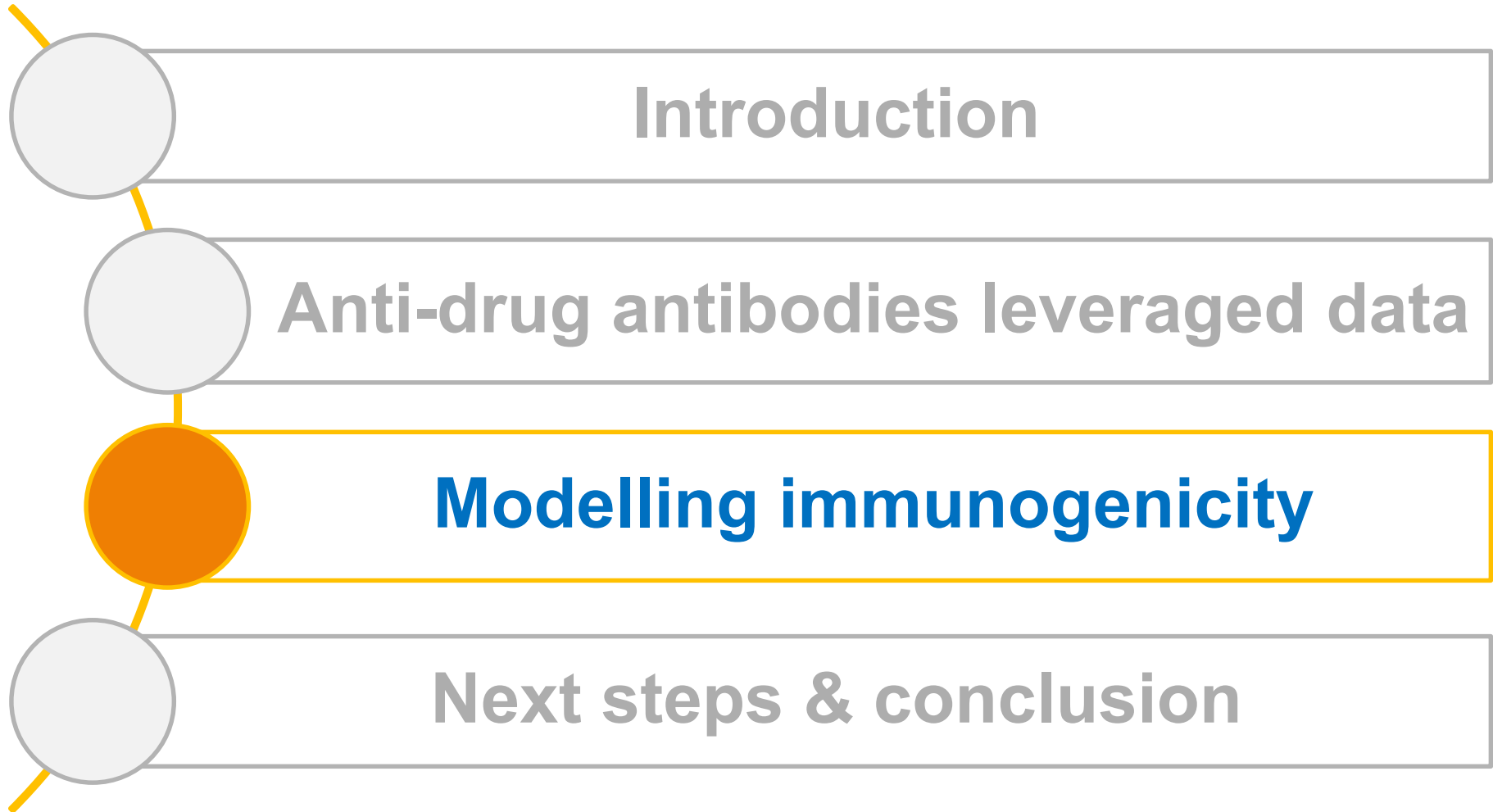


→ %ADA+ and %Nab+ stabilised after week 30

→ %Nab+/ADA+ remains ~50% across the study

ADA: Anti-drug antibody
Nab: Neutralising ADA

Overview



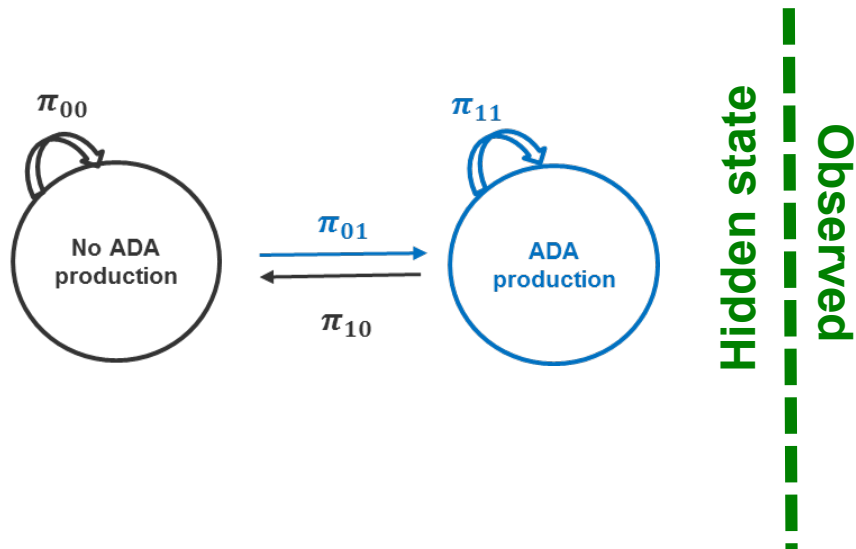
Mixed hidden Markov model (MHMM)

Describe a relationship between two stochastic processes:

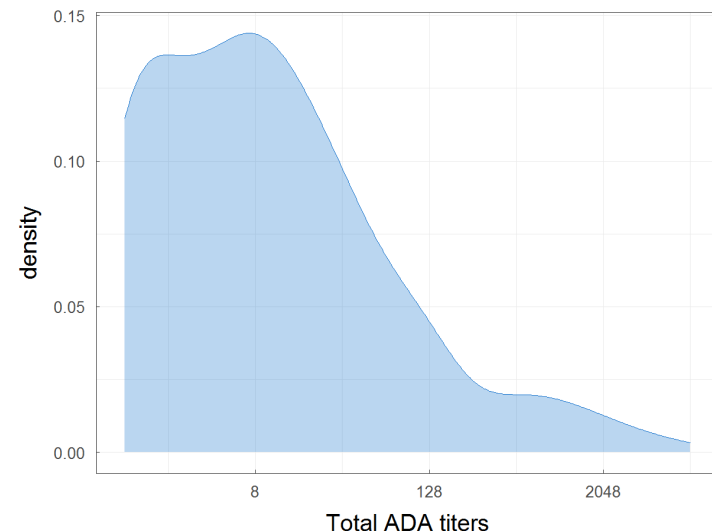
Hidden process → Underlying immunogenicity status

Observed process → ADA titers: most informative for prediction of immunogenicity status

Markov model



Characterised by its distribution

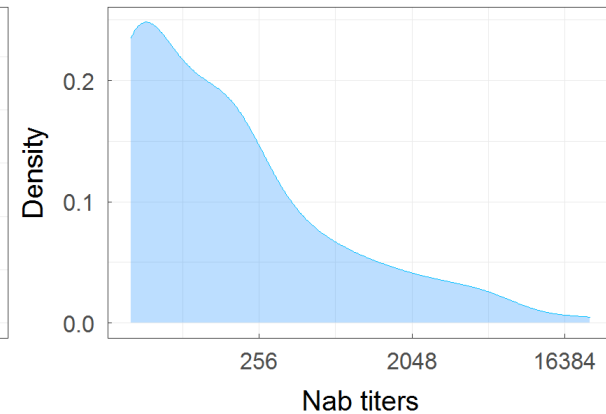
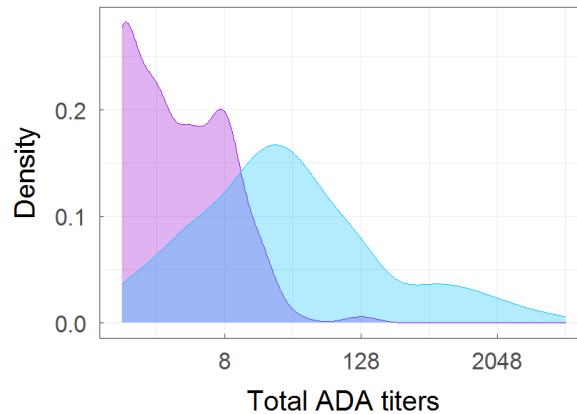
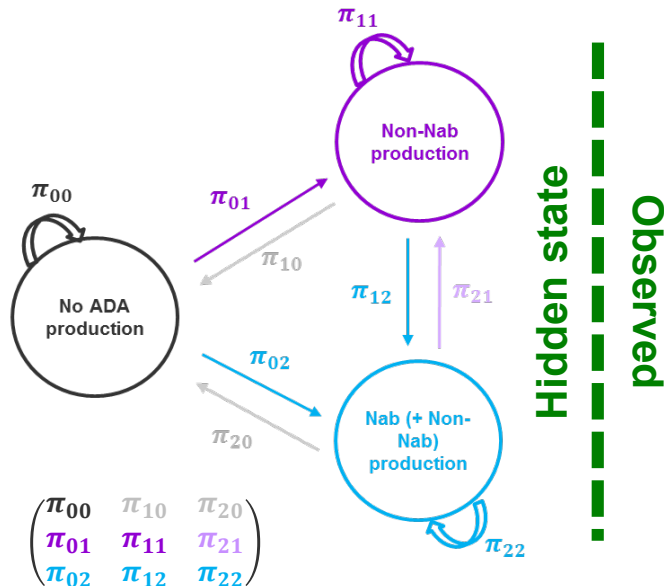


Model development strategy

1. Choice of hidden states
2. Choice of observed variables and the distribution to describe them

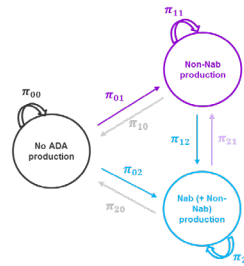
Hidden states	Observed variables	Distribution of the observed variables
3 states <ul style="list-style-type: none"> - No ADA production - Non-Nab production - Nab + Non-Nab production 	<ul style="list-style-type: none"> - Total ADA titers - Nab titers 	<ul style="list-style-type: none"> - ZTP distributions $P(X = k X > 0) = \frac{\lambda^k}{(e^\lambda - 1)k!}$

ADA: Anti-drug antibodies; Nab: Neutralising ADA; π : Transition probability; ZTP: Zero-truncated Poisson

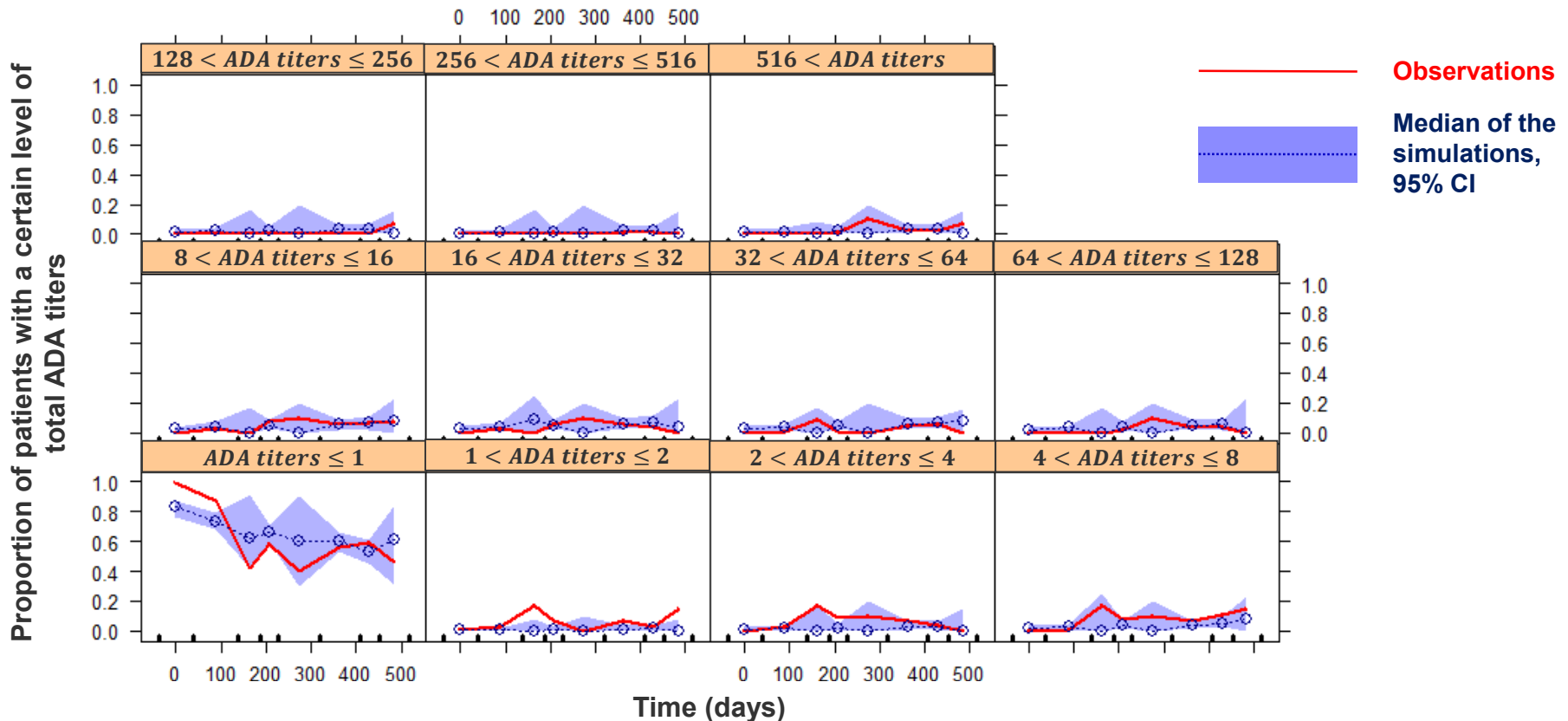
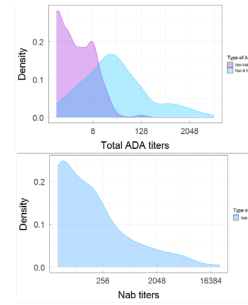


Model evaluation

- Three hidden states
- Two observed variable



Hidden state



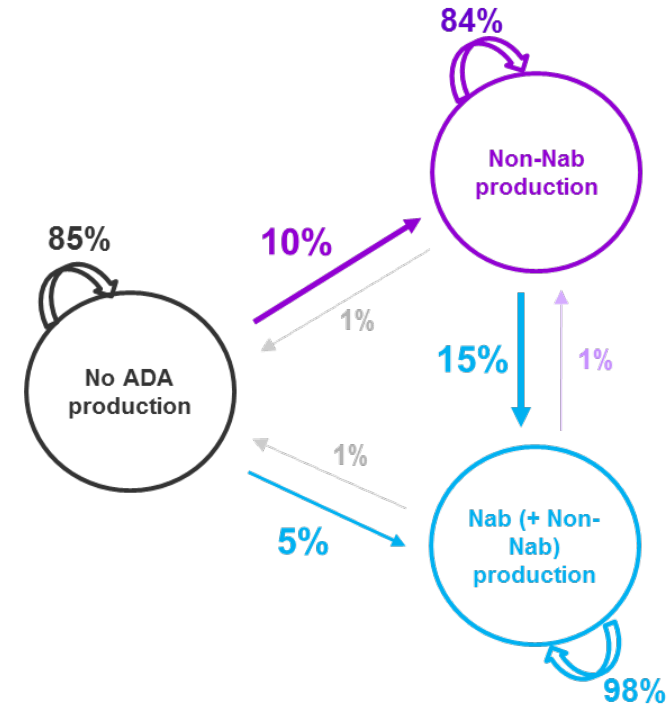
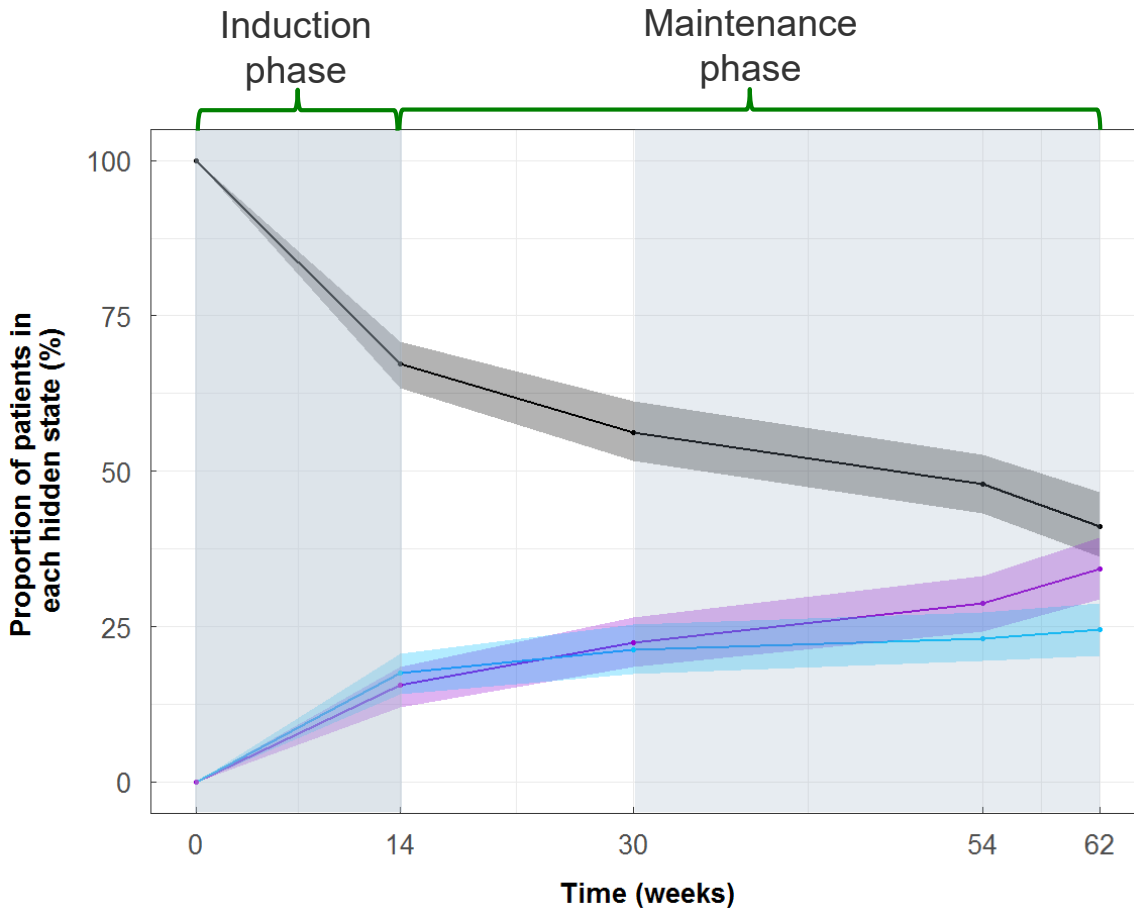
ADA: Anti-drug antibody
 Nab: Neutralising ADA
 π : Transition probability

Model results

Hidden states

- █ Non-producers
- █ Non-Nab producers
- █ Nab (+Non-Nab) producers

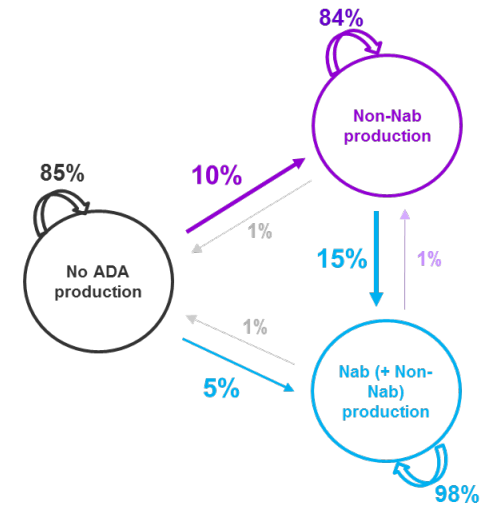
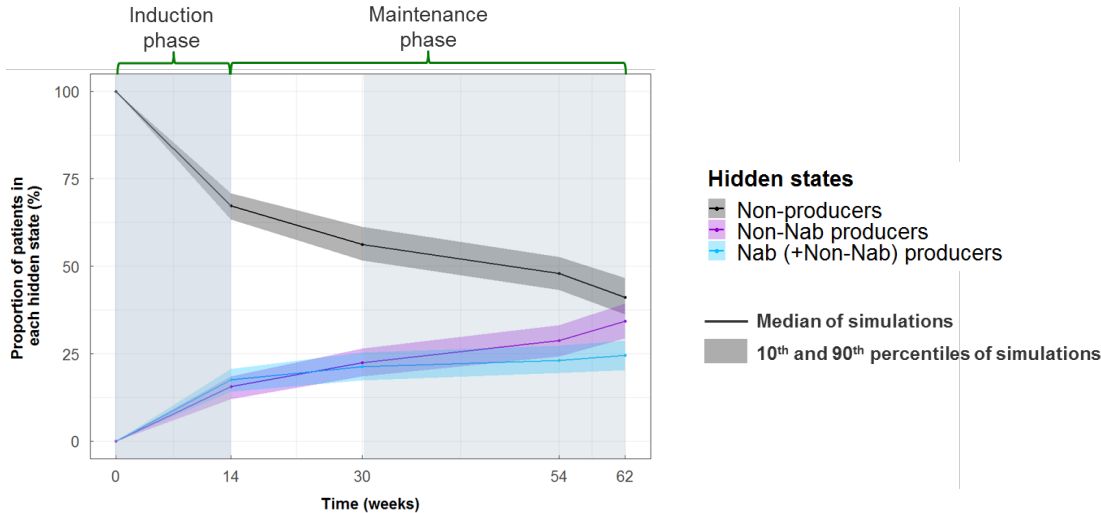
- Median of simulations
- 10th and 90th percentiles of simulations



ADA: Anti-drug antibody
 Nab: Neutralising ADA
 π : Transition probability

Model results

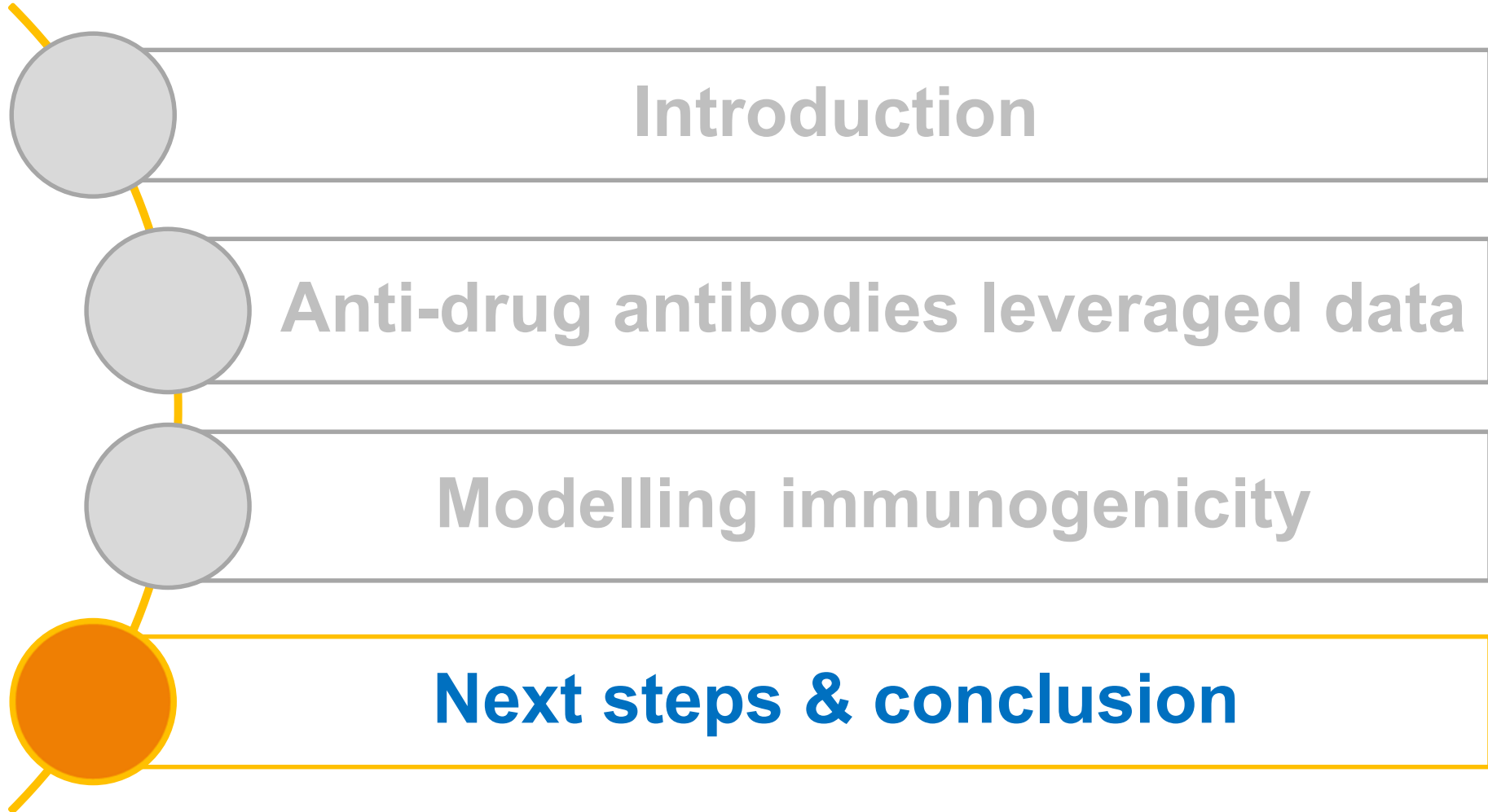
- Three hidden states
- Two observed variable



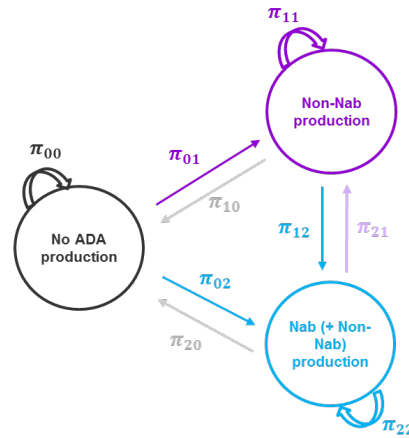
- 1/3 patients develop ADA during induction phase
- Stabilisation of ADA development at week 30 → Decreased risk of developing ADA
- ~ 50% patients present an immunogenic response
 - ~ 25% **Non-Nab**: ↑ risk of LOR due to PK
 - ~ 25% **Nab**: ↑ risk of LOR due to PK and PD

ADA: Anti-drug antibody
 Nab: Neutralising ADA
 π : Transition probability

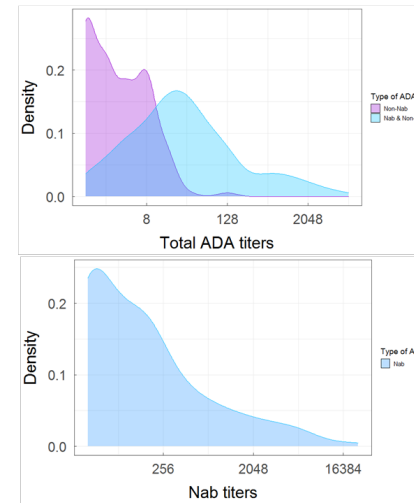
Overview



Next steps



Hidden state
Observed

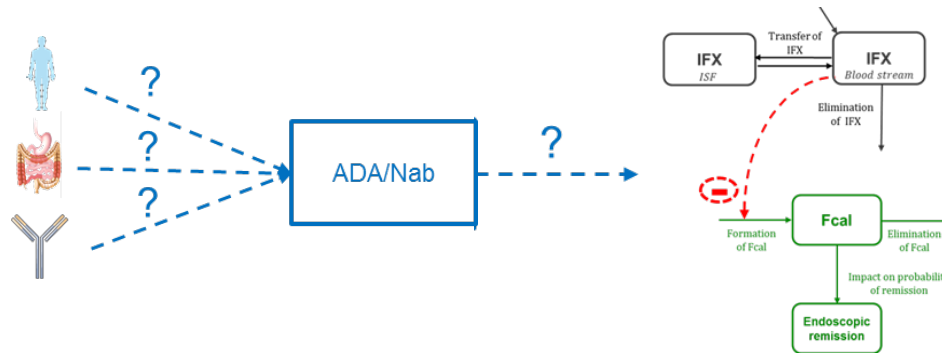


❖ ADA-Nab BV MHMM:

- Correlated bivariate ZTP
 - Need to describe the **correlation** between the two observed variables
- Inclusion **time dependency** on transition probabilities:
 - Hypothesis: After a while, risk of developing ADA/Nab decreases
- **Covariate** analysis :
 - Impacting factor on Non-Nab and Nab development?

ADA:	Anti-drug antibody	Nab:	Neutralising ADA
BV:	Bivariate	π :	Transition probability
MHMM:	Mixed hidden Markov model	ZTP:	Zero-truncated Poisson

- **Underlying immunogenicity in CD patients receiving IFX:**
 - Successfully described with **3 states**, comprising **Nab** development
- **MHMM:** comprehensive framework for predicting underlying immunogenicity
 - Predict sequence of unobservable states on an individual level: ADA and Nab development
 - Simulations of further studies possible: sampling of Nab
- **Final objective:** Assess impact of Non-Nab and/or Nab development on PK and PD



➔ **Impact of immunogenicity on treatment outcome: LOR**

ADA: Antidrug-antibodies
 CD: Crohn's disease
 IFX: Infliximab

Nab: Neutralising ADA
 PD: Pharmacodynamics
 PK: Pharmacokinetics

Acknowledgements

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