# Mathematical Modelling of Myeloproliferative Neoplasms and Hematopoietic Stem Cells

An overview of our recent work and a look to the future

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► Who am I?

► Postdoc at the PandemiX Research Center, working on COVID-19 and historical epidemics.



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- ▶ Ph.D. at Roskilde University: Mathematical modelling of blood cancers, with Johnny and Morten.





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► Postdoc at the PandemiX Research Center, working on COVID-19 and historical epidemics.

Ph.D. at Roskilde University: Mathematical modelling of blood cancers, with Johnny and Morten.





► This talk:

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► Who am I?

- Postdoc at the PandemiX Research Center, working on COVID-19 and historical epidemics.
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- ► This talk:
  - Illustration of how mathematical modelling could contribute to clinical practice in the future.

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► Who am I?

- Postdoc at the PandemiX Research Center, working on COVID-19 and historical epidemics.
- Ph.D. at Roskilde University: Mathematical modelling of blood cancers, with Johnny and Morten.





### ► This talk:

- Illustration of how mathematical modelling could contribute to clinical practice in the future.
- Focus on general aspects of mathematical modelling rather than mathematical details.

# Our purpose and goal

Myeloproliferative Neoplasms (MPNs): Group of diseases characterized by overproduction of blood cells.

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Myeloproliferative Neoplasms (MPNs): Group of diseases characterized by overproduction of blood cells. Believed to arise from mutations occurring in hematopoietic stem cells (HSC)

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► Myeloproliferative Neoplasms (MPNs): Group of diseases characterized by overproduction of blood cells. Believed to arise from mutations occurring in hematopoietic stem cells (HSC)

▶ Clinicians from Zealand University Hospital: MPN-patients treated with pegylated interferon- $\alpha$  obtain long-term normalization of cell-counts.

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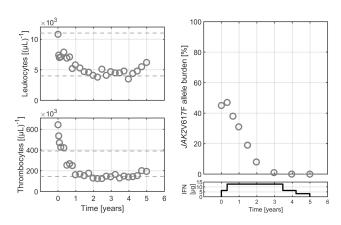
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- ► Myeloproliferative Neoplasms (MPNs): Group of diseases characterized by overproduction of blood cells. Believed to arise from mutations occurring in hematopoietic stem cells (HSC)
- ▶ Clinicians from Zealand University Hospital: MPN-patients treated with pegylated interferon- $\alpha$  obtain long-term normalization of cell-counts.
- ▶ **Data:** Clinical trial with patients treated with interferon- $\alpha$ .

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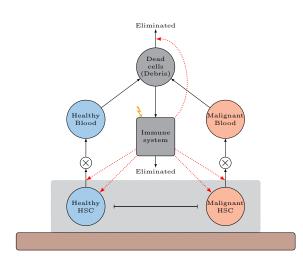
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Myeloproliferative Neoplasms (MPNs): Group of diseases characterized by overproduction of blood cells. Believed to arise from mutations occurring in hematopoietic stem cells (HSC)

- ► Clinicians from Zealand University Hospital: MPN-patients treated with pegylated interferon- $\alpha$ obtain long-term normalization of cell-counts.
- ▶ Data: Clinical trial with patients treated with interferon- $\alpha$ .
- ▶ **Goal:** Can the effects of interferon- $\alpha$  be accurately described by a mathematical model?

## Model description



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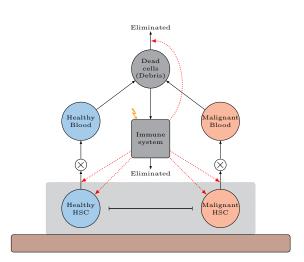
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(Pedersen et al, 2021)

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(Pedersen et al, 2021)

Based on previous model of Andersen et al (2017), combined with a model of HSC developed together with Thomas Stiehl (Roskilde University & RWTH Aachen).

► System of 6 ODEs, with about 16 parameters.

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- System of 6 ODEs, with about 16 parameters.
  - Most parameters determined from the literature.

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▶ Most parameters determined from the literature.

► General model behaviour in agreement with clinical intuition.

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  - ▶ Disease progression over a long period (15+ years).

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  - ▶ Disease progression over a long period (15+ years).
  - ► Chronic inflammation leads to faster disease progression.

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- System of 6 ODEs, with about 16 parameters.
- Most parameters determined from the literature.
- General model behaviour in agreement with clinical intuition.
  - ► No mutations → long-term health.
  - ▶ Disease progression over a long period (15+ years).
  - ► Chronic inflammation leads to faster disease progression.
- We can perturb specific model-parameters that relate to the biological effect of interferon- $\alpha$  treatment.



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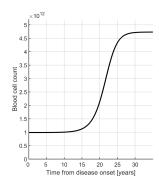
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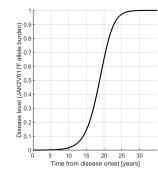
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Typical disease progression in the model.



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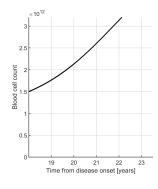
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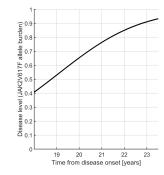
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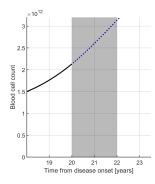
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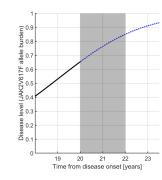
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Zooming in on year 20 after original mutation.





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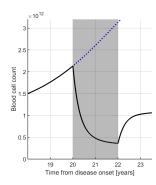
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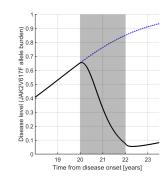
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Considering two years of treatment.





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Considering two years of treatment.

Perturbing model-parameters related to treatment.

# Fitting to data



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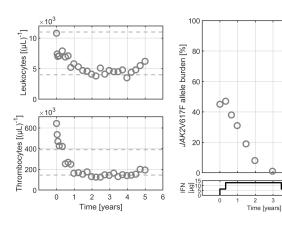
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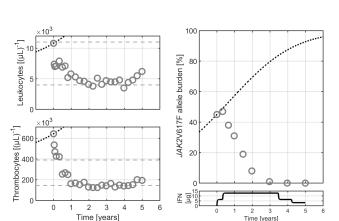
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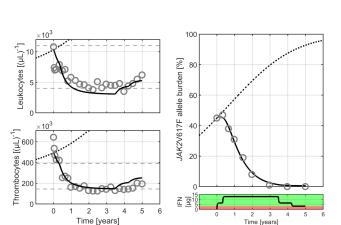
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Adding modelled scenario without treatment, time-shifted to agree with baseline data (and PK/PD-modelling of treatment)

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Dose-dependent fitting of model-parameters

# Halting treatment

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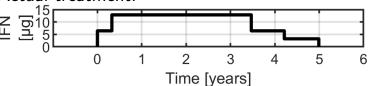
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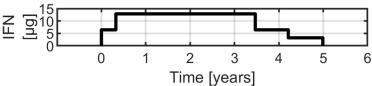
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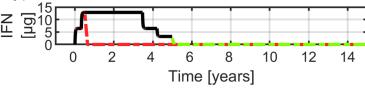
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## Hypothetical treatment:



# Halting treatment



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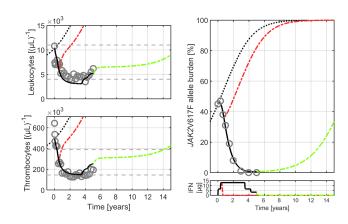
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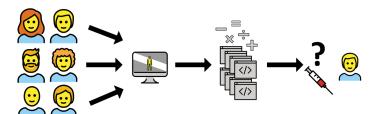
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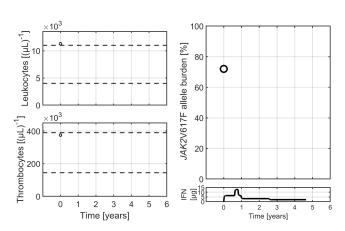
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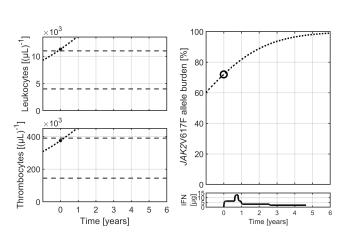
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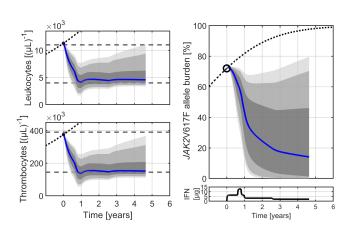
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Estimating stage of disease progression and prediction in absence of treatment.



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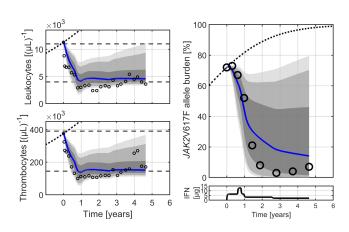
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Simulation 1000 virtual patients with same treatment-plan.

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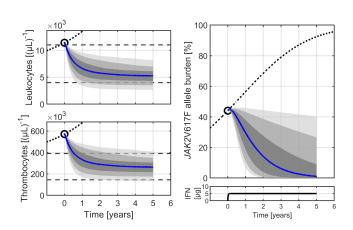
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Comparing to actual patient data

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Completely hypothetical patient: Baseline data from mean of cohort.

► Mathematical modelling can relate mechanistic

understanding of treatment to clinical measures.

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Mathematical modelling can relate mechanistic understanding of treatment to clinical measures.

► Individual patients' response to treatment can be captured by the model.

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Mathematical modelling can relate mechanistic understanding of treatment to clinical measures.

- Individual patients' response to treatment can be captured by the model.
- ► Combining the response of multiple patients could make forecasting possible.

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 Mathematical modelling can relate mechanistic understanding of treatment to clinical measures.

- Individual patients' response to treatment can be captured by the model.
- Combining the response of multiple patients could make forecasting possible.
- However, further validation is required before predictions can be used in the clinic.

### Comments on the future

be modelled succesfully.

► Highly complex diseases affecting the entire body can

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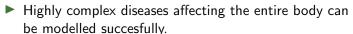
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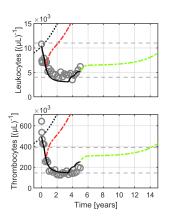
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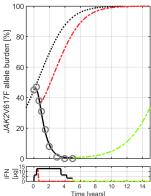
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► Mathematical models as expert-assistance tool.





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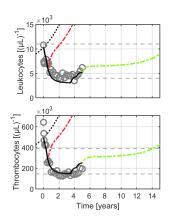
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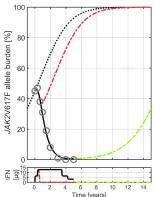
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► Highly complex diseases affecting the entire body can be modelled successfully.

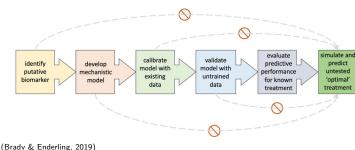
- Mathematical models as expert-assistance tool.
- ► And as a tool for patient-communication.





### Comments on the future

- ► Highly complex diseases affecting the entire body can be modelled succesfully.
- Mathematical models as expert-assistance tool.
- And as a tool for patient-communication.
- Promising for the future of oncology and heamatology, but careful and thoughtful validation is important.



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# Thank you for your attention.

## Any questions?



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### References and related articles

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