Model (PHREND[®]) for personalized prediction of treatment response in relapsing remitting multiple sclerosis (RRMS)

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Project goals

- In multiple sclerosis (MS), treatment decision is currently based on intuition from physicians.
- "Trial and error" takes up time, is cost intensive and accelerates the disease progression.
- **Patients have a desire to get a second opinion** on which treatment could work best for them based on the experience from similar patients.

'Marksman's shot'

Real-world evidence (RWE) data

- Since 2008 a Germany-wide network of physicians has maintained a database that currently documents more than 20'000 MS patients.
- The number of patients represents approx. 15% of the total market of MS patients in Germany (cf. Vfa 2014).

Patient ID	Age	Sex	Birth date	Date of diagnosis	EDSS value	Relapse (yes/no)	Factor n
1	56	Μ	26.11.1959	01.11.2004	2	yes	XXX
2	61	F	09.07.1954	26.06.2014	3	no	YYY
3	38	М	10.10.1977	06.04.2000	2.5	no	ZZZ
4	26	F	02.04.1969	13.05.2005	6	yes	QQQ
T, MI		U-U 15	Ų				10001010 01010101



- The database contains **demographic data**, such as patient's age and gender, as well as **clinical data**, such as patient's quality of life, diagnosis, treatments, side effects, rationale for a change of treatment, and several others over 1'000 variables in total.
- On average: >3 visits V_i per patient per year and 5 years observation period per patient.
- The database is active and with every half a year 500'000 entries are newly added.
- In total, more than 20'000 **therapy cycles**.

DISCLAIMER

< 3 | 4 BEHANDLUNGEN



Relapse EDSS = Expanded Disability Status Scale

'Shotgun blast'

Indicators of treatment response

- 1. Which treatment can reduce the **number of on-therapy relapses?**
- 2. Which treatment can reduce the **probability of an on-therapy confirmed disease progression (CDP)** based on the EDSS value?

Modelling approach

Business application

- Proposed solution: **Web-based App** providing personalized comparisons of treatment response using RWE data.
- **Input from physician:** currently 12 easy-to-enter variables summarising the patient's profile.
- **Options:** desired prediction period.
- Number crunching: predictive model used to provide two indicators of treatment response for all disease modifying MS-treatments (if sufficient RWE data is available).
- **Output:** probability of being relapse-/CDP-free + confidence interval for all available treatments.

• User friendliness:

model embedded in an enhanced tool to better guide the patient/ physician communication and decision.

PHREND						
ATIENTENDATEN						
PATIENT	MS DIAGNOSE					
GESCHLECHT	DATUM DER MS DIAGNOSE					
Männlich Weiblich	MM					
GEBURTSDATUM	AKTUELLE THERAPIE					
MM	Bitte auswählen	\sim				
	DAUER DER AKTUELLEN THERAPIE					
	Monate					
	THERAPIE VOR DER AKTUELLEN THERAPIE (OPTIONAL)					
	Bitte auswählen	\sim				
	ANZAHL THERAPIEN INSGESAMT					
	Keine krankheitsmodifizierende Therapie	\sim				
	AKTUELLER EDSS WERT	8				
	Bitte auswählen	\sim				
	WANN WAR DER LETZTE SCHUB?					
	Bitte auswählen	\sim				
HINWEIS Nur für Patienten mit der Diagnose BBMS Frühestens	ANZAHL SCHÜBE IN DEN LETZTEN 12 MONAT	EN				
sechs Monate ab Diagnosestellung. Derzeit nur für Patienten mit EDSS Wert bis 6.	Bitte auswählen	\sim				
W	EITER					
FRIKEIND		L⇒ ABMELDEN				

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ĀFERENZEN	\bigcirc			
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ANGERSCHAFT	2 Jahre	3 Jahre	4 Jahre	
ie Präferenz				
fikation bei Kinderwunsch	VORAUSS	GAGE FUR 100 AHNLICHE PA	TIENTEN	
likation bei				
wangerschaft				

- **Predictive models** based on the assumption that:
- 1. the number of on-therapy relapses follows a **negative binomial distribution;**
- 2. the CDP follows a **binomial distribution**.
- **Hierarchical generalized linear models (GLM)** are employed for both indicators of treatment response, with model parameters depending on patient's profile and treatment.
- The correlation between measurements coming from the same clinical site is incorporated through random effects.
- The duration of the therapy cycles is included as an offset term.
- Bayesian inference preferred due to the possibility of specifying (weakly informative) priors and preventing overfitting.



Model performance assessment

- **Calibration:** how well do predictions compare to observations?
- **Generalizability:** can the model be successfully applied to new data?
- 1. 10-fold cross-validation;
- 2. Leave-one-out cross-validation with respect to the clinical site;
- 3. Validation on test set.
- Comparison to **benchmark models** of decreasing complexity: is the knowledge of the patient's profile improving the predictions?
- Empirical test of the predicted treatment benefit: do patients profit from adhering to the recommendation?
- **Robustness:** is the model sensitive to different choices of the priors, to the characteristics of the patient population and to the sample size?
- Quality of predictions assessed using the following statistical measures:



Results and future prospects

Further developments

- Model refinement and extension (new variables and indicators of treatment response)
- Collection of new data
- Analysis of collected data on therapy decision

Status

- Currently in beta testing phase
- Roll out to German doctors' offices in 2018
- Results to be published in a scientific journal (in progress)
- Web-based App CE certified as medical device

Key success factors

- From doctors for doctors
- Use of RWE data
- Impact size
- Independence
- Living model
- Scalability to other diseases

Have a look at the PHREND[®] video:



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