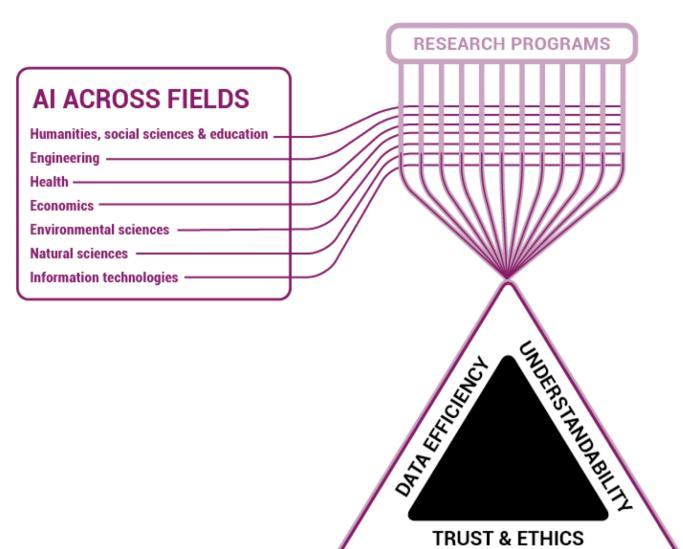


## Applications of AI in healthcare (Highlight B)

FCAI Highlight Program B creates AI tools to tackle real-world problems in healthcare together with expert collaborators from the respective fields



### Program objectives

#### Application B1: Al for genetics (Contact: Samuli Ripatti, UH)

We will create AI to analyze multivariate but structured genotype and phenotype data. The FinnGen project combines genetic data and electronic health records for 500,000 Finns. In collaboration with FinnGen we apply the AI tools to find genes modifying disease risk, progression, and comorbidities.

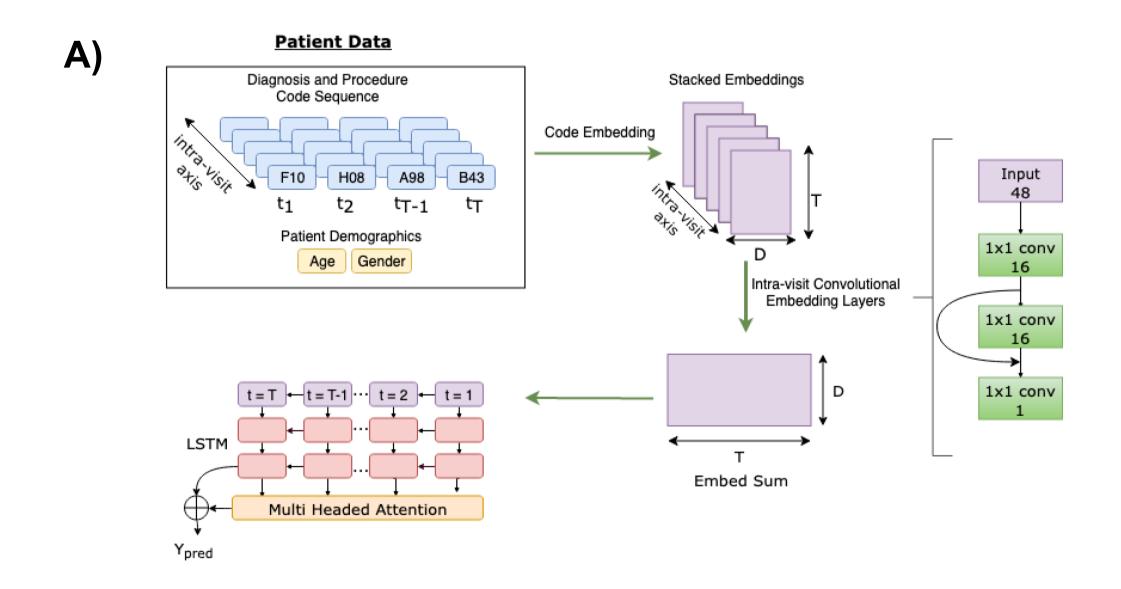
## **Application B2: Computational vaccines** (Contact: Jukka Corander, UH)

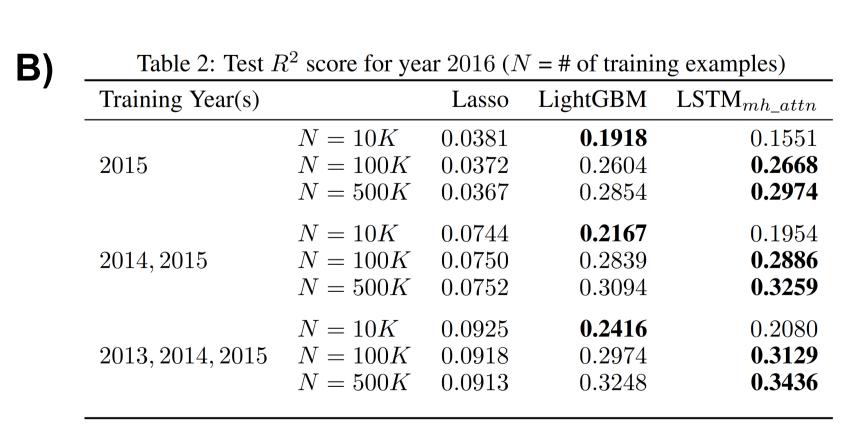
We will develop an Al-driven R&D tool for digital engineering of bacterial vaccines, which uses population genomic surveillance data combined with experiments to make probabilistic predictions of the campaign effects for candidate vaccines and to identify optimal formulations. The tool will significantly accelerate development of new vaccines and has large implications for global human health.

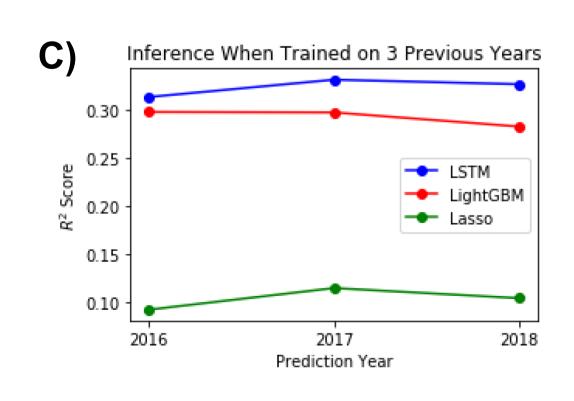
# **Application B3: Healthcare resource allocation** (Contact: Pekka Marttinen, Aalto)

We will create AI for prediction of healthcare services and train it on nation-wide healthcare register data, in collaboration with the National Institute for Health and Welfare (THL). The platform can predict healthcare costs of individuals and will be used to allocate resources to healthcare providers in a fair and efficient way. It will also be used to assess and compare treatment practices across the country to identify the most effective ones.

#### Selected results



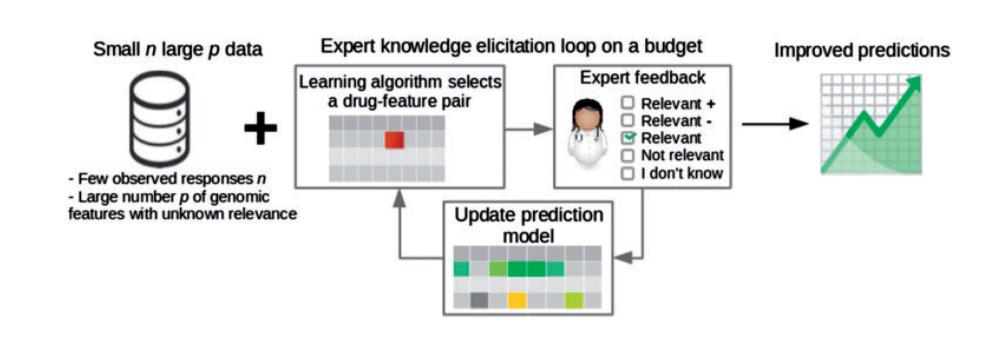




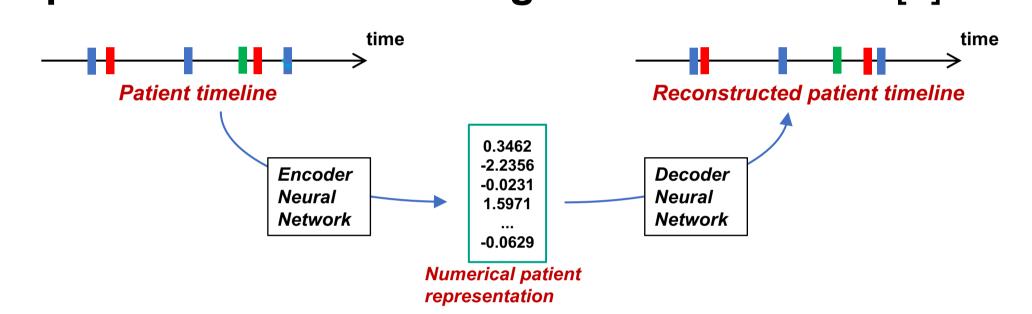
**A)** Outline of the model used in [3] for predicting the number of visits to a doctor next year, given individual treatment histories from previous years. The model was trained using pseudonymized out-patient data (AvoHILMO) on 1.4M elderly Finnish individual from years 2012-2018. **B)** Prediction accuracy of the model with different sizes of the training sets and lengths of treatment histories. **C)** These results demonstrate that the model generalizes well to future years (2017, 2018) from which no data were used when training the model.

## Example methodologies

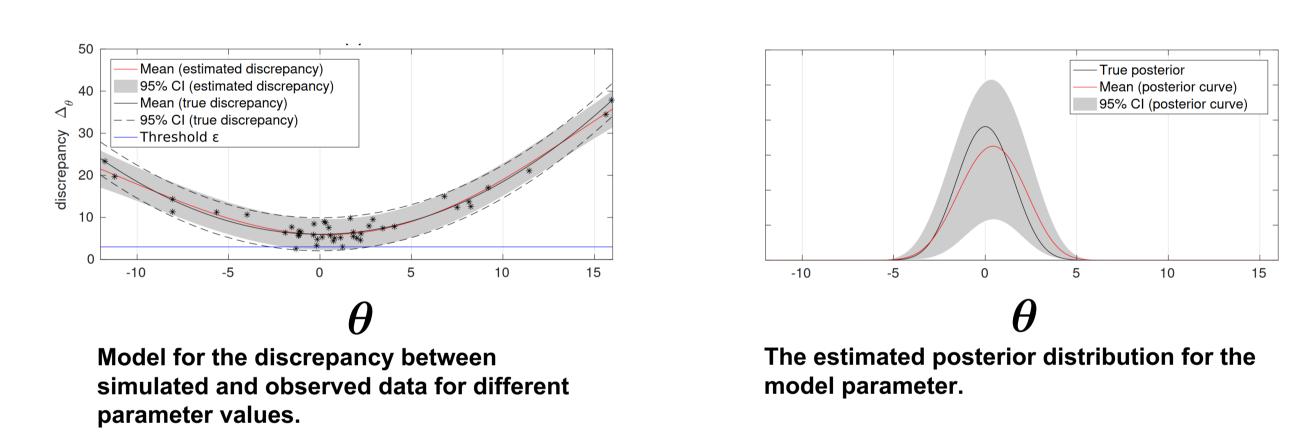
#### Interpretable and interactive machine learning [1,5]



#### Deep neural networks and generative models [3]



#### Model-based likelihood-free inference [2,4]



### Related publications

- [1] Cui, T., et al. (2019). Learning global pairwise interactions using Bayesian neural networks. arXiv:1901.08361.
- [2] Järvenpää, M., et al. (2019). Efficient acquisition rules for model-based approximate Bayesian computation. *Bayesian Analysis*.
- [3] Kumar, Y., et al. (2019). Predicting utilization of healthcare services from individual disease trajectories using RNNs with multi-headed attention. In: *PMLR Proceedings of the ML4H: Machine Learning for Health Workshop.* (to appear)
- [4] Lintusaari, J., et al. (2018). ELFI: Engine for likelihood-free inference. *The Journal of Machine Learning Research*.
- [5] Sundin, Peltola, et al. (2018). Improving genomics-based predictions for precision medicine through active elicitation of expert knowledge. *Bioinformatics*.

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