



Machine Learning for Personalized Medicine

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WORLDWEBFORUM 2017, Machine Learning

Part I. Motivation

What is machine learning?
What is personalized medicine?

Machine learning

A definition by Tom Mitchell The Discipline of Machine Learning. (2006)

... “[A] machine learns with respect to a particular **task** \mathcal{T} ,
performance **metric** \mathcal{P} ,
and type of **experience** \mathcal{E} ,
if the system reliably improves its performance \mathcal{P}
at task \mathcal{T} ,
following experience \mathcal{E} ...”

- Recommender systems, e.g. Netflix[®]
- Tailoring of ads and newsfeeds in social networks, e.g. Facebook[®]
- Web searches and ranking of pages, e.g. Google[®] search
- Spam filtering of e-mails, and many others

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Personalized Medicine

What is personalized medicine?



- Doctors recognize that every patient is unique → tailor treatment as best they can
- Important discovery: Matching a blood transfusion to a blood type
 - "...What if matching a cancer cure to our genetic code was just as easy, just as standard?"

Barack Obama. obamawhitehouse.archives.gov/precision-medicine

- Definition of personalized medicine: "...delivering the right treatments, at the right time, every time to the right person."
- Launch of the Precision Medicine Initiative (January, 2015)
 - +1,000,000 participants
 - biological samples, genetic data, lifestyle information linked to electronic health records

Science Magazine. doi:10.1126/science.aaa6436; "All of Us" - www.nih.gov/allofus-research-program

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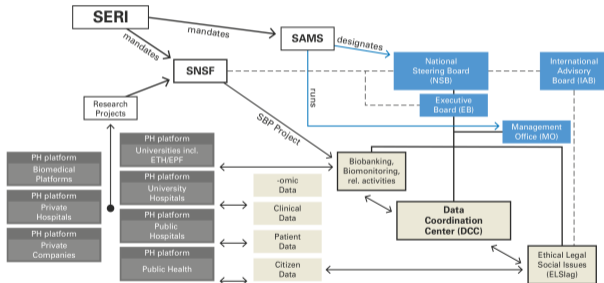
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Initiatives in Switzerland

Swiss Personalized Health Network (SPHN)



SAMS Bulletin 01.2016. www.samw.ch/en/Projects/Personalized-Health.html



Swiss Institute of
Bioinformatics

→ Establishment of standards for data production and storage

- Two phases:
 - 2017-2018: Funding of infrastructure
 - 2019-2020+: Funding of research projects
- Two clusters: Lausanne-Geneva and Zurich-Basel
- **Major goal:** Medical informatics systems in Swiss University Hospitals must be interoperable

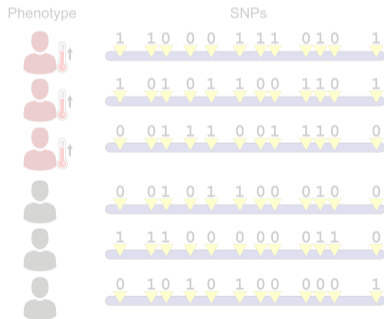
Part II. Applications

Significant pattern mining
Other ongoing projects

Significant pattern mining

Definition F. Llinares-López et al. KDD 2015

- The goal of *significant pattern mining* is to identify sets of items that occur statistically significantly more often in one class than in the other.

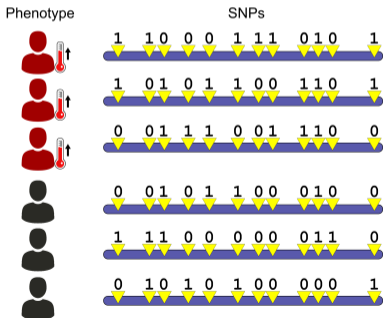


Llinares-López, et al. Bioinformatics (2015)

Significant pattern mining

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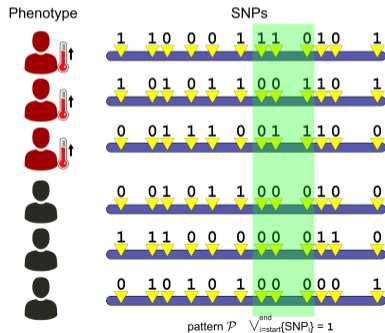
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Significant pattern mining

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What is not significant pattern mining

Frequent itemset mining



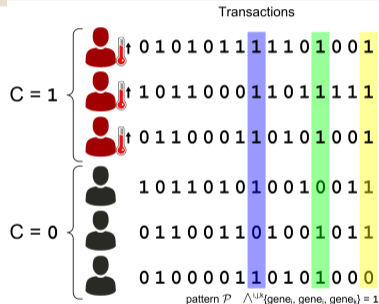
- **Goal:** Identify sets of products that are jointly bought by most customers
- Construct association rules of frequent itemsets

Significant pattern mining

Key aspects

	Pattern \mathcal{P} is present	Pattern \mathcal{P} is not present	
$C = 1$	a	$n_1 - a$	n_1
$C = 0$	$x - a$	$(n - n_1) - (x - a)$	$n - n_1$
	x	$n - x$	n

- Compute p -value based on a , x , n_1 and n
 - Use Fisher's Exact Test R.A. Fisher, 1922
- Must guarantee Family-wise Error Rate $< \alpha$
- Correct for multiple hypothesis testing
 - Exclude "untestable" hypotheses R.E. Tarone, 1990



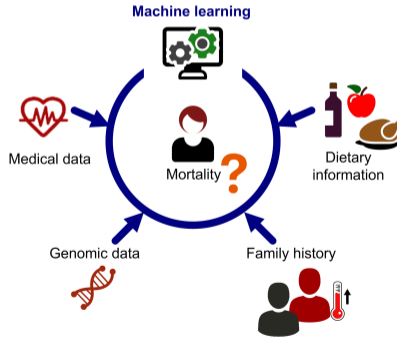
Papaxanthos et al. (2016) NIPS

Ongoing projects

Mortality prediction through machine learning (UK Biobank)

Goal

Predict the death risk of a patient in the near/mid future based on current health status data and genotypic data



- Integration of different sources of data (~ 150,000 patients)
 - Questionnaires: e.g. average weekly beer intake, happiness level, work/job satisfaction
 - Medical data: hospital admission/diagnoses
 - Genomic data: genotype calls
 - Death register: date of death, primary cause

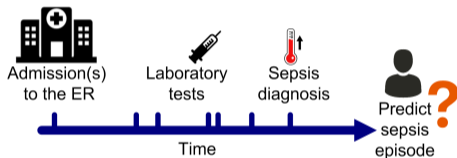
Ongoing projects

Septic shock prediction (Universitätsspital Basel)

Goal

Predict the occurrence of septic shock based on patients' laboratory tests and previous diagnoses

- No “magic” biomarker exists to predict early stages of the disease
- Mortality can reach up to 70% → early diagnosis has direct impact on life expectancy
- Data collected for patients since 2008.
- Total of ~ 1,800 laboratory tests conducted in successive hospital admissions
- Patients' outcomes are known



Part III. Challenges

Ever-increasing amount of data
Need for more complex tools to analyze the data
Insufficient data sharing

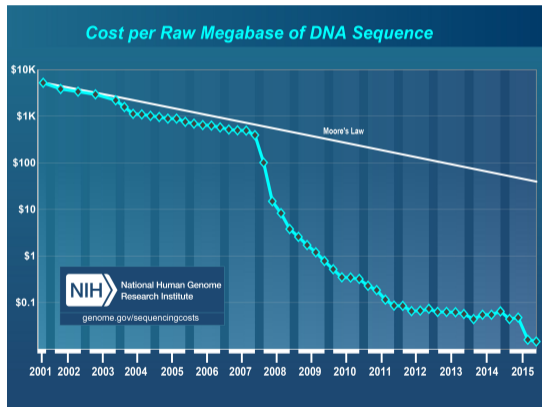
Decaying cost of sequencing/genotyping (in US\$)

Cost of sequencing one human genome

- 2003: \$500-1,000 million
 - Human Genome Project (estimated)
- 2006: \$20-25 million
- 2016: less than \$1,000

Cost of genotyping (array)

- ~\$100 per sample
 - HumanOmniExpress-24 BeadChips
 - 713,014 markers



K.A. Wetterstrand. DNA Sequencing Costs. Available at: www.genome.gov/sequencingcostsdata. Accessed January 2017.

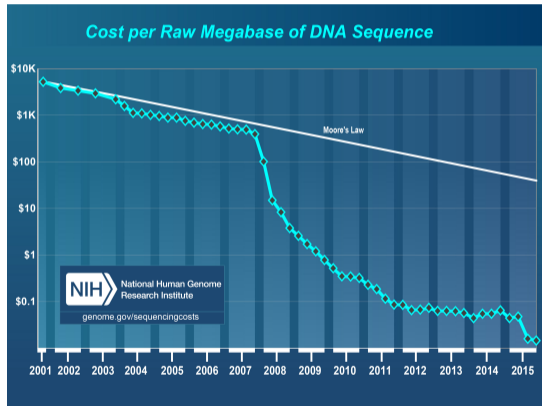
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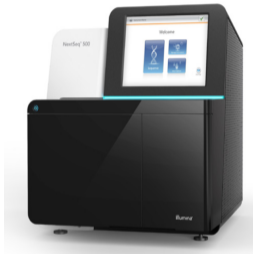


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www.genome.gov/sequencingcostsdata. Accessed January 2017.

Plethora of data

High throughput technologies (to name a few...)

- Measure expression levels of mRNA, rRNA, tRNA, and other non-coding RNA
- Identify genomic locations of DNA-binding proteins
- Measure abundance of metabolites
- Determine the 3-dimensional structure of chromosome in the nucleus



Illumina NextSeq 500, datasheet

Machine learning challenge

- Design better tools for data analysis
- Integration of patients' clinical information and *omics* data
- Design models for early disease diagnoses and predictors of response to treatment

Insufficient data sharing

Editorial in the New England Journal of Medicine

- “Research parasites”
 - Not involved in the study design
 - Potentially steal from research productivity envisioned by data gatherers, or
 - Disprove conclusions of the study

Longo, D.L., Drazen, J.M. Data Sharing. N Engl J Med 2016; 374(3): 276-7



Cartoon by Pécub. SIB.

Summary

Part I - Motivation

- Personalized medicine initiatives in the US and Switzerland

Part II - Application

- Significant pattern mining.
- Application I: Intervals of consecutive point mutations with pattern \rightarrow search space in $O(d^2)$
- Application II: Subsets of point mutations \rightarrow search space in $O(2^d)$
- Use **testability** to discard hypotheses and prune search space

Part III - Challenges

- Large amounts of *omics* data produced at unprecedented scale. Yet, insufficient data sharing

Thank You

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- Caroline Weis



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References

- Title slide: Photo by Peter Gartmann, <http://susanneminder.ch/susanne-minder-bildarchiv-basel-bilder/>
- Figures without acknowledgement were created by Damian Roqueiro using icons downloaded & altered from: flaticon.com (designed by Freepik)
- **Slide 3:** Tom Mitchell. 2006. The Discipline of Machine Learning. <http://www.cs.cmu.edu/~tom/pubs/MachineLearning.pdf>
- **Slide 4:** Picture extracted from <https://obamawhitehouse.archives.gov/administration/president-obama>
Text extracted from <http://obamawhitehouse.archives.gov/precision-medicine>
Definition of precision medicine: extracted from Science Magazine. doi:10.1126/science.aaa6436
Details about NIH Research Project “All of Us” <https://www.nih.gov/allofus-research-program>
- **Slide 5:** Swiss Academy of Medical Sciences, Bulletin 01.2016. <http://www.samw.ch/en/Projects/Personalized-Health.html>
- **Slide 7:** Definition of significant pattern mining: F. Llinares-López, M. Sugiyama, L. Papaxanthos and K. Borgwardt. (2015). Fast and Memory-Efficient Significant Pattern Mining via Permutation Testing. In Proceedings of the 21th ACM SIGKDD International Conference on Knowledge Discovery and Data Mining (KDD '15). ACM, New York, NY, USA, 725–734
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 - **Slide 9:** R.A. Fisher. (1922). On the interpretation of χ^2 from contingency tables, and the calculation of P. Journal of the Royal Statistical Society 85 (1): 87–94
 - R.E. Tarone.(1990). A modified Bonferroni method for discrete data. Biometrics 46, 515
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 - **Slide 13:** K.A. Wetterstrand. DNA Sequencing Costs. Available at: <https://www.genome.gov/sequencingcostsdata/>. Accessed January 2017.
 - **Slide 14:** Image of Illumina NextSeq System 500 extracted from datasheet: <https://www.illumina.com/systems/nextseq-sequencer.html>
 - **Slide 15:** D.L. Longo and J.M. Drazen. (2016). Data Sharing. N Engl J Med, 374:276–277
- Cartoon created by Pé cub. Property of the Swiss Institute of Bioinformatics.