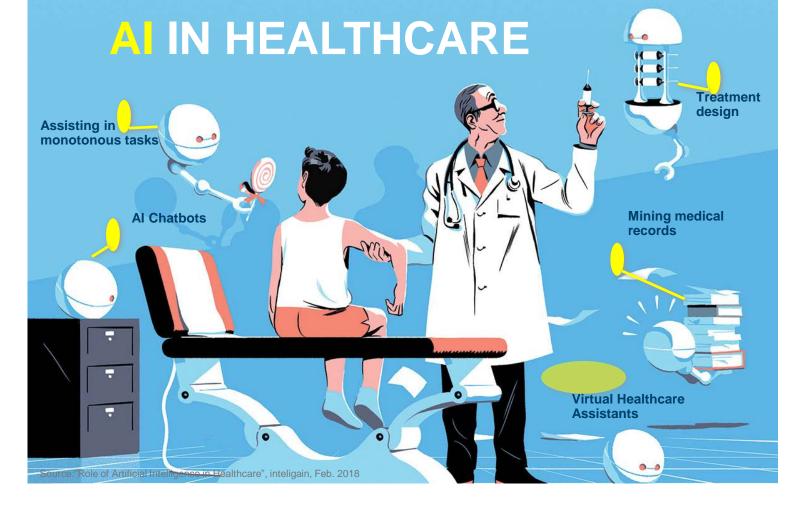
Al in Healthcare and Lifesciences

Maria Gabrani, Cognitive Healthcare and Lifesciences, IBM Research Zurich Swiss Medtech Day/ June 6th, 2019





Role of AI in Healthcare and Lifesciences: expected impact

\$6.6 Billion by 2021 the marker for Al in Healthcare (1)

\$150 Billion by 2026 annual savings for the US Healthcare economy (2)

50% reduced treatment costs (1)

30-40% improvement in outcomes (1)

Everything starts at the data



Al principles

- The purpose of AI is to augment human intelligence
- Data and insights belong to their creator
- New technology, including Al systems, must be transparent and explainable

Trust

Requires accurate and understandable evidence

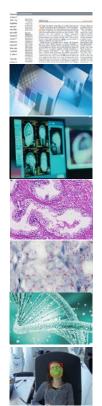
Compliance GDPR, FDA

Redress, Accountability, Liability Judicial system, healthcare

Safety

Need to predict behavior under changes in data

Al for Healthcare: Data driven, human centric, explainable, actionable



Data ingestion

- Data preprocessing
- Error correction
- Noise removal
- Data security and privacy
- Multiple languages
- Handwritten text
 - Scanned/faxed

Information extraction

- Ontologies
- Word embeddings
- Learn from small cohorts
- Annotation free
- Class unbalance
- Data uncertainty
- Continuous learning
- Diagram Understanding
- Interpretability
- Multimodal integration

Knowledge generation

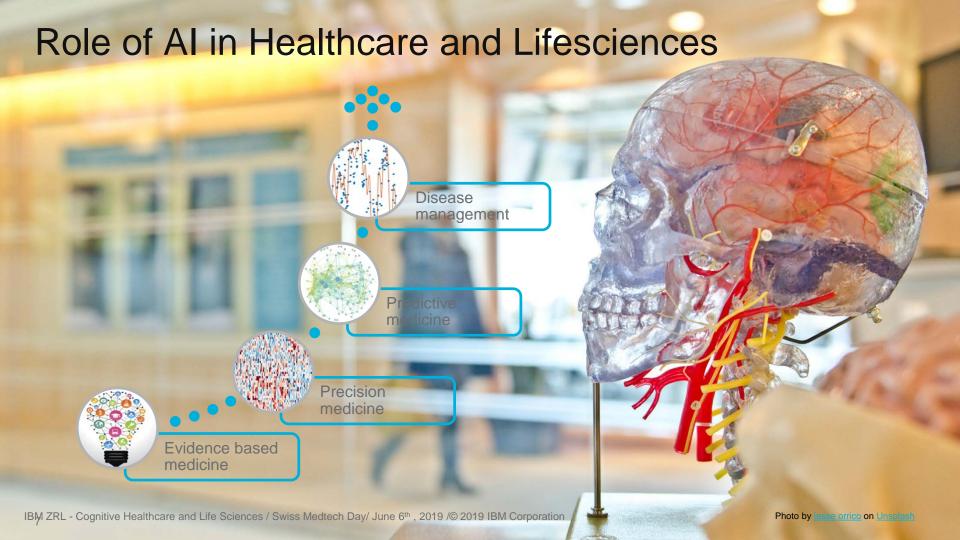
- Knowledge base
- *-*-interactions
 *={gene, protein, drug, ...}
- Semantic representation of knowledge
- Relation detection
- Explainable decisions
- Multimodal integration

Insights generation

- Reasoning
- Decision support
- Recommender systems
- Patient similarity
- Anomaly detection
- Behavioral analysis
- Personalized learning
- Monitoring
- Dialog ("Next best question")







Role of AI in Healthcare and Lifesciences Individual & sensor data Disease management Data integration Predictive medicine Molecular Data Big Data Precision medicine Evidence based medicine

Photo by jesse orrico on Unsplash



Improving Access Using AI

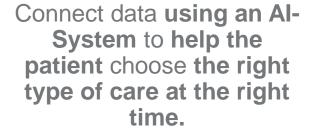








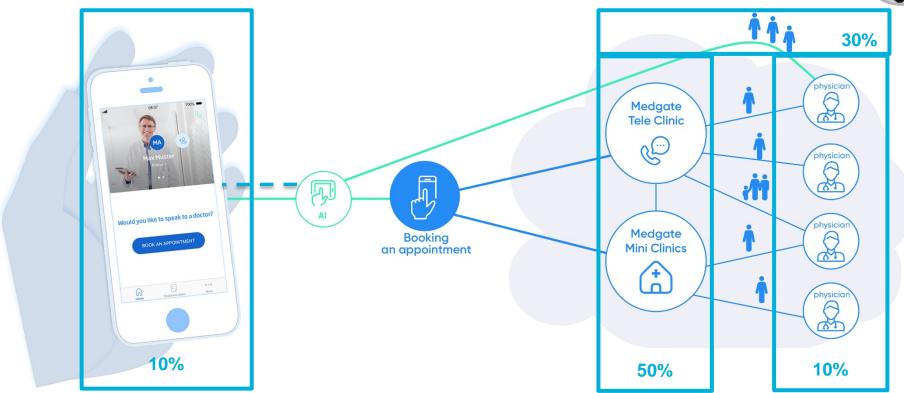
More than **7.4 million** teleconsultations **over 18 years** generated a huge amount of **data.**





Al for Decision Support System for Patients





Demo of the app







Al for Decision Support Systems – Explainable Al on clinical data



Visualization of attention factors from neural networks can be used to explain the rationale of a recommendation

smoker | blood | syndrome | headache | vomiting blood | vomit | heavy vomiting foreign body sensation | nothing visible | itching | hard object got into eye no peripheral neurological symptoms | diabetes typ 1 | angina pectoris | shortness of breath | chest pain on the right finger burn injury | movement possible | pressure on chest on the left | electric shock | pressure on chest impetigo | mouth rash | minimal state of consciousness | temperature 38.0-39.0 °C | vomiting



Each line represents the symptoms extracted for an urgent patient case file. The darker the colour, the higher the attention factor for a symptom.

Determine medical urgency from reported symptoms highlighting their importance

Girardi et al., Explainable Deep Learning on Clinical Data, LOUHI, 2018@EMNLP



Drug discovery



- Pharmaceutical industry has the largest R&D business spending of any industry (19% of the total R&D spending worldwide) [2]
- Anticancer compounds in particular, take the lion's share of drug discovery R&D efforts
- Anticancer compounds account for 34.1 % of all drugs in global R&D pipeline in 2018 (5212 of 15267 drugs) [2]
- Despite all R&D spending and efforts serendipity and chance still play a big role in new drug discovery

The low success rate has been attributed to lack of a systematic way to accumulate and leverage years of R&D to achieve higher success rates [1]

[1] E. Petrova, Innovation and marketing in the pharmaceutical industry, Springer, 2014, pp. 19–81.

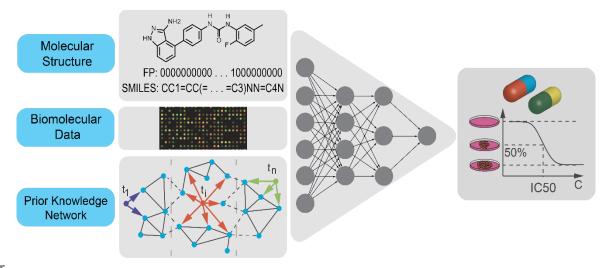
^[2] I. Lloyd, A. Shimmings, and P. S. Scrip, "Pharma R&D Annual Review 2017," Available Pharmaintelligence Inf. Comresourcesproduct-Contentpharma-Rd-Annu.-Rev.-2018 Accessed June 25 2018, 2017.

PaccMann – drug sensitivity prediction via deep learning



Multi-modal prediction of IC50 drug sensitivity.

Three key data modalities that influence anticancer drug sensitivity: biomolecular measurements of cancer cells (gene expression, copy number alteration etc.), a network of known interactions between the biomolecular entities and the chemical structure of the anticancer compounds.



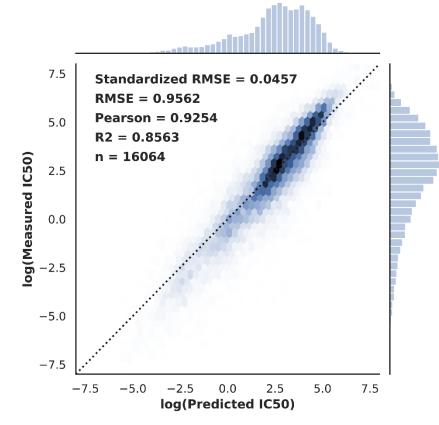
PaccMann Architecture – Multiscale convolutional attentive (MCA) encoder

MCA Encoder

- Three convolutional channels with variable kernel size and a residual channel
- MCA Enables capturing various molecular sub-structures
- Interpretable Via contextual attention mechanism on the SMILES and gene attention on the cell

PaccMann performs well in predicting IC50 drug sensitivity for unseen cell-drug pairs



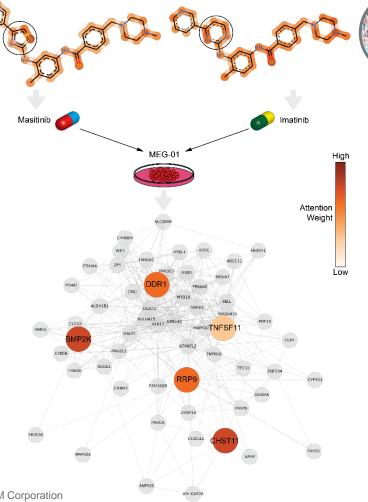




PaccMann Interpretability

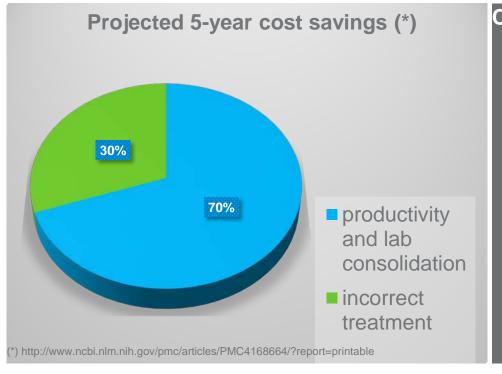
- The model is explainable and interpretable through gene and atom attention weights
- PaccMann gives high weight to important functional groups and relevant biomarkers in a leukemia cancer cell line

ibm.biz/paccmann-aas



Computational Pathology





Cost Savings:

"\$17.8 million the projected 5-year total cost savings

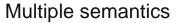
"\$3 million per year savings from FEDEX costs

Digital Cartography





- Mapping of information in multiresolution spatial data for geography representation
- Visualization of context at multiple levels (hemisphere, landscape, street level)
- Context relevant interaction information (e.g., location of closest restaurants or distance calculations)
- Enables personalized decision (e.g., area close to school vs. ability to bike to work)

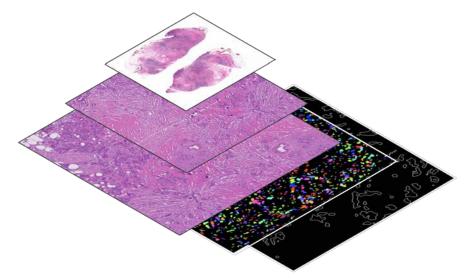




HistoCartography



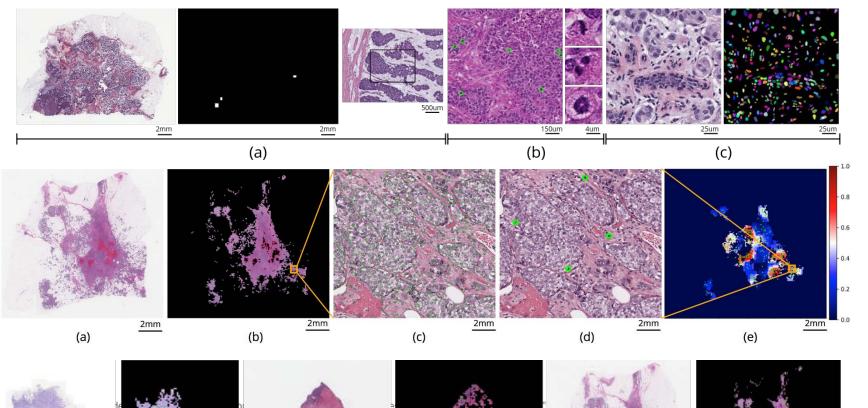
- Mapping of information in multiresolution phenotypic and molecular expression data for pathogenesis representation
- Visualization of context at multiple levels (organ, cellular formations, molecular level)
- Context relevant relational information (e.g., location of glands, or cell formations to region boundaries)
- Enables interpretable personalized decision at whole slide image and local scale (e.g., studying heterogeneity, grading or scoring)



Multiple semantics

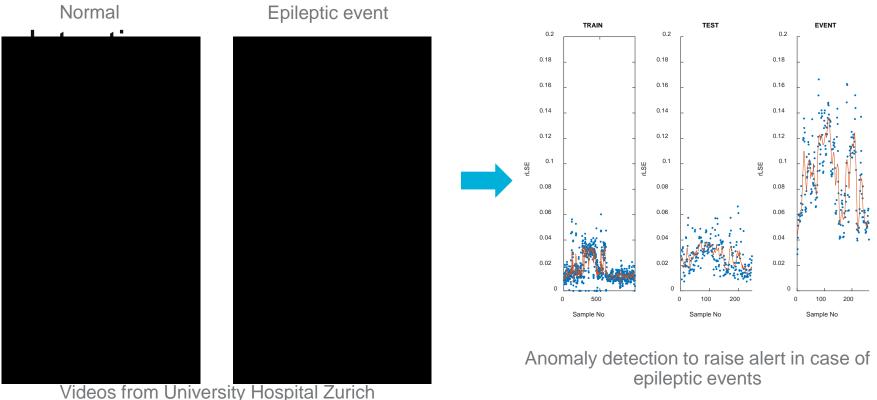


Context Aware Mitotic Activity

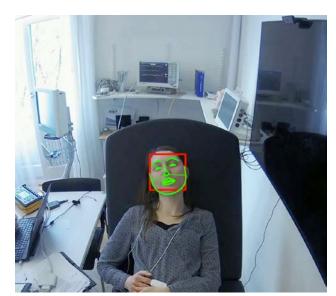


Al for Video-based patient monitoring – Epilepsy

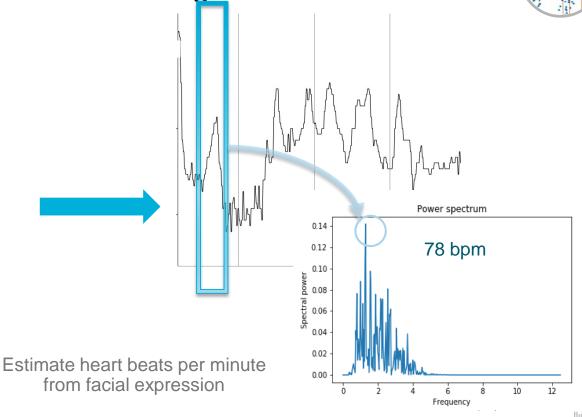




Al for Video-based patient monitoring — Hearth rate estimat



Video of team member



Summary: Al in Healthcare and Lifesciences









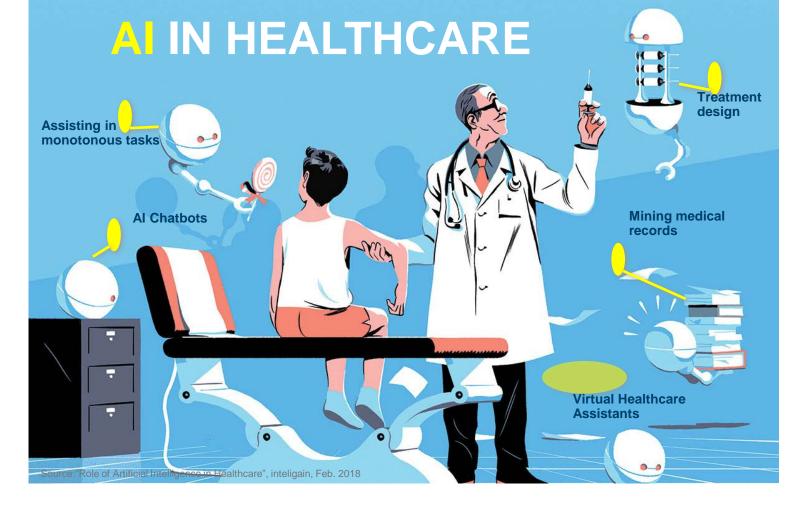
Learn from multimodal data

Build domain knowledge that enhances human understanding Provide interpretable insights

Provide actionable insights

Summary: Al in Healthcare and Lifesciences



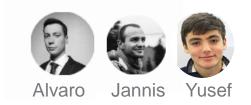


The team

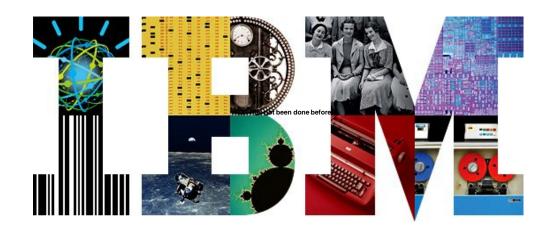


Andreas Antonio Chiara Doug Ivan Maria Marianna Matteo Matthias





Thank You!



IBM Research | Zurich

