

Precision medicine unit white paper v1.0

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1 Executive summary

The **Precision medicine unit** at Universitäts-Kinderspital Zürich is proposed to represent the forefront of medical science, using an integrated multi-omics approach to augment paediatric healthcare. This strategy extends beyond genomic data to incorporate a wide array of omics technologies, such as genomics, proteomics, metabolomics, and innovative methods like single-cell sequencing. Together, these technologies can provide a comprehensive understanding of disease mechanisms, which is instrumental in devising personalised treatment strategies for rare and complex paediatric conditions.

Comparable benchmarks have demonstrated clear cost-saving potential through precise diagnostics and targeted therapy. For rare diseases, approximately 40% of probands received a genetic diagnosis [13; 14]. Altered critical care management in 77% of diagnosed cases [9]. A well-managed work-flow can result in rapid whole-genome sequencing with a turnaround of 37 hours on average [12].

A forecast, projected into 2030, for the yearly cases of sepsis are shown in **figure 1**. Black and red values show the expected number of deaths with and without precise diagnostics and targeted therapy, respectively.

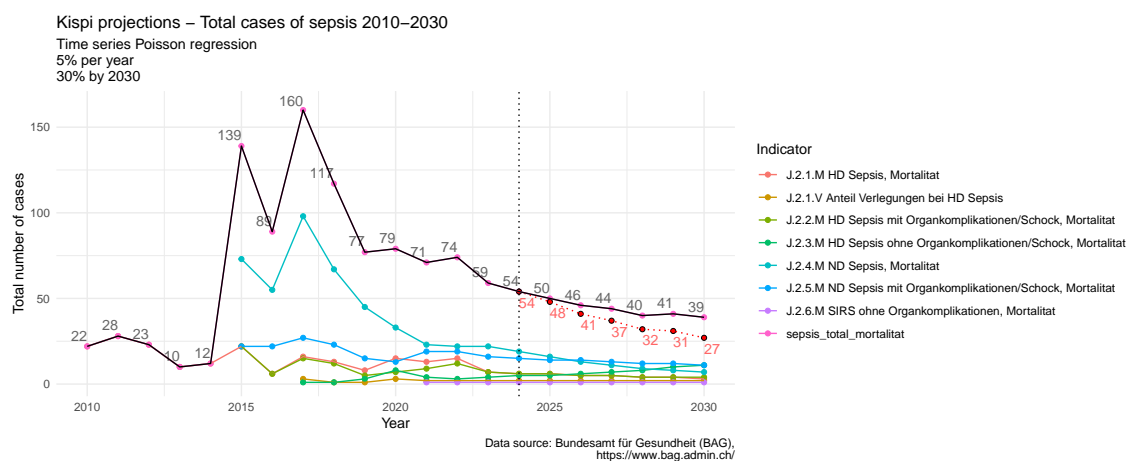


Figure 1: Year deaths due to sepsis at Universitäts-Kinderspital Zürich. This data is based on statistics reported by Bundesamt für Gesundheit (BAG), <https://www.bag.admin.ch/> for years 2010–2022. Time series was performed using Poisson regression to extrapolate the expected outcomes from 2010–2030. Predictions for the cost and number of cases were generated in section 14. DP: Diagnostic procedure. HD: Primary diagnosis. ND: Secondary diagnosis.

2 Introduction

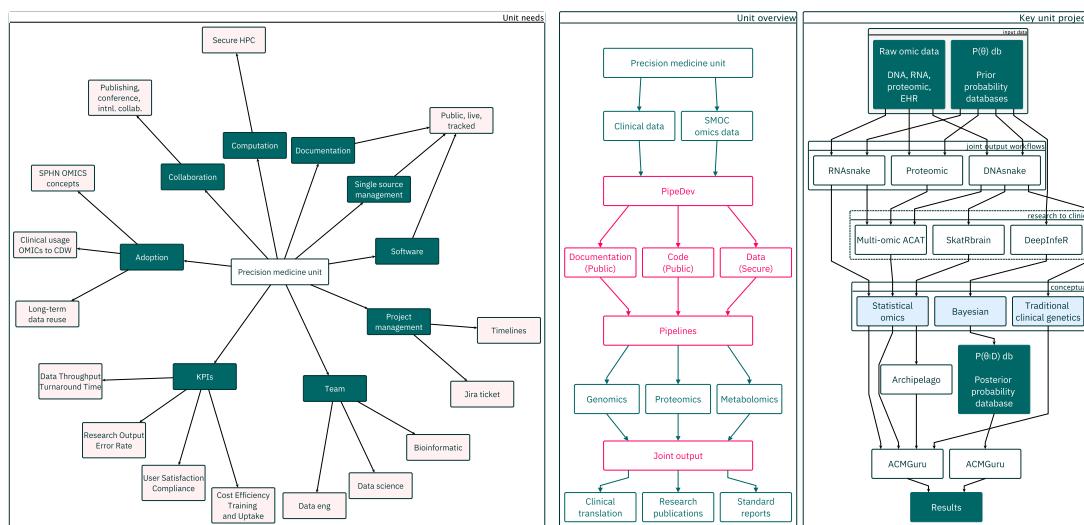


Figure 2: **Precision medicine unit** overview. The unit needs illustrate the management philosophy for minimal disruption for technical progress. The unit overview illustrates the main flow of information to key products. Key unit products illustrate the flow of information, with technical hurdles avoided using the principles of single-source management, and shared open development.

The **Precision medicine unit** is ready to update our paediatric healthcare by serving an array of advanced capabilities in multi-omics, pioneering research methodologies, purpose-built infrastructure, and rigorous data analysis techniques. Our approach sets a new benchmark for personalised treatment, employing cutting-edge omics services to provide tailored healthcare for the clinic and enrich basic science at Kinderspital Zürich. An overview of our unit is shown in **figure 2**.

Traditionally, medical practices have predominantly adhered to a standard approach, tailoring disease prevention and treatment strategies based on the average response anticipated from a general population. This conventional method has proven effective for certain patients and conditions, but often falls short when individual variances in genetics, environment, and lifestyle are significant factors influencing health outcomes. Precision medicine emerges as a revolutionary paradigm, diverging from the one-size-fits-all approach to embrace these individual differences. This innovative medical strategy integrates detailed layers of genomic, environmental, and lifestyle data to tailor personalized healthcare solutions. The evolution of precision medicine has been markedly propelled by significant advances in biomedical research, affecting millions worldwide. It aims not only to refine disease treatment and prevention but also promises enhanced outcomes by aligning more closely with each person's unique biological makeup. This approach has profound implications across various domains of healthcare, including oncology, pharmacogenomics, and the management of rare diseases, setting a new standard in personalized healthcare. Our partners are implementing precision medicine approaches at the [Swiss Institute of Bioinformatics \(SIB\)](#), [The LOOP Zurich](#), [CHUV](#), and all private industry leaders such as [Roche](#), [Genentech](#), and [Pfizer](#).

The **Precision medicine unit** at Kinderspital Zürich can be ready to collaborate closely with the Health 2030 Genome Center, part of the Swiss Multi-Omics Center (SMOC) (<http://smoc.ethz.ch>), to harness state-of-the-art genomic services. This partnership underpins our cutting-edge approach in pediatric healthcare, enabling us to deliver personalized treatment strategies through comprehensive multi-omic

data analysis. As a collaborative effort among major Swiss universities and university hospitals, including UNIBE, Inselspital, UNIL, CHUV, UNIGE, HUG, and EPFL, the center exemplifies a comprehensive approach to personalized health care and genomic research. SMOC integrates expertise across genomics, transcriptomics, proteomics, and metabolomics. This integrated multi-omic analysis is essential for the thorough exploration and understanding of clinical specimens. Accredited under ISO 15189, the center ensures the highest standards in sequencing and data analysis, leveraging state-of-the-art technologies such as the Illumina Novaseq6000 platforms and TruSeq DNA PCR-Free library preparation.

We will employ essential services such as:

- Clinical-grade whole genome sequencing (WGS)
- Clinical-grade whole exome sequencing (WES)
- Clinical-grade RNA sequencing (RNA-seq)
- Fast turnaround time for clinical-grade WGS, WES, and RNA-seq
- Low-coverage WGS with variant imputation
- Viral pathogen WGS
- Microbiome WGS

Our subsequent analysis occurs at the cutting edge of technology, employing the most reliable reference genomes based on GRCh38 and adhering to stringent guidelines for handling sensitive clinical data, as outlined by the SPHN/BioMedIT network on the sciCORE platform (<https://sphn.ch/network/projects/biomedit/>). This framework supports the SwissPedHealth initiatives, enhancing the capacity for robust clinical diagnosis and research. We are also thus preparing to meet the rapidly approaching national expectations. These expectations are demonstrated by projects like the Swiss Federated Genomics Network (SFGN) and Genome of Switzerland (GoS) which aim to sequence clinical-grade genomes on a national scale, contributing to European genomic initiatives and supporting the advancement of genomic medicine across the region.

By capitalising on these integrated expert services we can provide analysis within the **Precision medicine unit**:

- Sequencing data processing
- Genome variant calling
- Variant clinical annotation and prioritising
- RNA expression profiling
- Single cell RNA-seq
- Multi-omic joint analysis of DNA, RNA, proteomics

The **Precision medicine unit** will therefore be able to provide clinical-grade results with fast turnaround time - from DNA to variant interpretation and functional effect. Material will be shipped from Kispi to SMOC where high-throughput OMICs performed. OMIC analysis will then be performed on our secure high-performance computer infrastructure. Clinical reports containing actionable results can be provided to the clinic. Rich datasets will be generated for research and discovery.

3 Key products

Products of the **Precision medicine unit** include **ACMGuru (v1.0)** for variant classification and interpretation, **DeepInfer (v1.0)** for defining the posterior probability of every genetic determinant of disease based on all known public data, **Archipelago (v1.0)** for statistical interpretation of VSAT, **SkatRbrain (v0.2)** for automated statistical genomics, **multi-omic ACAT (v0.1)** for multi-omic joint analysis of VSAT, **DNAsnake (v0.1)** for DNA pre-processing, **RNAsnake (v0.1)** for RNA pre-processing, and the documentation of pipelines in **Pipe-Dev docs DNAsnake (v0.1)**.

3.1 Product example: advanced DNA sequencing data preprocessing

DNAsnake (v0.1) represents a vital product from our Precision Medicine Unit, meticulously designed to preprocess WGS DNA data for use in clinical genetics reporting, statistical analysis, and machine learning applications. Employing the Genome Analysis Toolkit (GATK), **DNAsnake (v0.1)** standardises the preparation of DNA sequencing data to ensure consistency and reliability across diverse analytical applications.

DNAsnake (v0.1) is engineered around the GATK best practices for DNA sequence data preprocessing. This workflow is integral to producing high-quality, clinical-grade DNA data outputs, which are crucial for downstream processes like variant interpretation in ACMGuru and assessing genetic determinants of disease with deepInfer.

The **DNAsnake (v0.1)** workflow is detailed and robust, encompassing several critical stages of DNA preprocessing:

1. **Quality Control and Pre-processing:** Initial receipt of raw FASTQ files followed by quality control assessments using tools like FastQC and subsequent trimming with Trimmomatic.
2. **Alignment:** Alignment of sequences to the GRCh38 reference genome using the Burrows-Wheeler Aligner (BWA).
3. **Post-alignment Optimization:** Includes marking duplicates with Picard Tools, and base quality score recalibration (BQSR) with GATK's BaseRecalibrator and ApplyBQSR tools.
4. **Variant Calling:** Utilizing GATK's HaplotypeCaller for calling germline SNPs and indels, followed by variant quality score recalibration (VQSR) to ensure high-quality variant calls.
5. **Output Generation:** Production of annotated, processed BAM and VCF files ready for comprehensive genetic analysis.

Processed outputs from DNAsnake feed directly into:

- **ACMGuru (v1.0)**, for detailed variant classification and interpretation in line with ACMG guidelines.
- **DeepInfer (v1.0)**, which utilizes the processed data to calculate the posterior probabilities of genetic variants, influencing disease phenotypes based on extensive public data repositories.

By automating the WGS DNA data preprocessing with **DNAsnake (v0.1)**, our unit will achieve:

- **Standardisation and reproducibility:** Ensures that all samples are processed through a uniform pipeline, reducing variability and enhancing the reliability of results.
- **Efficiency and scalability:** Capable of handling large-scale datasets with the flexibility to accommodate increasing data volumes without sacrificing performance.
- **Integration and Interoperability:** Seamlessly interfaces with other analytical tools and databases, promoting a cohesive and integrated approach to precision medicine.

DNAsnake (v0.1) exemplifies our commitment to delivering state-of-the-art solutions for genetic data preprocessing. As an element of our single-source management strategy, it not only supports but enhances the capabilities of our precision medicine initiatives, ensuring that data used across various platforms is of the highest quality and utility.

4 Multi-omics integration

Our unit will use whole genome sequencing (WGS) to identify both common and rare genetic variants, thereby establishing a foundational genetic blueprint of disease predisposition. In complement to WGS, RNA sequencing (RNAseq) provides insights into gene expression changes across different physiological states or in response to treatments, shedding light on the functional consequences of genetic alterations.

Furthermore, the incorporation of proteomics and metabolomics technologies offers a detailed examination of the proteome and metabolome, which are in closer correspondence to the phenotype. These methodologies facilitate the monitoring of biochemical activities and protein functions that have direct impacts on disease phenotypes, effectively bridging genotypic information with tangible biological and clinical outcomes.

In addition, routine clinical measurements and electronic health records (EHRs) will be integrated with omics data, augmenting the granularity and practicality of our findings. This synthesis enables tailored, real-time therapeutic interventions and continuous monitoring of disease progression, markedly enhancing the precision of diagnostics and personalised treatment plans.

5 Project management

We will manage intricate projects that encompass multiple stakeholders and complex data streams. Employing Jira, an advanced project management tool, we can clearly oversee tasks, timelines, and dependencies, ensuring that every phase of our projects conforms to our exacting standards and strategic objectives. This dynamic approach underpins our adaptability, facilitating the efficient execution of our research and clinical activities.

The documentation and outputs of our research will be rigorously curated from the first step to ensure transparency and accessibility, adhering strictly to our existing and future standards. Sensitive data will be secured and protected in BioMedIT to adhere with patient confidentiality and data integrity. Data transfers between BioMedIT and Kispi will follow standard guidelines and use secure methods such as sFTP backed by data transfer agreements. Stringent data stewardship is critical for maintaining trust and ethical standards in our research.

6 Data flow

The data generation and analysis process involves several stages, beginning with sample collection and culminating in the presentation of final results. This process is designed to integrate with existing clinical databases using newly developed OMICs concepts based on the SPHN ontology. The process is summarised in [figure 3](#).

We have worked with SPHN and [TheHyve](#) to develop new concepts which cover the generation of sequencing data and analysis, recently published in van der Horst et al. [[1](#)] as illustrated in [figure 4](#). We have also developed the concepts which cover the final outputs of downstream analysis results for omic results. This work enhances the integration of omics data into the SPHN Semantic Interoperability Framework, which primarily handles clinical routine data.

Our new genomics extension enriches the framework to include comprehensive descriptions of genomics experiments, encompassing both clinical and research applications. It outlines the entire omics process flow, detailing steps from sample processing to data analysis, including specifics like library preparation and sequencing analysis. The extension also integrates additional omics metadata, such as details on sequencing instruments and quality control metrics. By aligning with established semantic models and leveraging common biomedical vocabularies (e.g., EDAM, OBI, and FAIR genomes), it promotes semantic interoperability and aims to FAIRify data for shared use within the Swiss network, enhancing data reuse in a unified knowledge graph.

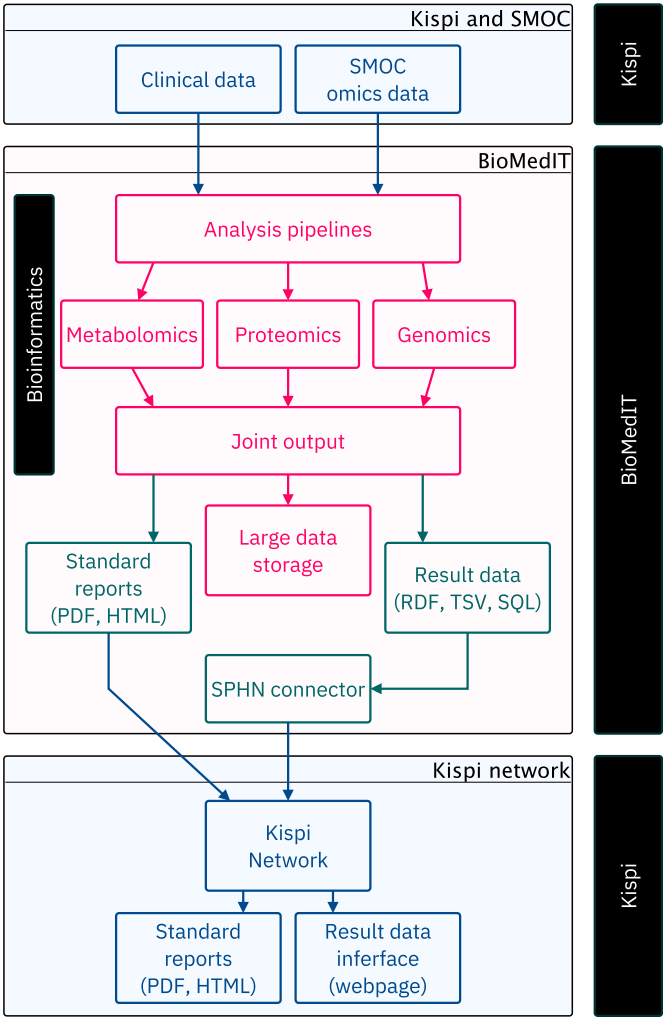


Figure 3: **Precision medicine unit** Data flow from sample collection to final result presentation.. Sample collection occurs in Kispi. SMOC typically processes most physical samples of DNA, RNA, serum, or other tissues. Multi-omic data is generated and transferred to BioMedIT. Bioinformatic analysis pipelines process the data and produce a main analysis output which is stored long-term. Key actionable results from this large dataset are prepared according to SPHN ontology, using concepts such as “sequencing assay” and “omic result”.

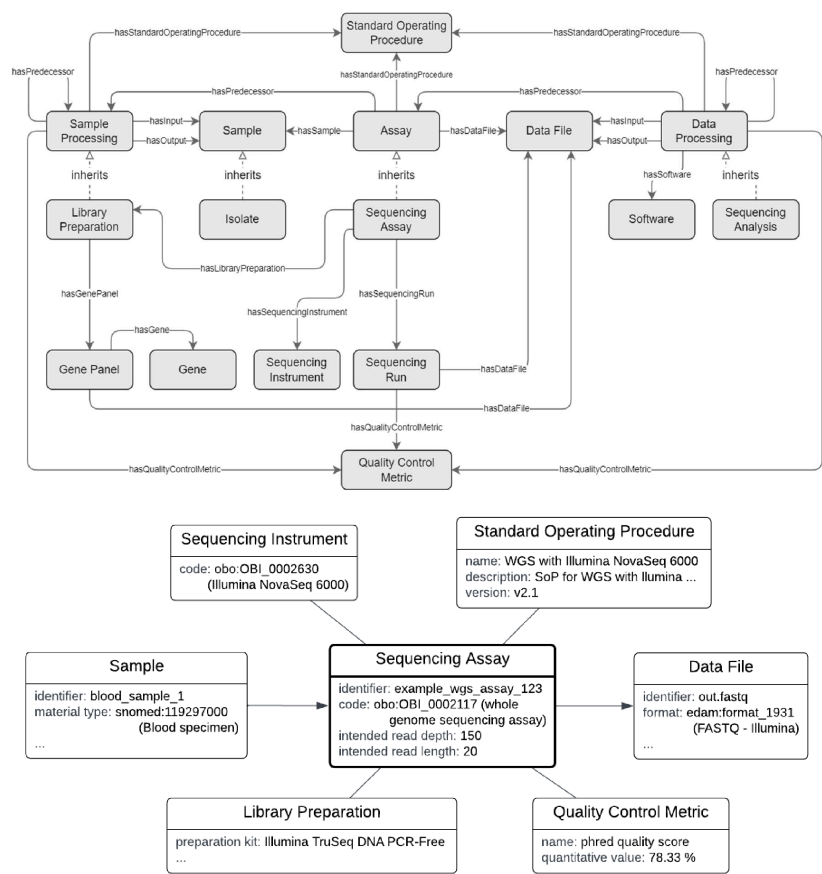


Figure 4: Figure extract from van der Horst et al. [1]. (A) Basic excerpt of the schema for the (gen)omics process flow. (B) Diagram visualizing an instance of a sequencing assay that analyzes one sample and produces one FASTQ file. ‘Sequencing Assay’ concept, together with its ‘Instrument’, ‘Library Preparation’, ‘Standard Operating Procedure’, and ‘Quality Control Metric’ concepts from which it is composed.

6.1 Sample collection and initial processing

- **Location:** Kispi Hospital.
- **Details:** Sample collection encompasses various biological materials including DNA, RNA, serum, and other tissue types. These are initially processed by the Swiss Multi-Omics Center (SMOC), which is responsible for the physical handling and preliminary omics data generation.

6.2 Data transfer and bioinformatics analysis

- **Transfer:** The raw multi-omic data generated by SMOC is transferred to BioMedIT using secure protocols such as sFTP.
- **Bioinformatics processing:** At BioMedIT, advanced bioinformatics pipelines are employed to analyse the data. This includes comprehensive analyses across metabolomics, proteomics, and genomics disciplines.
- **Outputs:** The main outputs from these analyses include:
 - A large dataset stored for long-term reuse and research purposes.
 - Standard reports generated in formats such as PDF and HTML.
 - Result data formatted in RDF, SQL, TSV, which are then adapted to meet the SPHN connector's requirements for merging into clinical data warehouses.

6.3 Data conversion and integration

- **Conversion:** Key actionable results are extracted from the large dataset and prepared according to our reporting evidence guidelines and formatted using the SPHN ontology. This preparation uses specific OMICS concepts such as “omic result” to ensure that the data can be seamlessly integrated and interpreted within the clinical framework.
- **Integration:** The processed results are converted to fit the database requirements of the hospital's clinical data warehouse.

6.4 Presentation of final results

- **Internal Network Transfer:** Outputs, including the standard reports and result data, are transferred back to the Kispi network. This step is crucial as clinicians do not have direct access to secured BioMedIT servers.
- **Access and Presentation:** Final analysis results are made accessible to clinicians through an internal webpage and downloadable PDF formats. This ensures that the results are readily available for clinical decision-making and further research.

By aligning with the SPHN RDF ontology and implementing it through newly developed OMICS concepts, this data flow ensures that genomic and other omic data types are integrated into the hospital's clinical operations, enhancing the capacity for precision medicine and personalised patient care. The entire process is illustrated in **figure 3**, providing a visual representation of the data flow from sample collection to final result presentation within the hospital's infrastructure.

7 Clinical implementation and future prospects

Our multi-omics data inform precision medicine protocols, guiding treatment plans that are specifically tailored to individual genetic profiles, thus enhancing treatment efficacy and minimising adverse effects. The prognostic models developed from our multi-omics data facilitate early interventions and enhance disease management strategies.

Looking to the future, we are dedicated to integrating single-cell sequencing and additional omic technologies, such as lipidomics and glycomics. These advancements are expected to considerably refine our understanding of diseases at a resolution that was previously unattainable. This development is anticipated to improve our diagnostic capabilities and therapeutic interventions, ensuring that Kinderspital Zurich continues to lead in paediatric healthcare innovation.

8 Strategic collaborations

Collaboration forms a cornerstone of our operational philosophy. We have partnerships both nationally and internationally, enhancing the capabilities and influence in paediatric precision medicine. These collaborations will extend beyond academic institutions; we will actively engage with healthcare providers, policymakers, and the pharmaceutical industry to ensure that our research has a wide-reaching and significant impact.

By participating in global research consortia and contributing to international databases, we share our findings and innovations, fostering a culture of open science and continuous improvement. These strategic partnerships amplify our research capabilities and enable us to remain at the forefront of technological and methodological advancements in precision medicine.

This proactive approach to project management and strategic collaboration positions by the **Precision medicine unit** will help use to become leaders in translating scientific research into clinical outcomes, significantly advancing the field of paediatric healthcare.

9 Funding model and sustainability

The **Precision medicine unit** will benefit from a diverse funding model that includes institutional support, competitive research grants, and strategic partnerships with industry leaders. This multifaceted approach provides a robust financial foundation that helps ongoing research and development in paediatric precision medicine. Institutional support from Kispi and its affiliated academic entities should furnish a stable base of funding that sustains core facilities and staffing requirements.

Beyond traditional funding avenues, we may proactively establishes collaborations with pharmaceutical companies and biotechnology firms. These partnerships frequently catalyse joint research projects and grant access to avant-garde technologies and additional streams of funding. By harmonising our research objectives with the needs of our industry partners, we not only secure essential resources but also ensure that our research outputs are pragmatically applicable and rapidly translatable into clinical settings.

Furthermore, we are dedicated to integrating sustainable practices to sustain all facets of our research activities, from data management and analysis to resource utilisation. Our goal is to build a system that is redundant to decay. To do so, we must produce tools that are sensible for newcomers to learn, provide a consistent method of production and usage so that they can be updated and maintained without dependency on single members.

The most common research funding pathways for academic members of the Precision medicine unit are shown in figure 5 and links to funding opportunities that are also supported by UZH assistance are provided in table 1.

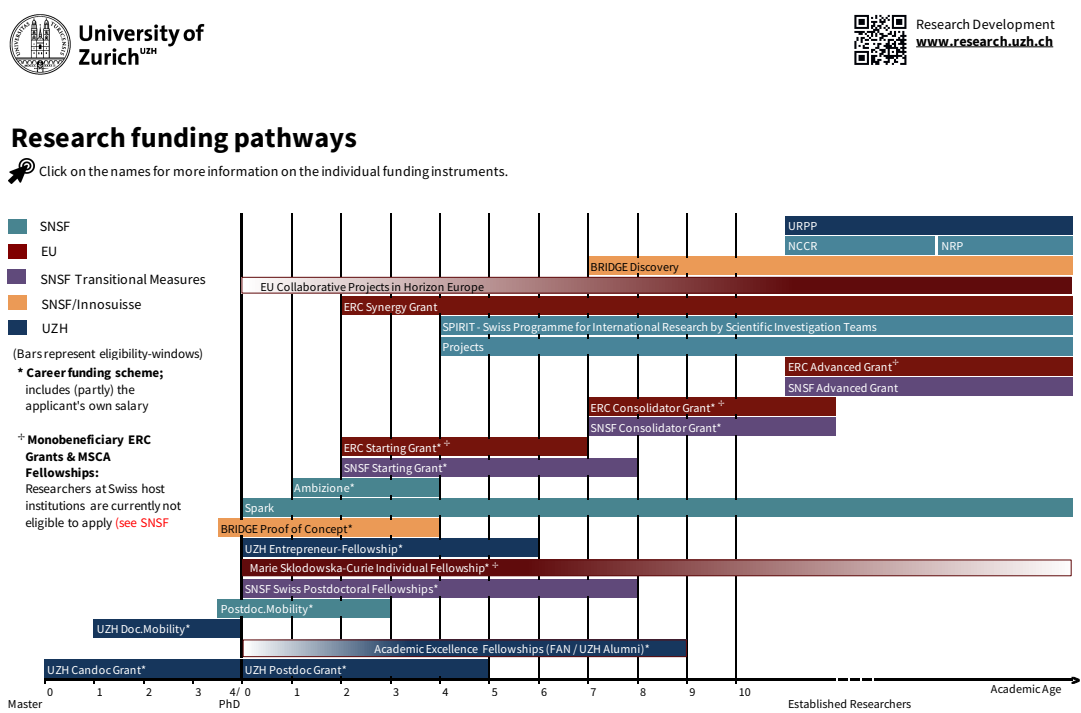


Figure 5: Most common research funding pathways for academic members.

| Program | |
|--|---|
| Alumni-Fonds, UZH | PRIMA (Promoting Women in Academia, SNSF) |
| Ambizione (SNSF) | Privatdozenten-Stiftung, UZH |
| AXA Research Fund | Research Equipment (R'Equip, SNSF) |
| Bridge: Discovery | Scientific Exchanges (SNSF) |
| Bridge: Proof of Concept | SNIS Project Funding |
| Centers of Competence | SNSF Advanced Grant |
| Citizen Science Seed Grants (UZH) | SNSF Consolidator Grant |
| Collegium Helveticum Early-Career Fellowships | SNSF Eccellenaz Professorial Fellowship |
| Collegium Helveticum Senior Fellowships | SNSF Project Funding |
| Competitive Sabbaticals, UZH | SNSF Starting Grant |
| Cotutelle de Thèse | SNSF Swiss Postdoctoral Fellowship |
| DIZH Innovation Program | Spark (SNSF) |
| DSI Infrastructure & Lab Call | SPHN (Swiss Personalized Health Network) |
| ERC Advanced Grant | SPIRIT (SNSF) |
| ERC Grants | Swiss 3R Competence Centre (3RCC) |
| FAN Academic Excellence Fellowships, UZH | Excellence Scholarships for Foreign Scholars |
| Foreign Government Scholarships | Research and Innovation Cooperation Programs |
| Foundation for Research in Science and the Humanities at UZH | Swiss Personalized Health Network (SPHN) |
| GRC Grants (UZH) | Swiss-European Mobility Programme |
| GRC Travel Grant (UZH) | swissuniversities: Project Contributions |
| GSPI Impact Collaboration Programme | TRANSFORM (UZH) |
| Heidi Ras | University Research Priority Programs (URPP) |
| Horizon Europe | UZH Candoc Grant |
| Innosuisse Projects | UZH Doc.Mobility |
| Marie Skłodowska-Curie Postdoctoral Fellowships | UZH Entrepreneur Fellowships |
| National Centers of Competence in Research (NCCR) | UZH Global Strategy and Partnerships Funding Scheme |
| National Research Programmes (NRP) | UZH Index of Foundations (German only) |
| NIH and other US Grants | UZH Postdoc Grant |
| Postdoc.Mobility (SNSF) | |

Table 1: List of funding opportunities which include organisational support by UZH. (Note hyper-links in electronic PDF version.)

10 Key performance indicators

To ensure that our research and clinical implementations are both effective and efficient, the **Precision medicine unit** will track performance using a well-defined set of Key Performance Indicators (KPIs). These KPIs are carefully chosen to reflect crucial aspects of our operations, from data management to user engagement and compliance. We also aim to take the production burden off individuals and instead place it on the joint group effort to determine our own pace, remaining flexible to for worker satisfaction and development.

Our KPI framework includes the following metrics, each designed to provide insights into specific operational aspects:

- **Data Throughput:** Measured by the total gigabytes (GB) of data processed weekly, this indicator helps assess our capacity to handle large-scale genomic datasets.
- **Turnaround Time:** This KPI tracks the duration from sample receipt to the delivery of the report, crucial for evaluating our efficiency in processing and reporting.
- **Research Output:** Annually quantified by the number of papers published and tools developed, reflecting our contribution to scientific knowledge and tool innovation.
- **Error Rate:** This is calculated as the number of errors per 100 analyses, providing a clear measure of the precision and reliability of our data analyses.
- **User Satisfaction:** Determined by averaging scores from user surveys, this metric gauges the satisfaction of clinical staff and researchers with our tools and reports.
- **Compliance:** Measured by the percentage of successful compliance checks, this KPI ensures that our operations adhere to all relevant regulatory and ethical standards.
- **Cost Efficiency:** This is evaluated by the cost per analysis or report, helping us monitor financial efficiency in our operations.
- **Training and Uptake:** The percentage of cases using new tools measures the adoption rate of our latest methodologies and technologies, indicating the effectiveness of our training programs.

11 Single-source management in precision medicine

In the fast-evolving field of precision medicine, managing documentation, data, and software development requires a robust approach to maintain consistency and accuracy across various studies, clinical trials, and software platforms. Our group adopts a **single-source management** philosophy, centralizing shared components to reduce redundancy and prevent discrepancies.

11.1 Principles of single-source management

Single-source management is predicated on the idea that all derivative documents and systems should draw from a common repository of information, ensuring that changes in one area are automatically propagated to all related areas. This approach is particularly advantageous in environments where:

- Documentation needs to be consistent across multiple reports and publications.

- Software developments are closely tied to ongoing clinical trials and studies.
- Data from these trials and studies are frequently updated and must remain synchronized across platforms.

Figure 6 illustrates why this is important. If a user requires a piece of critical information (for example, the number of patients in a study), it must be checked and copied. If that information later updates then all of the manual steps to find that information must also be repeated. When scaled across multiple different projects/uses, it rapidly degrades productive time into “busy work” for the user as well as leading to numerous critical errors. Figure 6 illustrates why this is important. shows a two simple examples of single-source management for two and six use cases, respectively.

If, for example, the contents of “raw data” is updated as a project updates, all of the subsequent uses of this information can be automatically updated in a single step rather than repeating work numerous times.

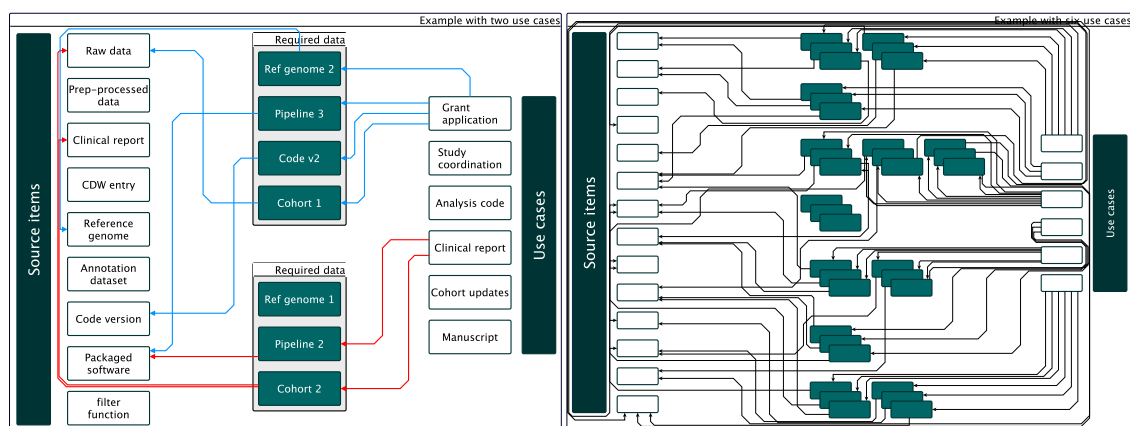


Figure 6: Single-source management examples within the Precision medicine unit. As the number of uses for information increases, the scale of tracking grows massively. Without automation this bottleneck can halt progress.

11.2 Implementation example in our documentation

Our LaTeX documents exemplify this single-source management system. The document that you are reading has been generated using this system. For example, all of the **highlighted key texts** used in this document are sourced from a central repository so that they are always up-to-date. Key elements such as common variables, formatting settings, and version-controlled content are centralized. We maintain a structured directory layout for efficient resource management:

```
/resources
  /head.tex          % Standardized layout and color definitions
  /variables          % Common variables across projects
    /variables.tex    % Definitions of constants, pipeline names, etc.
/main_document_1.tex % Specific document drawing from the resources
/main_document_2.tex % Another document sharing the same resources
```

The same approach is taken in our software development, analysis protocols, data-flow, etc. There is no philosophical separation between any type of information - a document name is treated with the same respect as raw data from a patient. There is a single source of truth for information and all subsequent uses of that information directly source the origin, ideally including tracked changes. Numerous sources of variables can be used, including the database repository, document layouts, and other configurations. The variables source contains declarations for key terms and concepts used across various documents, highlighted using a specific color to indicate their importance and frequent usage. These variables serve not only as identifiers within texts/code/etc. but also help to connect the reader/user with recurring themes and important concepts across multiple documents.

11.3 Benefits

The benefits of single-source management in our precision medicine research are manifold:

- **Increased Efficiency:** Updates need only be made in one place, automatically propagating to all documents and systems.
- **Enhanced Accuracy:** Reduces the risk of discrepancies between documents, as all draw from a unified source.
- **Improved Collaboration:** Team members can work simultaneously on different documents, confident that they are using the most current and consistent information.

Implementing this system requires rigorous discipline and a well-organized file management structure, but the return on investment in terms of time saved and error reduction is substantial.

12 Future directions and innovation

As we look to the future, the **Precision medicine unit** is set to incorporate more advanced technologies, such as artificial intelligence (AI) and machine learning algorithms, into our workflow. These technologies are expected to substantially enhance our capability to analyse and interpret the copious amounts of data we generate, thereby facilitating the development of more accurate diagnostic tools and more efficacious therapeutic strategies.

In parallel with technological expansion, we are planning to initiate new collaborative projects that utilise our multi-omics platforms to delve into previously unexplored areas of paediatric health and disease. These projects will strive to identify novel biomarkers, elucidate complex disease mechanisms, and forge innovative treatments tailored to the genetic and molecular profiles of individual patients.

Our ongoing commitment to innovation and excellence ensures that we will continue to enhance health outcomes for children and further afield, significantly shaping the future of paediatric medicine with each discovery we make.

13 Future products

- Noninvasive prenatal testing using circulating DNA and RNA [2].
- Pharmacogenomics drug dosages.
- Pharmacogenomics drug interactions.
- Pharmacogenomics cost effectiveness.
- DeepInfer

13.1 Pharmacogenomics at Kispi

Pharmacogenomics has the power to find direct effects of personal genetics on the metabolism and effectiveness of drugs used in the clinic. **Figure 7** illustrates an example of variant effect from gene-to-protein and how some gene-drug interactions are identified. Our pharmacogenomics strategy has a tripartite approach which is designed to personalise and enhance treatment efficacy while managing healthcare expenditures effectively. Pirmohamed [3] has reviewed the current status and future perspectives of such approaches.

Firstly, the pharmacogenomics of drug dosages focuses on adjusting medication based on genetic profiles to optimise therapeutic effects and minimise adverse reactions [4]. By analysing a patient's genetic markers, the **Precision medicine unit** can predict metabolic rates and drug absorption, ensuring that each child receives a dosage that maximises benefit while reducing the risk of toxicity. This tailored approach not only promises better health outcomes but also reduces the trial and error typically associated with standard dosing practices.

Secondly, understanding pharmacogenomic drug interactions is crucial in a paediatric setting where polypharmacy can be common, especially in complex cases such as cancer or chronic diseases. The **Precision medicine unit** leverages genetic insights to foresee and mitigate adverse drug-drug interactions. This is achieved by mapping genetic variants that affect drug metabolism enzymes, transporters, and receptors, thereby providing a clear picture of potential interactions before they can cause harm.

Lastly, the aspect of cost-effectiveness in pharmacogenomics cannot be overstated. By integrating pharmacogenomic data into clinical decision-making, the **Precision medicine unit** aims to reduce not only the financial burden associated with ineffective treatment but also the costs related to adverse drug reactions. This strategic incorporation of pharmacogenomics is expected to lead to more efficient use of healthcare resources, shorter hospital stays, and fewer unnecessary tests and procedures [5].

The **Cytochrome P450** (CYP) enzymes are an especially useful target of pharmacogenomics. These are crucial for the metabolism of many xenobiotics and endogenous compounds, making them key players in drug metabolism. Variations in the 15 encoding genes significantly influence drug interactions and efficacy. The **Pharmacogene Variation** (PharmVar) consortium maintains a repository that catalogues these genetic variations, aiming to standardize nomenclature across the pharmacogenetics community [6]. This facilitates understanding and predicting drug metabolism, disposition, and response. Other essential resources in this field include the **Pharmacogenomic KnowledgeBase**, <https://go.drugbank.com>, and the **Clinical Pharmacogenetic Implementation Consortium**, which further support pharmacogenomic research and clinical implementation.

Our methods have taught for the last six years (2019-2024) in the EPFL master's degree coursework and

hands-on projects for [BIO491](#) - MSc New tools & research strategies in personalized health and [tutorials](#).

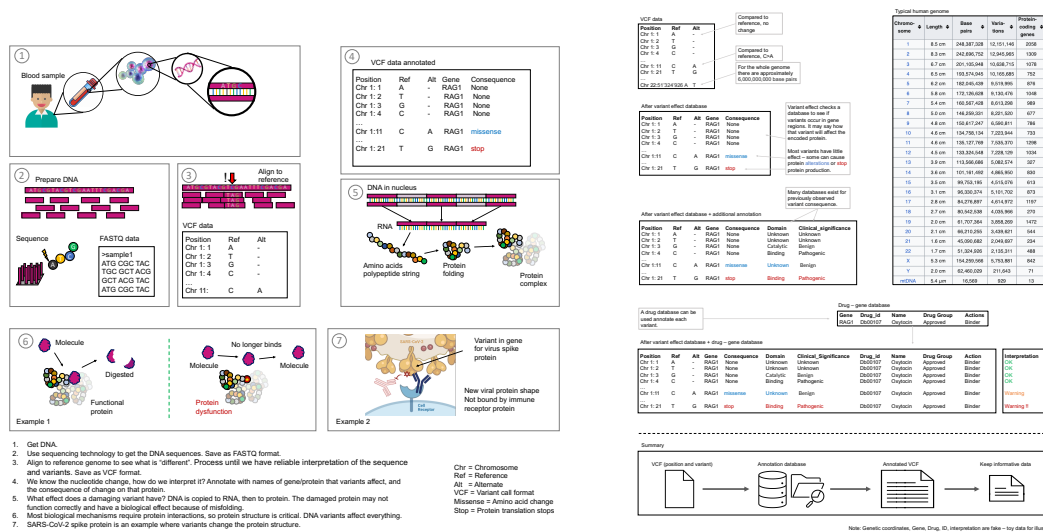


Figure 7: **Precision medicine unit** pharmacogenomics. Known genetic variants produce measurable effects on protein and function. By applying the pharmacogeomic knowledge base to patient's personal genomic data, we can determine the gene-drug effect. This allows us to provide information about drug metabolism, dosage, indication, and cost effectiveness.

13.2 DeepInfer

A significant development within the **Precision medicine unit** is the **DeepInfer (v1.0)** initiative, which embodies our forward-thinking approach in genetic research. This project aims to pre-calculate the probability of each genetic variant, whether known or novel, in causing any given set of diseases using a comprehensive Bayesian framework. This initiative sets a new benchmark in predictive medicine by integrating extensive prior knowledge and observational data into a cohesive statistical model.

The methodology behind **DeepInfer (v1.0)** involves several rigorous steps:

- Variant Data Collection and Annotation:** All known nucleotide variants are collected and annotated using tools such as VariantAnnotation. This annotation details the functional consequences of each variant, providing a foundational dataset for subsequent analyses.
- Probability Estimation:** We estimate the frequency of each variant using population data from databases like gnomAD. Additionally, we calculate probabilities for random, novel variants that may not yet be catalogued.
- Incorporation of Prior Information:** The model integrates biological data (e.g., clinical pathogenicity from ClinVar and structural data from UniProt) as Bayesian priors. This step enhances the model's ability to provide accurate estimates by including comprehensive background information.
- Bayesian Inference:** Our approach updates the probability of each variant causing specific diseases by applying Bayesian inference techniques. We use statistical packages such as brms and rstan for this purpose, employing Bayes' theorem to combine prior knowledge with the observed data, resulting in a detailed set of posterior probabilities.

This quantitative expression of disease association likelihood for each variant allows for preemptive strategies in pediatric care tailored to the genetic profiles of individual patients, significantly transforming the landscape of pediatric healthcare. The **DeepInfer (v1.0)** initiative not only broadens our understanding of genetic influences on disease but also markedly enhances our capacity to intervene effectively before clinical manifestations occur.

By continuing to refine and expand the **DeepInfer (v1.0)** initiative, we anticipate it will become a cornerstone in our predictive capabilities, providing new insights into the genetic basis of diseases and enhancing our ability to offer personalized care.

14 Benefit analysis

Economic impact of sepsis - [7]

- **Patient data and impact:** Review covers studies reporting costs associated with adult sepsis patients globally, indicating significant economic burden.
- **Methodology:** Systematic review under PRISMA guidelines, with data from PubMed, EMBASE, and Cochrane.
- **Key statistics:** Reports wide cost range from €1,101 to €91,951 per sepsis patient, reflecting international healthcare system variances.

Genomic insights in critical care for infants - [8]

- **Patient data and impact:** Involves 278 critically ill infants; molecular diagnosis achieved in 36.7% of cases, with higher rates (50.8%) in critical trio exome cases.
- **Methodology:** Utilises proband exome, trio exome, and critical trio exome sequencing.
- **Key statistics:** Impacted medical management in 52.0% of diagnosed cases.

National scale multi-omics for rare diseases - [9]

- **Patient data and impact:** Involves 290 critically ill infants and children; diagnostic yield from WGS initially at 47%, increased to 54% with extended analysis.
- **Methodology:** Rapid whole-genome sequencing integrated with transcriptomic data.
- **Key statistics:** Altered critical care management in 77% of diagnosed cases.

Genomic lifespan association in Iceland - [10]

- **Patient data and impact:** Study includes 57,933 participants, identifying 2,306 individuals with actionable genotypes.
- **Methodology:** Whole-genome sequencing focusing on 73 genes from ACMG Secondary Findings recommendations.
- **Key statistics:** Actionable genotypes linked to a decrease in median lifespan by approximately three years for carriers.

Genome analysis in neurodevelopmental disorders - [11]

- **Patient data and impact:** Encompasses 465 families, finding causal variants in 36% of 489 affected individuals.
- **Methodology:** Combines short-read and long-read genome sequencing.
- **Key statistics:** Long-read sequencing crucial for resolving complex variants.

Rapid whole-genome sequencing in UAE - [12]

- **Patient data and impact:** Initial feasibility study with five infants, three of whom were successfully diagnosed.
- **Methodology:** Rapid whole-genome sequencing with a turnaround of 37 hours on average.
- **Key statistics:** Demonstrates the utility of rWGS in diverse populations within a Middle-Eastern context.

Genome sequencing for rare diseases - [13]

- **Patient data and impact:** Involves 822 families with rare monogenic diseases, achieving a diagnostic yield of 29.3%.
- **Methodology:** Focuses on broader genomic coverage including structural and non-coding variants.
- **Key statistics:** Identifies causative variants in 8.2% of cases previously undetected by exome sequencing.

UK and Ireland genomic diagnostics in paediatrics - [14]

- **Patient data and impact:** The DDD study involved over 13,500 families; approximately 41% of probands received a genetic diagnosis.
- **Methodology:** Integrated genomic data analysis with clinical phenotyping.
- **Key statistics:** Highlights cost-saving potential through precise diagnostics and targeted therapy.

These studies show the potential of precision medicine, particularly through genomic sequencing and multi-omic technologies, in diagnosing and managing rare and complex conditions. By implementing known methods, we can improve patient outcomes, enable targeted therapies, and potentially reduce healthcare costs through more accurate and faster diagnoses. This integration promises not only to improve patient care but also to provide critical insights into the genetic basis of diseases, ultimately informing both treatment and prevention strategies.

The **Precision medicine unit** cost analysis overview is shown in figure 8. Sepsis was specifically modelled using federal statistics from Bundesamt für Gesundheit (BAG, Federal office for public health) from 2010-2022 as shown in figure 9, 10, 11.

A forecast model from 2010-2030 to the number of deaths due to sepsis in Kispi is shown in figure 1.

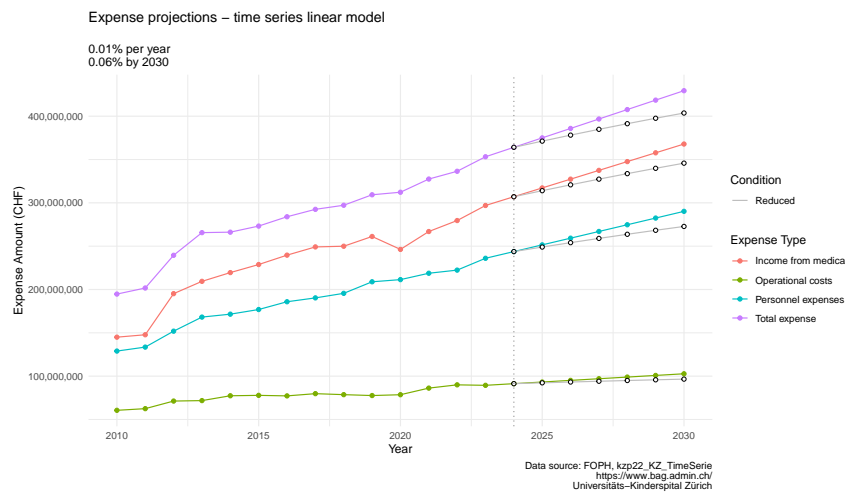


Figure 8: Projections for Kispi with a **Precision medicine unit** (2010-2030). Federal statistics from Bundesamt für Gesundheit (BAG, Federal office for public health) were modelled and projected to 2030. A modest benefit effect size was applied. In the most similar application to our **Precision medicine unit**; Lunke et al. [9] showed a 54% diagnostic yield and an altered critical care management in 77% of diagnosed cases. We estimated modest 1% increase in actualised savings per year after successful implementation starting in 2024.

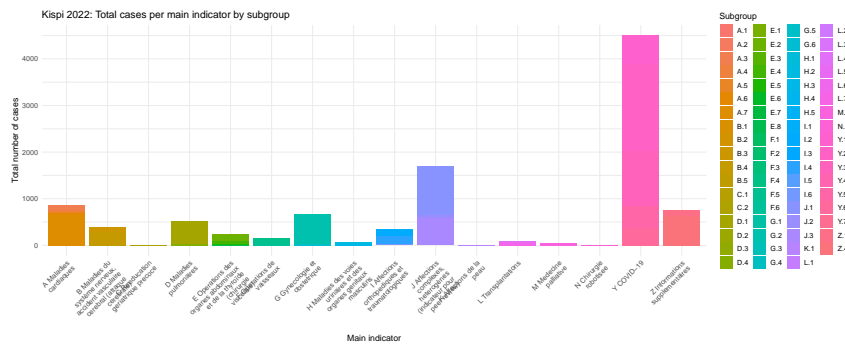


Figure 9

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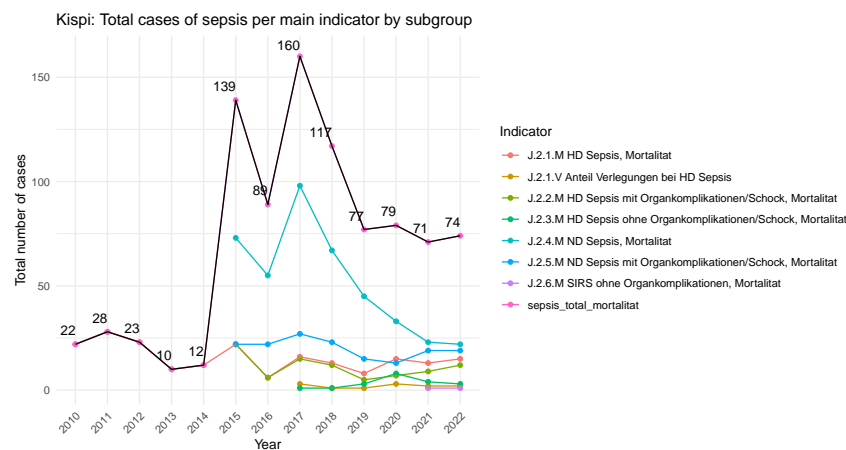


Figure 10

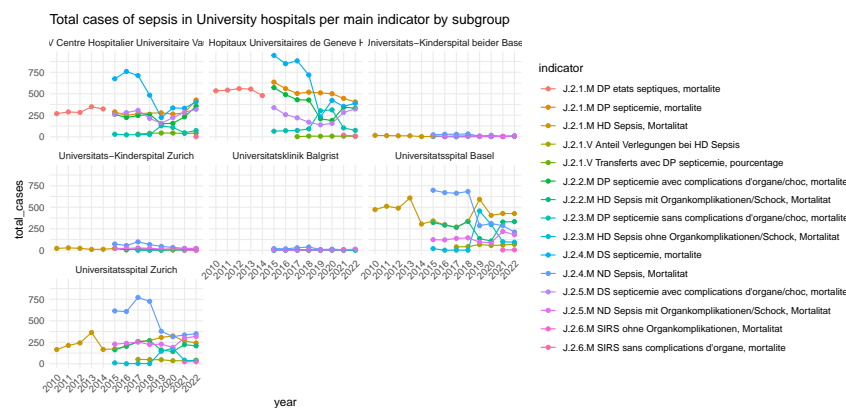


Figure 11

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