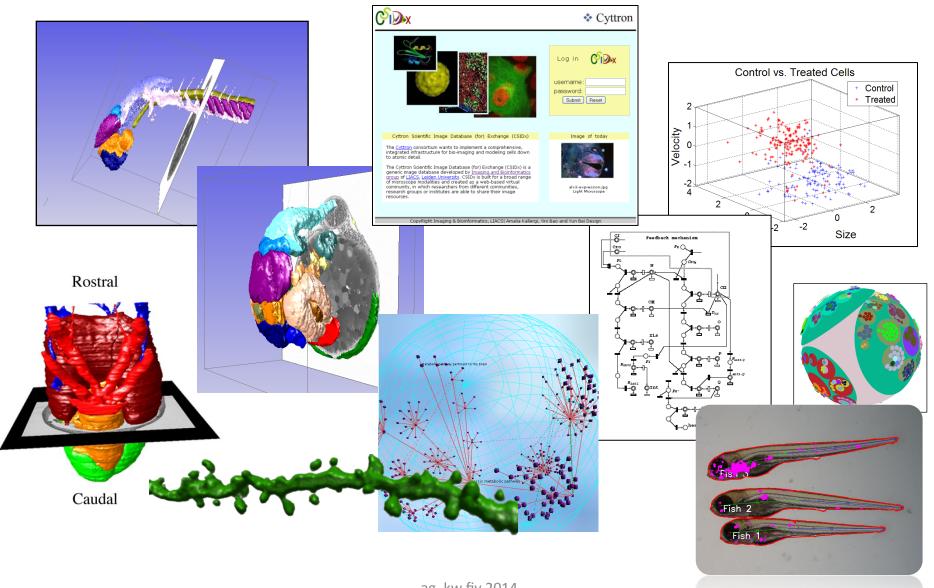
Imaging & BioInformatics

FJ. Verbeek, K. Wolstencroft, A. Gultyaev,J Slob, R Carvalho, L. Cao,C. Fuyu, E Larios, M. Tleis, Z. Xiong

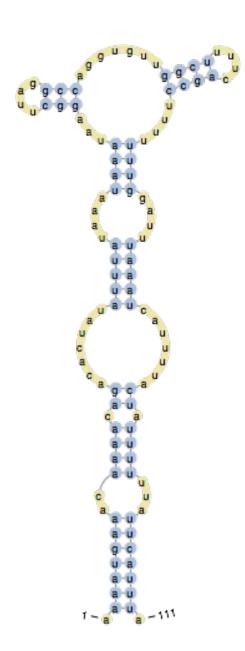
Imaging & BioInformatics @ LIACS



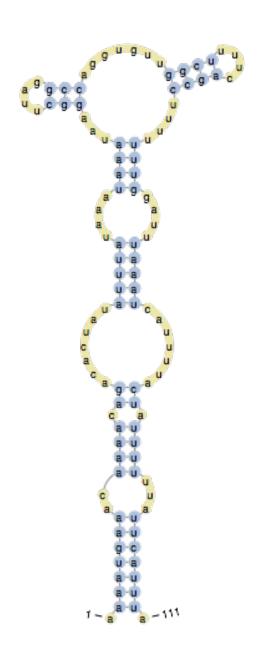
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RNOMICS

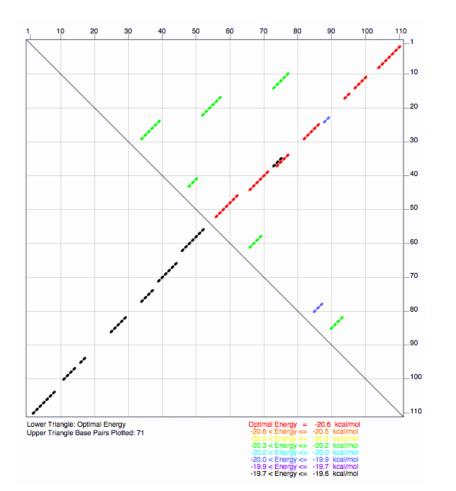
RNomics: identification of functional RNAs encoded in genomes



RNA secondary structure prediction



Dynamic programming algorithm can compute both optimal and suboptimal structures that can be shown in various ways in e.g. "dot-plots":



Comparative RNA structure analysis

A powerful approach in RNA structure prediction, in particular, due to RNA-specific patterns of variation, nucleotide covariations.

An example of two covariations in three related RNA's:

n n	n n	n n
n n	n n	n n
n-n	n-n	n-n
G-C	U-A	A-U
n-n	n-n	n-n
n-n	n-n	n-n
A-U	G-C	C-G
RNA 1	RNA 2	RNA 3
A nn G nnnnnn C nn U	RNA 1	
Gnn U nnnnnn A nn C	RNA 2	
${f C}$ nn ${f A}$ nnnnnn ${f U}$ nn ${f G}$	RNA 3	
((((()))))	consensus	"bracket view"

Detecting conserved structures in related RNAs (prediction of "consensus" structures)

Consensus structures can be computed from sequence alignments using information from suboptimal structures, base probabilities and covariation patterns

Input: Sequence alignment

Calculation: suboptimal structures/partition functions/base probabilities for individual sequences; detection of common patterns and their scoring

Output: The "consensus" structure, (ideally) conserved in all sequences of the dataset.

For instance, a fragment of the output of RNAalifold algorithm:

```
NP_gullMD77/1-1565
NP_gsGD96/1-1565
NP_eqMiami63/1-1565
NP_Victoria75/1-1565
NP_swTN77/1-1565
```



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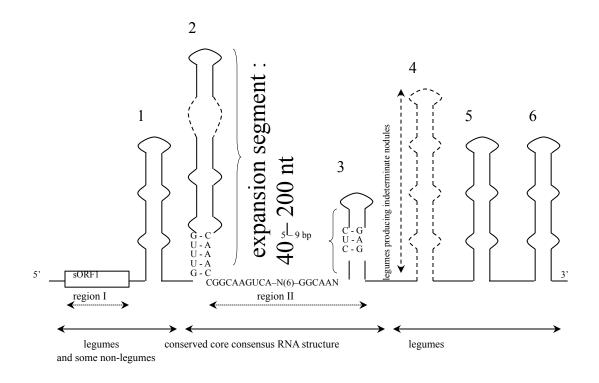
```
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NP_gsGD96/1-1565
NP_eqMiami63/1-1565
NP_victoria75/1-1565
NP_swTN77/1-1565
NP_swTN77/1-15
```

Such structure-annotated alignments allow one to identify covariations.

Plant enod40 gene

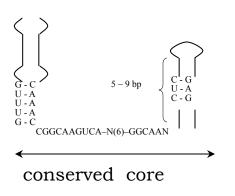
- One of the most intriguing genes involved in the regulation of symbiotic interaction between plants and bacteria or fungi;
- Initially identified at the early stages of formation of nitrogen-fixing root nodules of legume plants;
- Also found in non-legume species;
- Apparently has a dual function:
 - encodes small conserved peptides
 - contains conserved RNA structural elements

- Enod40 RNA is highly structured in legumes.
- The consensus core structure turns out to be very conserved in other plants as well.



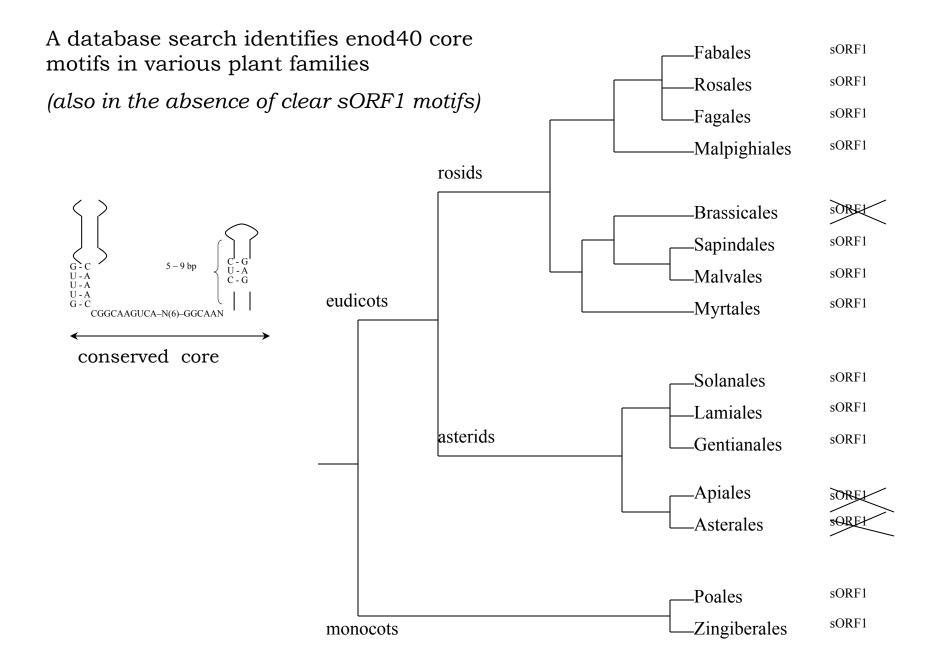
(Gultyaev & Roussis, 2007)

- Enod40 RNA is highly structured in legumes.
- The consensus core structure turns out to be very conserved in other plants as well.
- This consensus can be used for the search of new enod40 genes in the nucleotide databases (GenBank, EST, WGS).

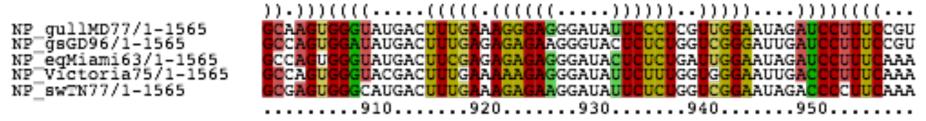


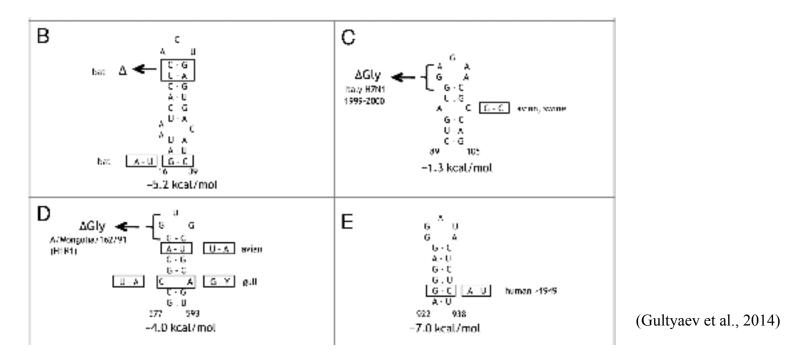
A search protocol:

- -Using the conserved core sequence of a known enod40 RNA as a query, BLAST the databases.
- Analyze the structures flanking the putative enod40 RNA cores in the BLAST hits (including those with rather high E-values). The presence of a structure consistent with consensus is an evidence for enod40 motif.
- New (putative) enod40 sequences are used again as BLAST queries to find enod40s in other species.
- NB. Due to low sequence similarities, a straightforward BLAST is not efficient (reliable BLAST hits with low E-values are yielded only for close relatives, while many hits with high E-values are false-positive results).



Detecting conserved structures in the influenza virus genome





Important similarities and differences between virus strains can be identified

WORKFLOWS & SEMANTIC INTEGRATION

Data and Knowledge Integration

Scientific workflows

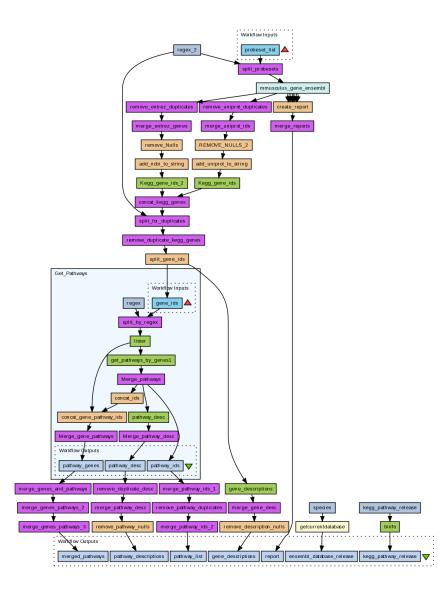
- Distributed computing
- Experimental reproducibility
- Analysing and processing high-throughput (bio)informatics data

Semantic integration

- Using semantic annotation to explore and understand complex, heterogeneous biological data
- Make new connections and inferences from existing data

Taverna Scientific Workflows

- Sophisticated analysis pipelines
- A set of services to analyse or manage data (either local or remote Web services)
- Automation of data flow
- Iteration over data sets
- Control of service invocation
- Provenance collection
- Experimental protocols



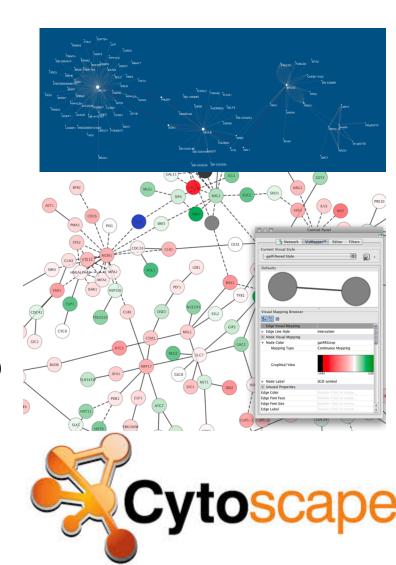
Taverna Database Integration

- Data processing is by strings and lists
 - Data transformation services required
 - What we often require are tables
- Develop a plugin to enable database population from Taverna workflows
 - local desktop
 - Server, cloud/grid
- Evaluation with case studies in high throughput bioinformatics analysis

Taverna and Cytoscape Integration

Cytoscape Network Visualisation tool

- Execute Taverna workflows through a cytoscape plugin to allow ingestion and integration of multiple data sources ('biological layers')
- Explore plugin in relation to systems biology case studies



Semantic Annotation and Matching of CellML Models

CellML is an XML standard for describing mathematical models of biological processes (e.g. cardiovascular circulation, metabolism, neurobiology).

- Over 1000 published models
- Models have structured metadata, but biological entities have no semantic descriptions.
 - Difficult to find relevant models and identify where models may overlap
 - Difficult to find experimental data to compare to model simulations

Objectives:

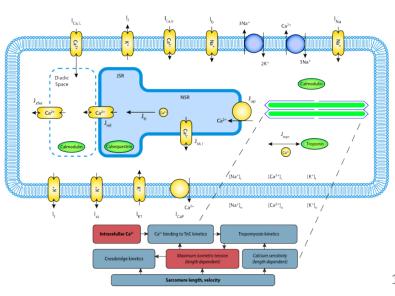
- Develop an automated approach to semantically annotating CellML models with terms from biological ontologies
- Perform semantic matching and similarity measures between CellML models
- Link models to relevant data, using scientific workflows

Prerequisites (helpful, not mandatory):

- Databases
- Ontologies
- Classification and semantic similarity
- An interest in bioinformatics

Advisor: Dr Katy Wolstencroft

Contact: k.j.wolstencroft@liacs.leidenuniv.nl



COMPUTATION

Leiden Life Sciences Cluster (LLSC)



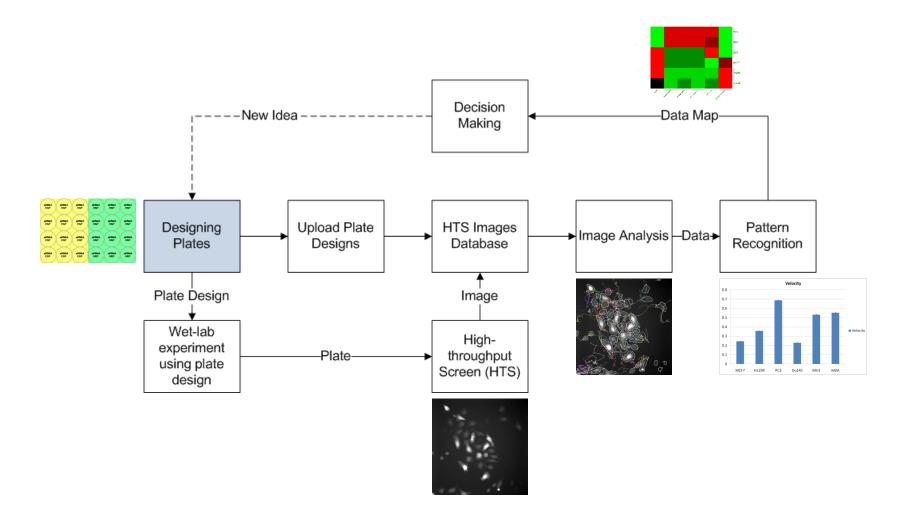
LLSC: Scientific Cluster

- 48 Dell 2u 8x cores
- Cluster
 - Head Node
 - Main Node
 - Worker Nodes
- Map computational intensive programs to Cluster
 - image processing/analysis
 - pattern recognition
 - bioinformatics jobs from workflows
- RESEARCH
 - Develop strategy/template for web-services
 - Develop strategy for Parallelization

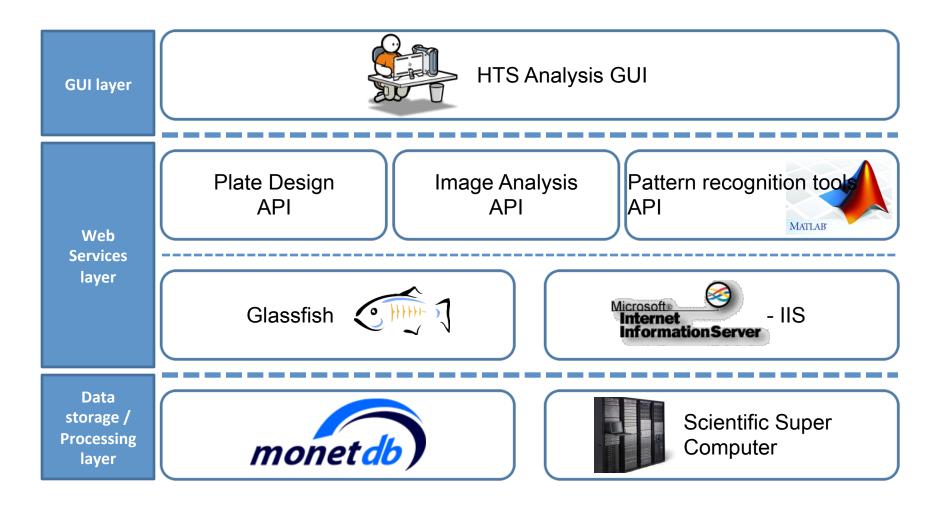
Cluster Computation

- NeCEN
 - Netherlands Centre for Electron Nanoscopy
- Cell Observatory
 - Collaboration within Faculty of Science
 - IBL, LIC, LION, LIACS
- High-Throughput imaging
 - 3D zebrafish imaging

Systems for High-Throughput



Description of Solution



Specific Projects

- API from database to Cluster
 - API Architecture; Requirements
 - Technical Solutions, builds on previous results
 - What is the architecture required from MonetDB to HTS-processing
- Image Processing Software
 - Datastructures / Code optimization
 - Better fit it for cluster/parallel computation
 - Where and how can we get computational benefit

Project Summary

- Analyze software to be ported to the Cluster
- Develop mapping to the cluster architecture
- Apply the results to a dataset
- Generalize
- Collaboration with Kris Rietveld

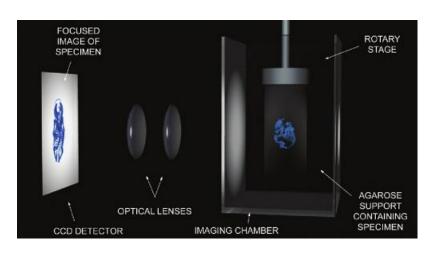
– NeCEN: Java, C, Python

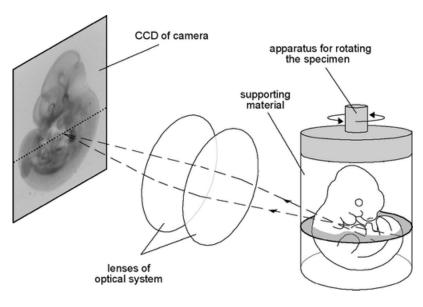
– Cell Observatory: Java

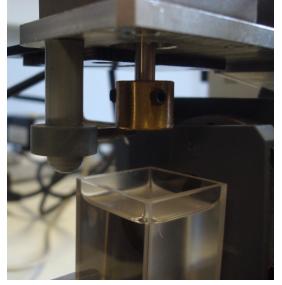
– High Troughput: Java, Python, SQL

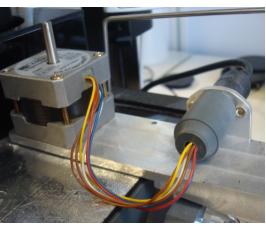
DATA ACQUISITION

OPT microscopy



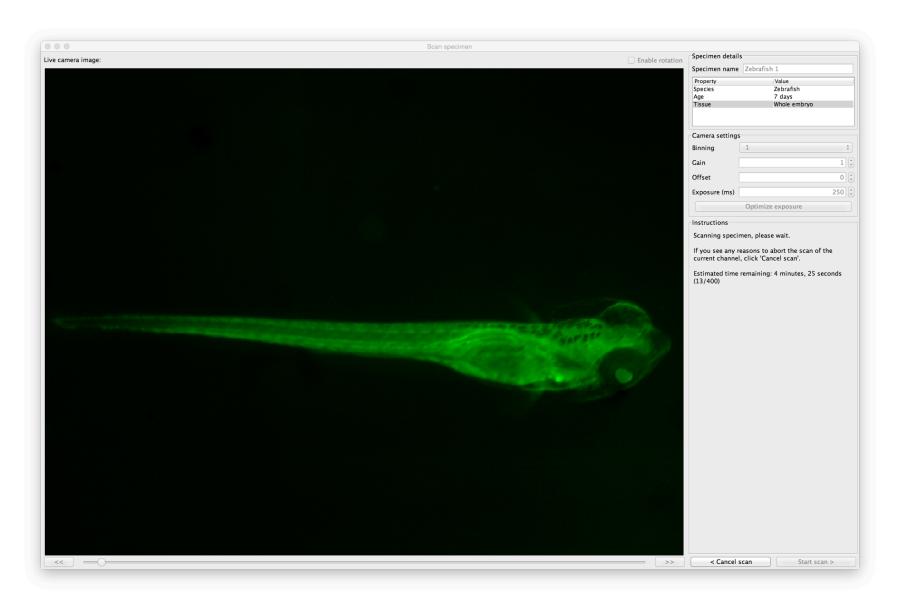






Object Clearance

OPT Software - Method



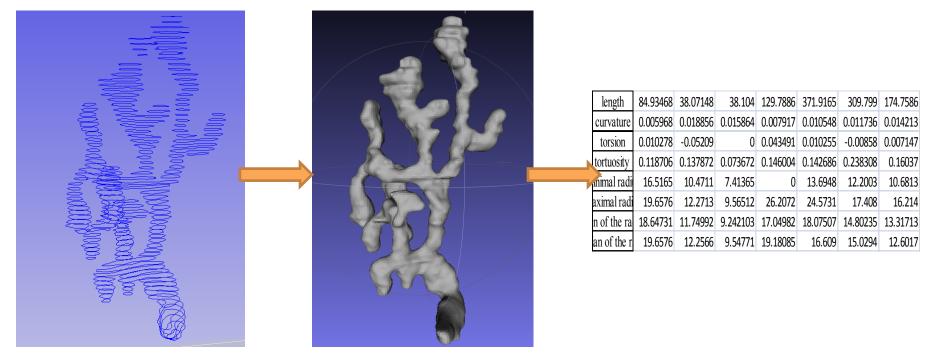
Project Summary

- Calibration system optimization
- Develop Back Projection of tomogram to 3D volume = inverse Radon Transform
- Investigate parallelisation
- Implement on Cluster
- Workflow optimazation for the imaging
- Project with Sander Hille (Mathematics)
- Dubble Bachelor team Math-CS

ANALYSIS & VISUALIZATION

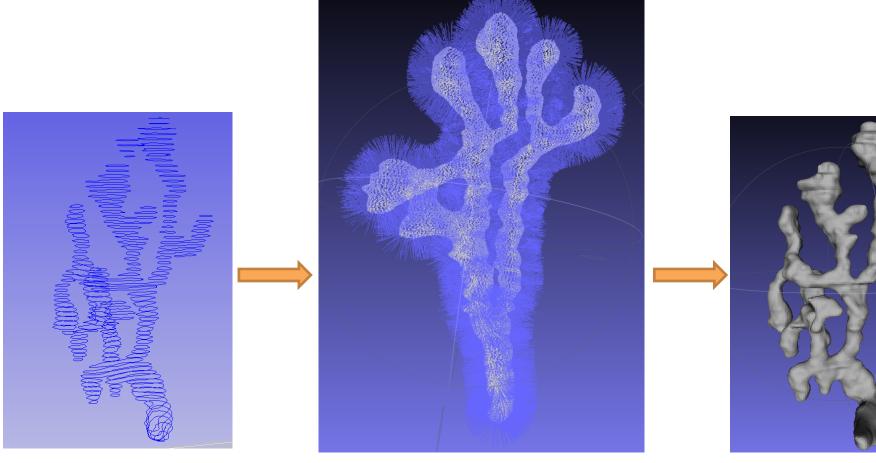
3D Model optimization

- We start from 3D models obtained from various sources
- Measurement



Surface Reconstruction

Poisson Reconstruction



Contour representation

Point Cloud representation

Surface representation

ag, kw fjv 2014

Connecting Components

- Read files
- Reconstruction & Optimization
 - Poisson reconstruction
 - (L.Cao & FJ Verbeek, Electronic Imaging 2013)
- Visualization
 - VTK (visualization toolkit)
 - Geometrical data-structures
- GUI components

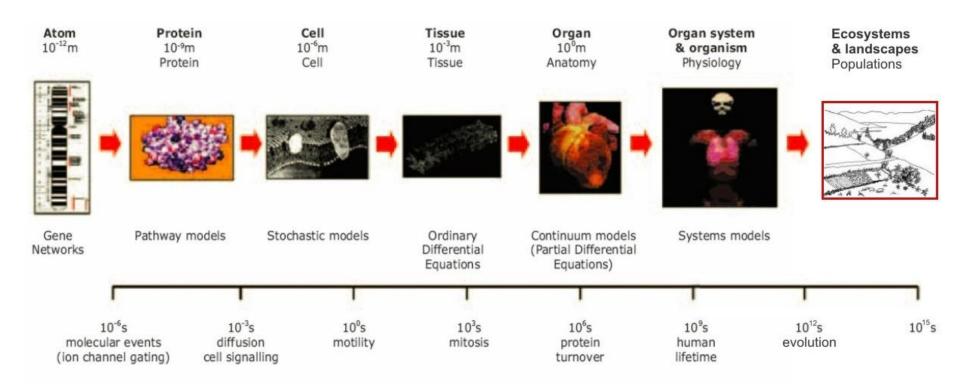
Project Summary

- Analyze components
- Develop infrastructure that fits workflow for 3D modelling
- Develop interface
- Connect components

• C, C++, QT, VTK

SOFTWARE AGENTS

Modeling Different Levels of Biology



Intelligent Software Agents

- Images are found on different scales in biology
- These scales need to be connected
- Force the connection with software agent
- Annotate large collections
- Integrate in existing database (Cyttron)
- Test and develop agent
- From prototype increase intelligence.

Contacts

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 - k.j.wolstencroft@liacs.leidenuniv.nl
- Sacha Gultyaev
 - A.P.Gultyaev@liacs.leidenuniv.nl

URL

http://bio-imaging.liacs.nl/projects