

GE32/MM12 - Data Reuse

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Availability of slides

- All materials are freely available (CC BY) after the lectures:
 - StudIP: GE32/MM12
 - GitHub: https://github.com/bpucker/teaching
- Questions: Feel free to ask at any time
- Feedback, comments, or questions: b.pucker[a]tu-braunschweig.de

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What are the advantages of re-use?



What are the advantages of re-use?

- Cost-effective
- Immediately available
- Extremely large datasets



Challenges with data re-use?

- Lacking metadata
- Unknown details/issues
- Mislabeling possible
- Not perfectly matching needs
- Technology outdated



Find gene expression data sets

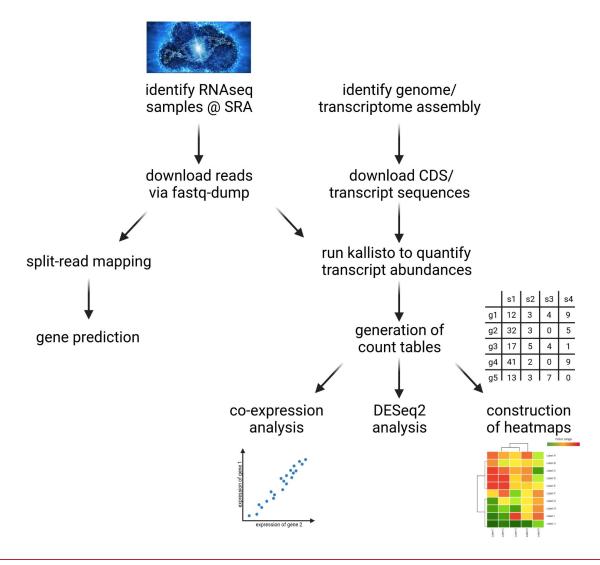
- SRA read selector
- Gene Expression Omnibus (GEO) search
- Publications: data availability statements & supplementary tables

How to retrieve data?

- Preprocessed data sets (count tables @ GEO)
- Cloud solutions (Galaxy)
- Fastq-dump



Workflow



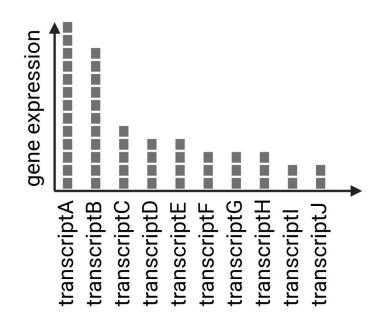


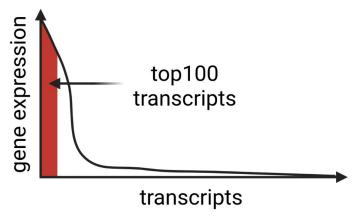
How to check RNA-seq data sets?



RNA-seq quality control

- Percentage of reads mapped to individual mRNA sequences
- Distribution of abundance across transcripts
 - Substantial coverage of top100 transcripts
- Metadata assessment / marker gene check
 - Highly expressed marker genes like RuBisCO







How to analyze the distribution of species?



GBIF

- Sources have heterogeneous quality
- Extensive filtering required
- What types of issues can be expected?



GBIF

- Coordinates located in the sea
- Coordinates located in zoos/botanical gardens
- Coordinates located at the center of grids/in capitals
- Entries of fossils / too old entries
- Entries containing likely typos

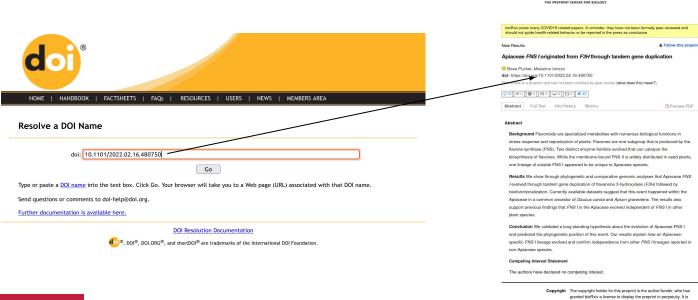


How to cite data sets?



Digital Object Identifier (DOI)

- DOI = Digital Object Identifier
- Unique and short way to point to a publication or a data set
- How to resolve a DOI? https://dx.doi.org/



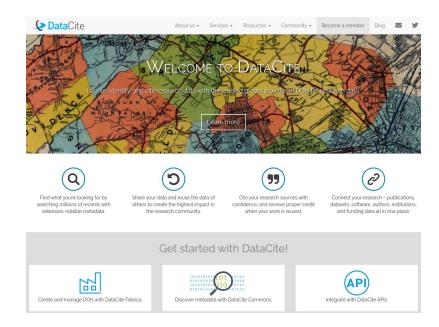
CSH Spring bio Rχiv

made available under a CC-BY 4.0 Interns



DataCite

- Central service for generation and management of DOIs for research data sets
- Enable connection and reuse of data sets





Managing references

- Entries about publications / data sets are stored in a local database
- Convenient citation when writing manuscripts
- Support for comments and key words assigned to entries
- Examples:
 - Zotero: free
 - Mendeley: free, but belongs to Elsevier
 - Citavi: commercial, but campus license
 - EndNote: commercial



Data reuse examples



Benchmarking of NOVOPlasty

- Public sequence read data sets are used for plastome assemblies
- Data sets can be selected based on specific criteria
- Pure bioinformatics groups can work with real data sets
- No costs for generation of data sets



Benchmarking SANPolyA

- SANPolyA detects Poly(A) signals through deep learning
- Comparison of SANPolyA results against results of other tools
- Freely available output of tools is required for systematic comparison



Pangenome of hexaploid bread wheat

- Pangenome = combination of all genomes of a species
- Integration of public data sets to identify presence/absence of genes in wheat cultivars
- Power of pangenomes increases with number of analyzed



Single plant GWAS + bulk segregant analysis

- spGWAS = single plant genome-wide association studies
- Bulk segregant analysis = comparisons of plant groups with contrasting phenotypes
- Comparison of different approaches to identify genetic basis of a trait (plant height)
- Evaluation of findings against previous reports



Co-expression gene network analysis

- Genes belonging to the same pathway show similar expression patterns
- Bamboo genome sequence was integrated with RNA-seq data sets
- Identification of co-expression modules associated with development



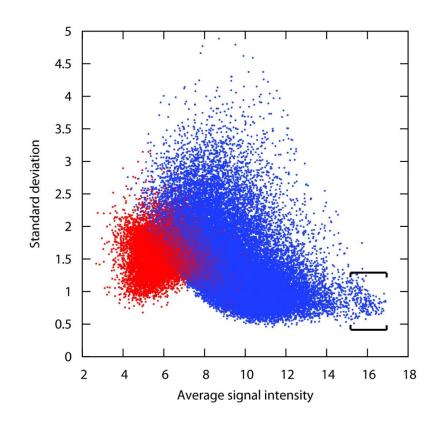
Integration of GWAS and co-expression analysis

- GWAS allows the identification of genomic regions associated with a trait
- Co-expression analysis can help to identify individual genes in these regions
- RNA-seq data sets of most species are available for free



Identification of RT-qPCR reference genes

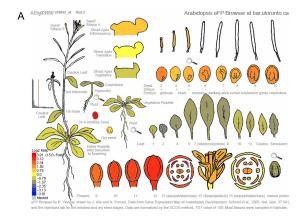
- RNA-seq data sets can be analyzed to identify constantly expressed genes
- Reference genes with constant expression across samples is crucial for RT-qPCRs
- Reference genes can be specific for certain conditions/tissues

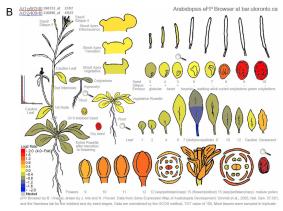




Electronic Fluorescent Pictograph browser

- Integration of existing microarray datasets
- Option to integrate other large scale datasets
- Basis for hypothesis generation
- Visualization of gene expression in many different plant tissues/conditions

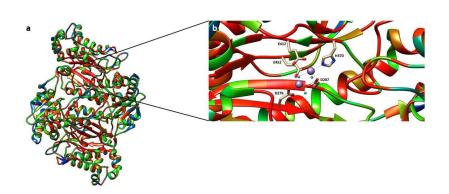






Identification of conserved amino acid residues

- Identification of orthologous sequences in hundreds of species
- Comparison of sequences to identify highly conserved amino acid residues
- 3D modeling based on known structures of homologs



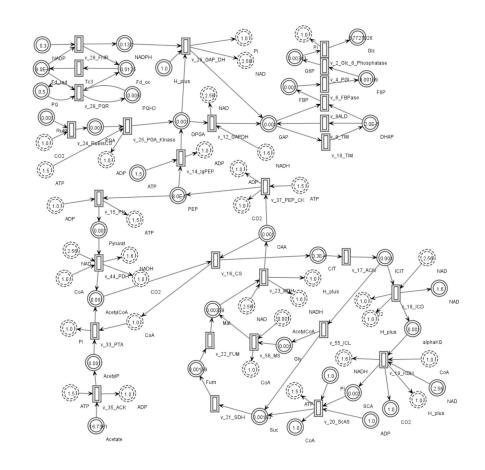
D276	94	99	100	100	100
D287	94	98	100	100	100
H370	94	98	100	100	100
E412	94	96	100	100	100
E452	91	97	100	100	100
T289	94	97	100	100	97
T410	93	96	100	79	100
H377	94	98	100	100	97
R398	93	98	89	10	57
W107	88	98	96	0	96
Y241	94	96	100	2	90
1244	93	98	97	88	100
H255	94	98	100	100	100
V376	89	1	38	81	94
C58	58	64	0	0	0
C158	40	1	0	0	0
	Animals	Plants	Fungi	Archaea	Bacteria



Schilbert et al., 2018: 10.1101/423475

VANESA: enzyme properties for model

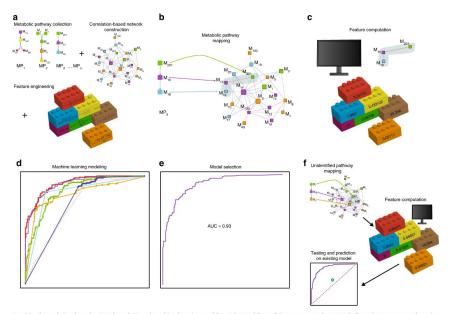
- Pathway structure taken from KEGG
- Integration of enzymatic properties (K_M, V_{max}) from BRENDA
- Freely accessible metabolic modeling solution





Inference of metabolic pathways from metabolite data

- Correlation-based inference of metabolic pathways from metabolite data sets
- Machine learning was deployed
- Identification of the most important features

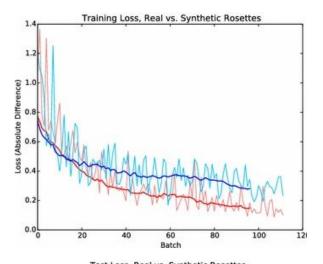


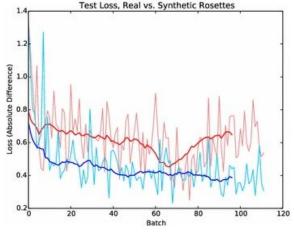
combined correlation-based network analysis and machine learning workflow. The workflow of the current study: a Metabolic pathways were gathered from existing repositories. In parallel, correlation-based networks of metabolites were constructed for the tissue of the organism of interest (here, the tomato pericarp). In addition, a vector of features was engineered based on network properties. b Metabolic pathways with partial to full coverage in the correlation networks were mapped to the networks. Each pathway was considered as a single instance. Training and test sets were proposed based on the existence of the pathways in the tomato. c A set of features was computed for each instance in the training set (for the current study 148 * 3 networks = 444 features in total). d The training set was used to generate different ML models. The model that generated the best performance measures (the AUC) was selected. The ML model was validated in silico using cross-validation. f Test set instances were mapped onto the networks with subsequent feature computation. The proposed ML model was used to predict the potential existence of unidentified pathways in the tomato pericarp



Leaf counting

- Automatic leaf counting in rosette plants
- Deep convolutional neural networks require large and diverse data sets for optimization
- Synthetic plants are proposed as an alternative



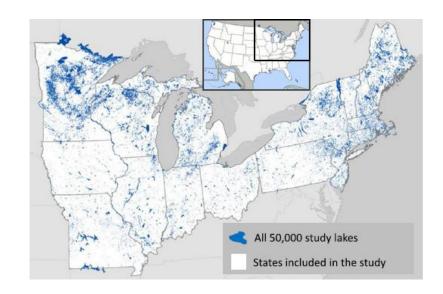


Ubbens, et al., 2018: 10.1186/s13007-018-0273-z



LAGOS: Ecology database

- LAGOS = LAke multi-scaled GeOSpatial and temporal database
- Combination of site-based ecosystem datasets with national geospatial datasets
- Best practice example for lake related data sets covering about 50,000 lakes
- 100 individual datasets about water quality





Time for questions!



Questions

- 1. What are the advantages/disadvantages of reuse?
- 2. How can you assess the quality of RNA-seq data sets?
- 3. How to filter GBIF data sets?
- 4. How can you cite data sets?
- 5. Which studies used existing data sets to gain novel insights?

