

UiO : **Department of Biosciences**
University of Oslo

MBV4410/9410 Fall 2016

Bioinformatics for Molecular Biology



General information

Course coordinator: Jon Bråte

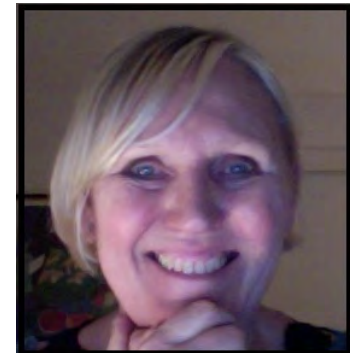
Email: jon.brate@ibv.uio.no

Phone: +47 922 44 582



Course administrator: Torill Rørtveit

Email: torill.rortveit@ibv.uio.no



Course web page:

https://wiki.uio.no/projects/clsi/index.php/MBV-INF410_2016

Purpose of the course

Goals:

- Learn how to *obtain* and *analyse* different types of *biological data*
- Learn basic file handling, and how to run and use programs on the Unix command line

Background:

- Primarily molecular biology and biochemistry. No programming skills required

General information

	Level	Credits	Exam	Oblig	Extra oblig
MBV-INF4410	MSc	10	Yes	Yes	No
MBV-INF9410	PhD	10	Yes	Yes	Yes (at least 2500 words)

Home exam:

- Sent out to all participants at 15:00 Friday December 9 by email
- Must be returned at latest 15:00 Friday December 16 by email (NB! To Torill – not Jon!)

Oblig:

- Assignment (including essay) must be returned by 23:59 Friday December 2 at by email

Obligatory assignment (“oblig”)

- Exercise for oblig will be handed out at the end of course week 2.
- Will be relatively easy and similar to exercises in course weeks 1 and 2.
- Must be returned before the first lecture in course week 4 (December 5).
- PhD students (MBV-INF9410) must in addition write an essay (> 2500 words).
 - Describe how you would use 2 or more of the methods covered in the course in your own research.
- **Obligatory assignment must be approved before you can take the exam!**

Exam

The exam for this course will be a week long take-home exam. Only students who have completed and passed the obligatory assignment are allowed to take the exam.

The exam will be sent to all qualified participants at 15:00 December 9 by email.

The completed exam must be returned at latest 15:00 on December 16 by email to Torill Rørtveit (torill.rortveit@ibv.uio.no) - **NOT TO JON!**. Please put the course code and your name in the subject field (e.g. "Exam MBV-INF4410 Your Name").

The exam must be handed in as a single PDF document. The document should be marked with the date, course code and your name.

If necessary for evaluating the exam, a small oral examination may be arranged.

MBV-INF4410: Grade scale A-F (F = fail)

MBV-INF9410: Pass/fail (Pass = B or better)

Curriculum

- All lectures
- All exercises, demos and computer labs
- Obligatory assignments
- [Articles](#) listed on the wiki

General information

Course web page

https://wiki.uio.no/projects/clsi/index.php/MBV-INF410_2016

General information

- Everyone need to send me an email with this subject header (to jon.bråte@ibv.uio.no):

“Course version” “email address” “full name”

Like this:

MBV-INF9410 jon.bråte@ibv.uio.no Jon Bråte

NB!

I will send you the
obligs, exam and
important
information to this
address!

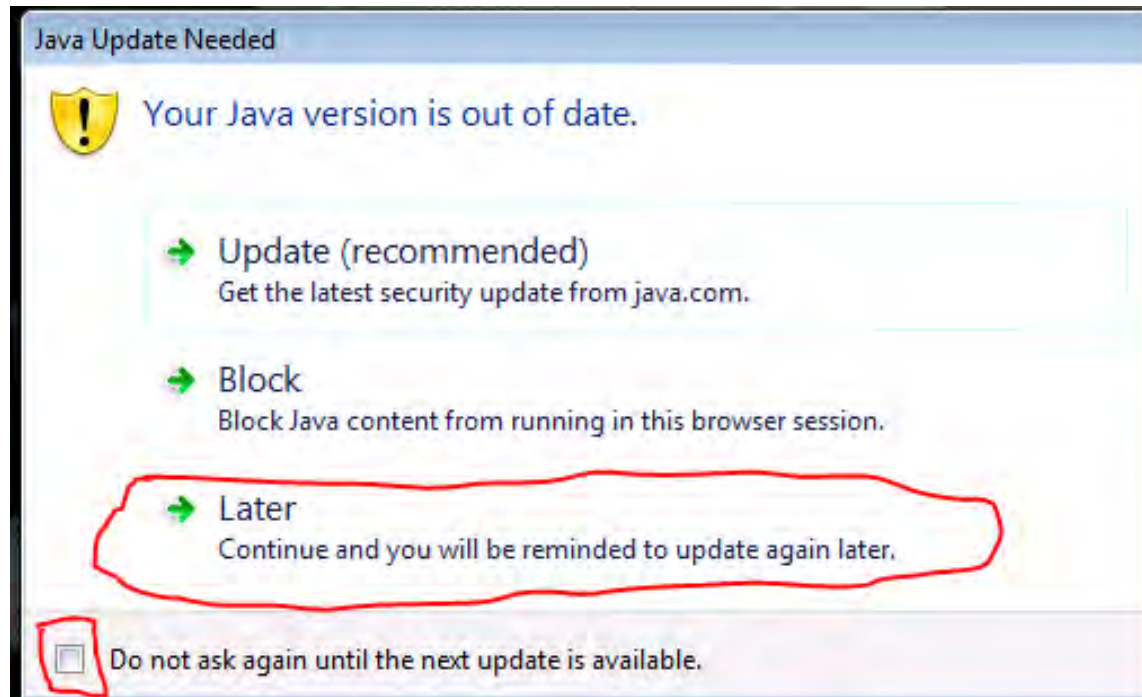
Laptops

I urge you to bring your own laptops

- We don't have access to computer rooms every day
- You need permissions to install software on it
- Please bring and **external mouse**
- You **must** have access to the UiO network (UiO username and password – see [here](#))
- All files should be stored on your UiO home directory – not locally on your laptop!



Don't update Java on the Desktops!!

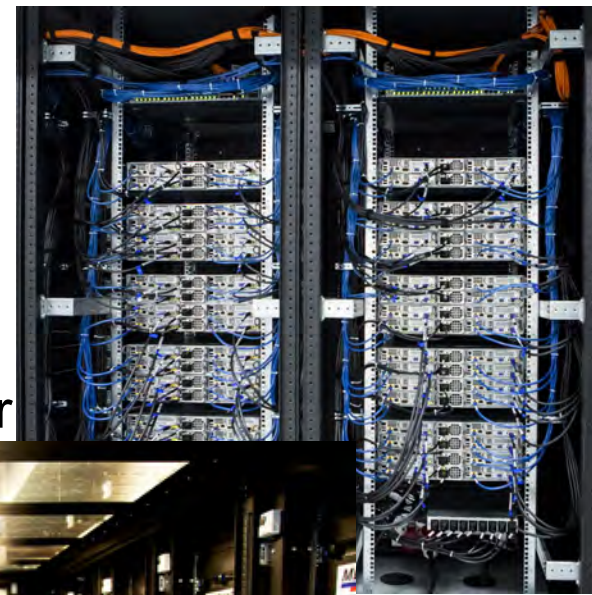


Computational resources at UiO

- Linux cluster
- Abel was (in June 2012) number 96 on the list of the most powerful computers in the world
- 258 TFLOP/s theoretical peak performance
- We will use Freebee – a "small corner" of Abel.

[Abel home page](#)

The Abel supercomputer



Defining transcribed regions using RNA-seq

Brian T Wilhelm^{1,4}, Samuel Marguerat^{2,4}, Ian Goodhead³ & Jürg Bähler²

¹Institute for Research in Immunology and Cancer (IRIC), Université de Montréal, Montréal, Québec, Canada. ²Department of Genetics, Evolution & Environment and UCL Cancer Institute, University College London, London, UK. ³Unit for Functional and Comparative Genomics, School of Biological Sciences, University of Liverpool, Liverpool, UK. ⁴These authors contributed equally to this work. Correspondence should be addressed to J.B. (j.bahler@ucl.ac.uk).

Published online 21 January 2010; doi:10.1038/nprot.2009.229

Next-generation sequencing technologies are revolutionizing genomics research. It is now possible to generate gigabase pairs of DNA sequence within a week without time-consuming cloning or massive infrastructure. This technology has recently been applied to the development of 'RNA-seq' techniques for sequencing cDNA from various organisms, with the goal of characterizing entire transcriptomes. These methods provide unprecedented resolution and depth of data, enabling simultaneous quantification of gene expression, discovery of novel transcripts and exons, and measurement of splicing efficiency. We present here a validated protocol for nonstrand-specific transcriptome sequencing via RNA-seq, describing the library preparation process and outlining the bioinformatic analysis procedure. While sample preparation and sequencing take a fairly short period of time (1–2 weeks), the downstream analysis is by far the most challenging and time-consuming aspect and can take weeks to months, depending on the experimental objectives.

Modern biology

Wet-lab: 1 week.

Dry-lab (analysing data): months...

ARTICLES

The sequence and *de novo* assembly of the giant panda genome

Travelled around in China
and took blood samples
from pandas

Wet lab?

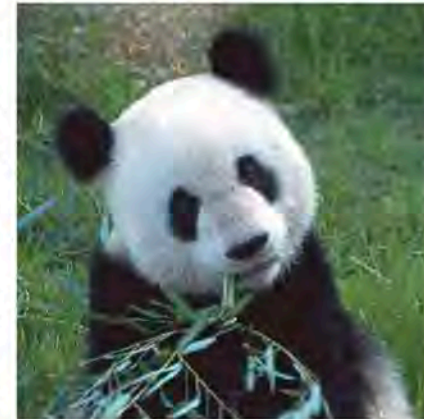
Ruiqiang Li^{1,2*}, Jun Wang¹, Bo Li¹, Yinqi Bai¹, Zhaolei Zhang¹¹, Rasmita Ghosh¹², Frederick Chi-Ching Leung¹⁵, Yan Zhou¹⁶, Jianjun Cao¹, Xiao Sun¹⁶, Yonggui Fu¹⁷, Xiaodong Fang¹, Xiaosen Guo¹, Bo Wang¹, Rong Hou⁸, Fujun Shen⁸, Bo Mu¹, Peixiang Ni¹, Runmao Lin¹, Wubin Qian¹, Guodong Wang^{3,6}, Chang Yu¹, Wenhui Nie⁶, Jinhuan Wang⁶, Zhigang Wu¹, Huiqing Liang¹, Jiumeng Min^{1,7}, Qi Wu⁹, Shifeng Cheng^{1,7}, Jue Ruan^{1,3}, Mingwei Chen¹, Wen¹, Binghang Liu¹, Xiaoli Ren¹, Huisong Zheng¹, Dong Dong¹¹, Kathrin Glatz¹, Yingrui Li¹, Tommy T. F. Wong¹, Timing Gong¹, Hongde Liu¹⁶, Dejin Zhang¹⁶, Yuanyuan Ren¹, Guojie Zhang^{1,3,6}, Michael Yang Zheng^{1,3}, Yongyong Shi⁵, Zhiqiang Li⁵, Feng Tian¹, Xiaoling Wang¹, Haiyin Wang¹, Siu-Ming Yiu²², Shiping Liu²³, Hemin Zhang¹, Junyi Wang¹, Nan Qin¹, Li Li¹, Jingxiang Li¹, Maynard Olson²⁶, Xiuqing Zhang¹, Songgar

Using next-generation sequencing technology a
giant panda genome

Mostly bioinformatics,
isn't it?

Our
using next-gen
genomes.

Author Contributions R.L., W.F., G.T., Ho.Z., L.H. and Jin.C. contributed equally to this work. Ju.W. and Ji.W. managed the project. Zhi.Z., R.H., F.S., He.Z., De.L., Ya.H., Jin.C., W.N., Jin.W. and W.W. prepared the panda DNA sample. X.Z., G.T., Jin.L., L.L., M.J., Da.L., Z.X., Jia.C., B.W., B.M., Z.W., Hu.L., X.R., Hu.Z., Si.L., Q.Z., Ju.Z., Y.R., Qin.L., Y.C., X.L. and Y.Z. performed sequencing. Ju.W., R.L. and W.F. designed analysis. Ho.Z., P.N., W.Q., G.S., S.Z., Run.L., F.T., J.R., M.Wa., Z.S., M.We., Xiao.W., H.W., L.X., T.-W.L. and S.-M.Y. performed genome assembly. Q.H., Q.C., Jia.L., J.M., Bi.L., Qib.L., Yu.H., Yang.Z., Ji.Z., W.G., X.X., Zu.L., X.S., Ho.L., D.Z. and Ni.Q. performed genome annotation. Ju.L., Bo.L., Y.B., Z.Y., S.C., Zha.Z., D.D., K.C., R.N., C.K., T.V., N.A., Sh.L., G.Z. and L.M. performed comparative genomics. Yap.Z., W., F.W., Q.W., M.W.B., L.H., Y.S., Zh.L., C.C.S., O.A.R., F.C.-C.L., T.T.-Y.L., Y.W., H., Y.F. and A.X. analysed genes related to panda-specific phenotypic characteristics. X.F., He.L., F.W., X.G., C.Yu., Hao.Z., Han.Z. and Y.L. identified heterozygous SNPs and performed panda historical population analysis. G.L., J.T., L.F., C.Ye. and T.G. performed data submission and database construction. Ju.W., Ji.W., R.L. and W.F. wrote the paper. X.W., G.Y., Y.G., Z.J., Juny.W., Na.Q., G.K.-S.W., L.B., M.O., K.K., So.L. and H.Y. revised the paper.



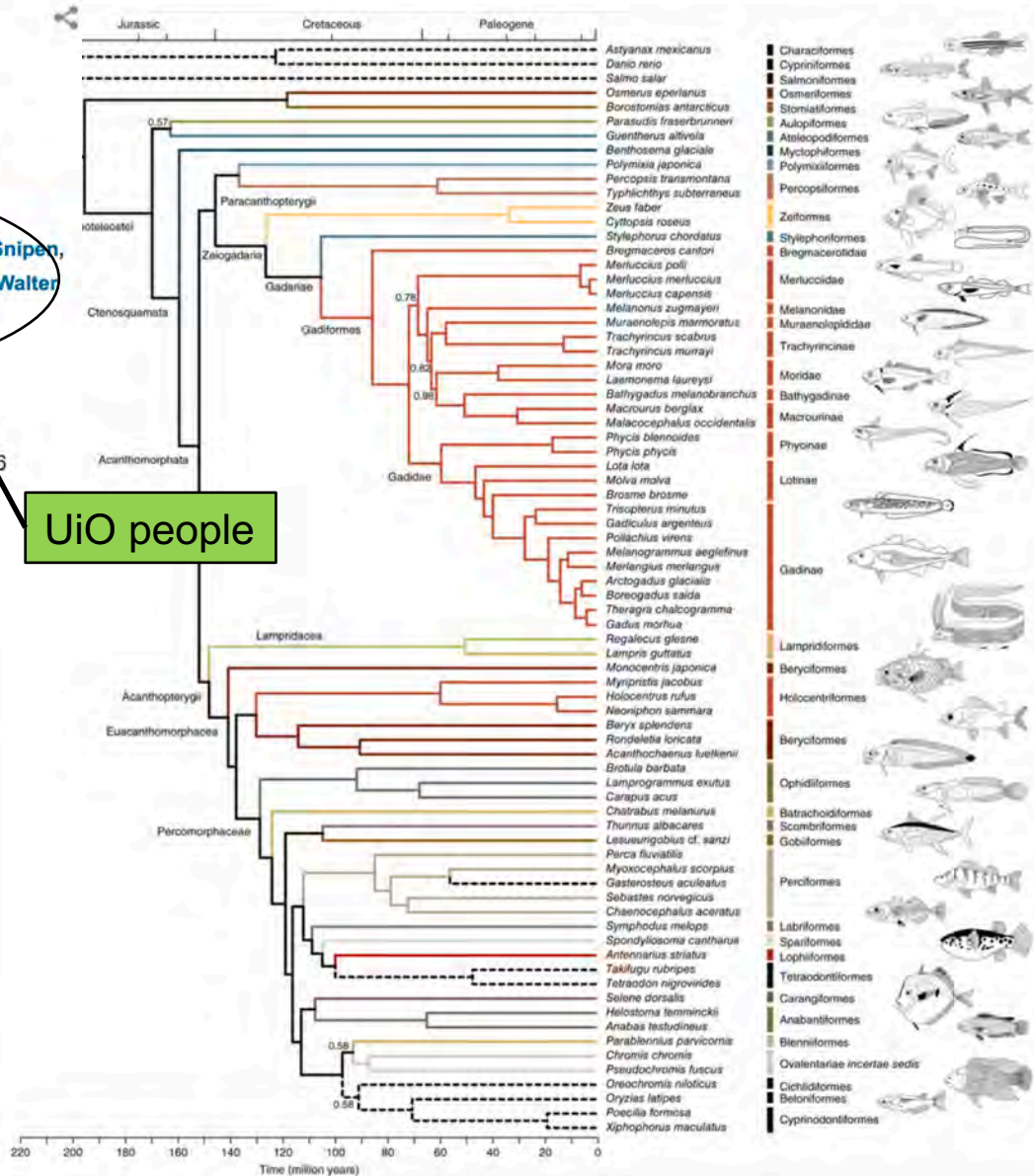
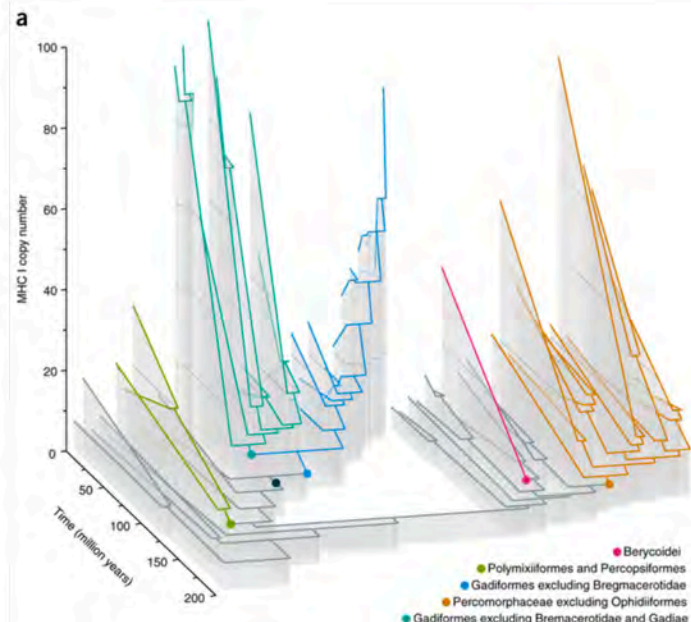
Evolution of the immune system influences speciation rates in teleost fishes

Martin Malmström, Michael Matschiner, Ole K Tørresen, Bastiaan Star, Lars G Snipen, Thomas F Hansen, Helle T Baalsrud, Alexander J Nederbragt, Reinhold Hanel, Walter Salzburger, Nils C Stenseth, Kjetill S Jakobsen & Sissel Jentoft

Affiliations | Contributions | Corresponding authors

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Tissues, sequencing and assembly.

Genomic DNA was obtained from various tissues of the different species in this study. Most tissue samples were provided by museums and other collections, while some come from commercially caught fish in collaboration with local fishermen (see [Supplementary Table 1](#) for a full list of tissues and contributors). A single paired-end library, with an insert size of ~400 bp, was created for each species, using the Illumina TruSeq Sample Prep v2 Low-Throughput protocol. All species were sequenced (2 × 150 bp) to >9× coverage on the Illumina HiSeq 2000 platform, and sequences were assembled using the Celera Assembler²⁶ ([Supplementary Note](#)). Draft genome assembly quality, in terms of gene space completeness, was assessed using CEGMA²⁷ and BUSCO²⁸ ([Supplementary Table 3](#) and [Supplementary Note](#)).

Gene mining of draft genome assemblies.

All draft genome assemblies were mined for genetic content on the unitig (UTG) assembly level, as assembly parameters are stricter for UTGs than for contigs or scaffolds. The presence or absence of each gene was determined through an automated pipeline, using full-length amino acid sequences for 27 immune-related genes and 3 control genes, from ten teleost genomes (Ensembl gene identifiers are listed in [Supplementary Table 6](#)). Potential genes were detected using TBLASTN with an acceptance level of $e\text{ value} = 1 \times 10^{-10}$ and eventual identification of ORFs predicted by the software Genescan⁵⁷. All ORFs were then compared to the UniProt database ([Supplementary Note](#)), and reciprocal BLASTN hits were recorded. Hits were potentially correct if their $e\text{ value}$ was below 1×10^{-10} . All recorded annotations for each gene were then manually inspected, and the best hit is reported (see the [Supplementary Note](#) for details and [Supplementary Table 7](#) for the location of each identified ortholog).

Copy number estimation of MHC I genes.

High sequence similarity and conserved regions make the different MHC I genes difficult to assemble correctly. To estimate the number of copies of these genes in each of the sequenced genomes, we applied a new method for copy number estimation, based on a comparison of raw read counts for target and reference sequences. For MHC I U- and Z-lineage genes, we used 270 bp of the conserved $\alpha 3$ domain as the target and equivalently sized fragments from 14 single- or low-copy genes as references (see [Supplementary Table 9](#) for a full overview of all reference regions). MHC I target sequences were prepared through consensus by majority for all hits detected in the individual draft genome assemblies with TBLASTN (e-value cutoff set to 1×10^{-5}) using U- and Z-lineage MHC I $\alpha 3$ -domain sequences from ten teleost reference genomes as queries. The number of copies of each of the target genes was determined on the basis of the number of unique sequencing reads mapping to this region, relative to the number of reads matching each of the reference gene regions. The copy numbers of each of the reference gene regions were estimated first, using an iterative method and four different BLAST stringencies. Not all reference regions fulfilled our criteria, and some references were discarded for some species (see the [Supplementary Note](#) for details and [Supplementary Table 11](#) for a full list of the references used for each species). Copy numbers for both MHC I lineages were then estimated by comparing the number of raw reads matching both the target and reference sequences and taking estimated genome size, coverage variation and total number of reads into account. The uncertainties of all copy number estimates were assessed with a double-bootstrapping procedure ([Supplementary Note](#)).

Phylogenetic inference

Strict filtering criteria were applied for the identification of orthologs. For the 33,737 annotated zebrafish genes in n1, we selected the longest transcript if it had at least five non-stop length. We removed genes that could not be assigned to a clade which teleost fishes did not form a monophyletic group. Genes for which the Ensembl gene tree indicated gene loss were included all ten teleost species of Ensembl v78 (Supplemental Table S1). Genomes of Ensembl were used to calculate TBLASTN using the BLAST+ v2.2.9 suite of tools⁴⁷. Exon-specific orthologs were defined on the basis of this bitcore information of the known orthologs had bitcores lower than this threshold. Genes with less than five remaining exons were discarded, which resulted in 302 zebrafish genes that were then used as queries in the new teleost draft genome assemblies, the 10 Ensembl genomes of salmon⁴⁸. For each species, the best hits to their TBLASTN bitcores were above the exon-specific threshold. Alignments of TBLASTN hits for the 2,251 exons were then used to identify nonsynonymous to synonymous substitutions (dN/dS) and

None of the authors are informaticians (all are biologists)...

on the number of congruent exon trees (Supplementary Note 2). For the analysis of the evolution of the *Utricularia* genus, we estimated the coefficient of variation of rate (0.25 v2.2.0 (ref. 52)) and removed the genes with the highest variation (Supplementary Note). After this step, our strictly filtered dataset contained 567 exons from 111 genes, with a total alignment length of 1,000,000 bp. To assess the consequences of strict filtering on phylogenetic inference, we compared maximum-likelihood phylogenies based on the strictly filtered dataset (302 genes, 252,442 bp, 18.2% missing data) with phylogenies based on a dataset of all genes (1,000,000 bp, 7.3% missing data) with phylogenies based on a dataset of all genes (1,000,000 bp, 7.3% missing data) strictly filtered (302 genes, 252,442 bp, 18.2% missing data). Both data sets were inferred with the software RAxML v8.2.12 (Stamatakis 2014) (Supplementary Note). The strictly filtered dataset was used to estimate divergence times with the software BEAST v2.2.2 (Bouckaert 2015) (Supplementary Note). Calibration was calculated with the BEAST add-on ClockLab (Bouckaert 2015) (Supplementary Note). Divergence rates and the fossil sampling rate. The divergence rates and the fossil sampling rate were estimated using the BEAST add-on ClockLab (Bouckaert 2015) (Supplementary Note). The divergence rates were identified and used to constrain the BEAST analysis. Calibration densities, taking into account the uncertainty in the divergence rates, were used to constrain the BEAST analysis (Supplementary Note). We further used coalescent-based methods to assess the potential for potentially misleading phylogenetic signal due to incomplete lineage sorting (ILS). We conducted both with individual gene trees and with trees inferred from the whole genome data to the binning approach of Mirarab *et al.* 55. Maximum-likelihood phylogenies were inferred using the maximum-likelihood approach of RAxML v8.2.12 (Stamatakis 2014) (Supplementary Note). The maximum-likelihood phylogenies resulting from BEAST v2.2.2 (Bouckaert 2015) (Supplementary Note) were used for species tree inference with the software *ASTRAL* v2.2.2 (Ogilvie 2015) (Supplementary Note). Fig. 2, Supplementary Table 5 and Supplementary Note 2.

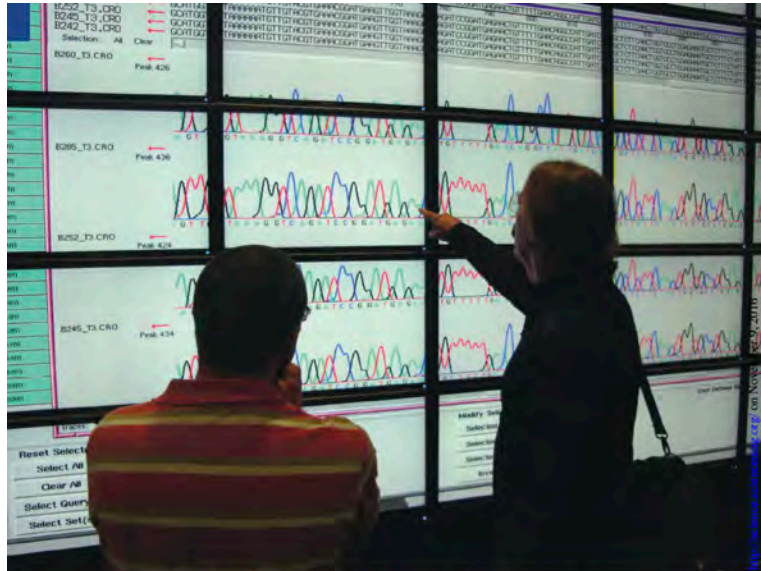
Rate shifts in MHC I copy number evolution.

Phylogenetic signal in MHC I copy number evolution was assessed with Blomberg's K statistic⁵⁸, calculated using the phylosignal function of the picante R package v.1.6-2 (ref. 59), and with Pagel's lambda⁶⁰, calculated with function phylosig of the phytools R package v.0.4-45 (ref. 61) (Supplementary Note). The fits of four general models of trait evolution were compared on the basis of their sample-size-corrected Akaike information criterion (AICc), using the function fitContinuous of the geiger R package v.2.0.3 (ref. 62): a white noise model, a Brownian motion model, an early-burst model⁶³ and a single-peak Ornstein–Uhlenbeck model^{40, 64} (Supplementary Note). The reversible-jump Bayesian approach of the bayou R package v.1.0.1 (ref. 65) was used to perform MCMC sampling of locations, magnitudes and numbers of shifts in multiple-optima Ornstein–Uhlenbeck models (Supplementary Fig. 5 and Supplementary Note). On the basis of the results of the bayou analysis, explicit hypotheses for shift combinations were tested in a likelihood framework, using the SLOUCH R package^{41, 42}. For each shift combination, the likelihood of the best fitting combination of optimum, half-life and stationary variance was compared on the basis of model comparison based on AICc scores (Supplementary Table 13 and Supplementary Note). The ancestral states of log-transformed MHC I copy numbers were estimated at internal nodes of the time-calibrated phylogeny, on the basis of the best fitting model (Supplementary Note).

Diversification rate analyses

Patterns of species diversification were analyzed with the Bayesian framework implemented in BAMM v.2.2.0 (ref. 66), on the basis of the time-calibrated phylogeny and counts of species richness in each of the 37 mutually exclusive clades of teleost fishes (Supplementary Table 14). The 'MEDUSA-like' model of diversification, assuming constant speciation and extinction rates within specific shift regimes⁶⁷, was used for this analysis (Supplementary Fig. 8 and Supplementary Note). To test whether high MHC I copy numbers are associated with lineages that have high diversification rates, we carried out BiSSE analyses⁶⁸ with the diversitree R package⁶⁹. In these analyses, species were grouped into two categories for high and low MHC I copy numbers, on the basis of a given threshold value. Analyses were repeated for 26 equally spaced copy number threshold values between 10 and 60. As diversitree allows terminal clades with extant diversities of no more than 200 species, we used birth–death models of diversification in combination with the diversified sampling scheme of Höhna *et al.*⁷⁰ to stochastically resolve subclades of all clades with more than 200 extant species, which was repeated 25 times. BiSSE analyses were conducted for each of the 25 resulting phylogenies and with each of the 26 copy number thresholds, assuming symmetric transition rates between high and low copy numbers and identical extinction rates in taxa with high and low copy numbers (Supplementary Note and Supplementary Data).

No more wet lab biology?



Biology's Dry Future

The explosion of publicly available databases housing sequences, structures, and images allows life scientists to make fundamental discoveries without ever getting their hands "wet" at the lab bench

Most life scientists single-mindedly focus their careers on a particular organism or disease—even just a specific molecular pathway. After all, it can often take months of training to master growing a particular cell type or learn a new laboratory technique. Atul Butte, however, wanders from topic to topic—and reaps scientific successes along the way. Though only 44 years old, he has earned tenure at Stanford University's School of Medicine in Palo Alto, California, based on advances in diabetes, obesity, transplant rejection, and the discovery of new drugs for lung cancer and other diseases.

Butte's lab is different, too. It isn't crowded with cell cultures and reagents. His tools look like those of an engineer or software developer: Most often, he's simply working on a Sony laptop, although at times he does turn to a large computer cluster at Stanford and supercomputers elsewhere when in need of massive processing power. Instead of growing cells and sequencing DNA, Butte, his students, and postdocs sift through massive databases full of freely available information, such as human genome sequences, cancer genome readouts, brain imaging scans, and biomarkers for specific diseases such as diabetes and Alzheimer's.

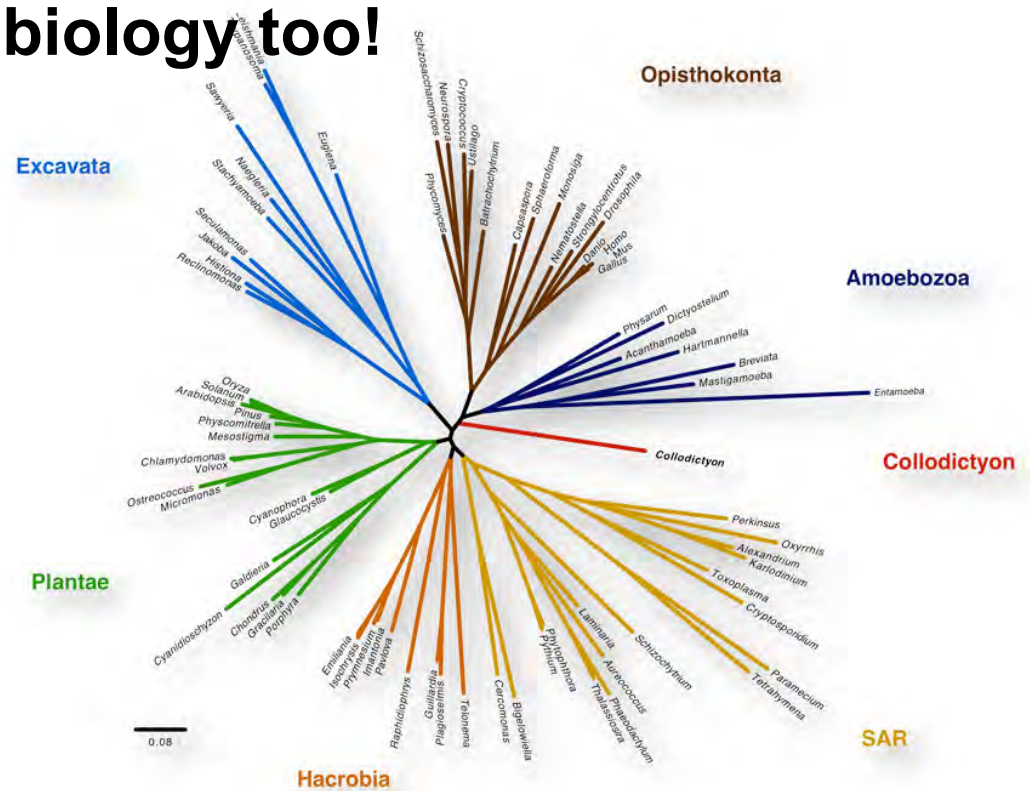
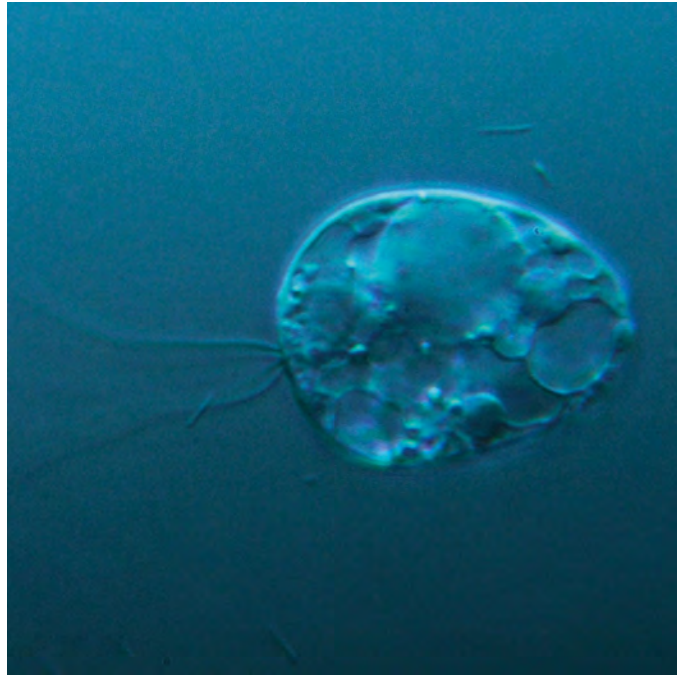
Many call this type of research "dry lab biology," to contrast it with the more hands-on "wet" traditional style of research. Although statistics on the number of dry lab biologists are hard to come by, these data hunters believe they are a growing minority. Butte is one of its top practitioners. Using publicly available data, for example, 2 years ago Butte and his colleagues surveyed the activity of large sets of genes in people affected by 100 different diseases and in cultured human cells exposed to 164 drugs already on the market. By comparing patterns of genes flipped on or off by the diseases and by the drugs, the team drew unexpected connections. They found clues



**"I'm like a kid in a candy store.
There is so much we can do."**

—Atul Butte, Stanford University School of Medicine

We need traditional biology too!



Collodictyon – collected in Ås in the 1980's. Placed on the Tree of Life in 2012.

COMMENT

Open Access



Gene name errors are widespread in the scientific literature

Mark Ziemann¹, Yotam Eren^{1,2} and Assam El-Osta^{1,3*}

Abstract

The spreadsheet software Microsoft Excel, when used with default settings, is known to convert gene names to dates and floating-point numbers. A programmatic scan of leading genomics journals reveals that approximately one-fifth of papers with supplementary Excel gene lists contain erroneous gene name conversions.

Keywords: Microsoft Excel, Gene symbol, Supplementary data

Abbreviations: GEO, Gene Expression Omnibus; JIF, journal impact factor

frequently reused. Our aim here is to raise awareness of the problem.

We downloaded and screened supplementary files from 18 journals published between 2005 and 2015 using a suite of shell scripts. Excel files (.xls and .xlsx suffixes) were converted to tabular separated files (tsv) with *ssconvert* (v1.12.9). Each sheet within the Excel file was converted to a separate tsv file. Each column of data in the tsv file was screened for the presence of gene symbols. If the first 20 rows of a column contained five or more gene symbols, then it was suspected to be a list of gene symbols, and then a regular expression (regex) search of the entire column was applied to identify gene symbol errors. Official gene symbols from Ensembl version 82, accessed November 2015, were obtained for

TextWrangler File Edit Text View Search Go Window #! \$ Help

genes.fpkm_tracking

~/Desktop/genes.fpkm_tracking

	tracking_id	class_code	nearest_ref_id	gene_id	gene_short_name	tss_id	locus	length	coverage	FPKM	FPKM_conf_lo	FPKM_conf_hi	FPKM_status
1	CUFF.1	-	-	CUFF.1	-	-	ML0001:1114-1859	-	-	3.40091	1.78902	4.55977	OK
2	CUFF.2	-	-	CUFF.2	-	-	ML0001:3149-5023	-	-	3.17083	2.01129	4.41652	OK
3	CUFF.3	-	-	CUFF.3	-	-	ML0001:5201-6008	-	-	3.4317	1.78902	4.94611	OK
4	CUFF.4	-	-	CUFF.4	-	-	ML0001:10567-17012	-	-	95.1855	66.665	83.1454	OK
5	CUFF.5	-	-	CUFF.5	-	-	ML0001:16842-25200	-	-	8.3306	6.73692	9.99184	OK
6	ML00014a	-	-	ML00014a	-	-	ML0001:18182-18727	-	-	0	0	0	OK
7	ML00015a	-	-	ML00015a	-	-	ML0001:20704-21437	-	-	0.310551	0	0.579303	OK
8	CUFF.6	-	-	CUFF.6	-	-	ML0001:116596-117203	-	-	84.1429	30.5411	48.7587	OK
9	CUFF.7	-	-	CUFF.7	-	-	ML0001:117529-120080	-	-	27.7801	24.287	31.2763	OK
10	CUFF.8	-	-	CUFF.8	-	-	ML0001:96723-100218	-	-	19.537	17.3757	21.7	OK
11	CUFF.9	-	-	CUFF.9	-	-	ML0001:95229-96857	-	-	0	0	2.6981	OK
12	ML000120a	-	-	ML000120a	-	-	ML0001:170230-171641	-	-	26.3709	18.6629	26.7691	OK
13	ML000117a	-	-	ML000117a	-	-	ML0001:160376-163637	-	-	17.0622	14.1038	18.4302	OK
14	ML000118a	-	-	ML000118a	-	-	ML0001:164238-166002	-	-	7.58874	5.87355	8.85847	OK
15	CUFF.12	-	-	CUFF.12	-	-	ML0001:158656-160173	-	-	99.6587	88.9378	110.403	OK
16	ML000121a	-	-	ML000121a	-	-	ML0001:171781-174164	-	-	18.474	14.6806	19.5556	OK
17	ML000122a	-	-	ML000122a	-	-	ML0001:174306-176040	-	-	17.8856	14.4971	18.8561	OK
18	CUFF.16	-	-	CUFF.16	-	-	ML0001:166667-169953	-	-	12.5275	11.1472	13.9171	OK
19	CUFF.10	-	-	CUFF.10	-	-	ML0001:124631-147621	-	-	27.2018	24.4382	30.0801	OK
20	CUFF.17	-	-	CUFF.17	-	-	ML0002:1471-1804	-	-	8.86287	2.29529	7.90601	OK
21	CUFF.18	-	-	CUFF.18	-	-	ML0002:53-1326	-	-	12.3905	7.24842	13.3478	OK
22	CUFF.11	-	-	CUFF.11	-	-	ML0001:101488-113483	-	-	0	0	2.68119	OK
23	CUFF.13	-	-	CUFF.13	-	-	ML0001:113060-114810	-	-	101.412	93.2058	109.982	OK
24	CUFF.15	-	-	CUFF.15	-	-	ML0001:104169-106175	-	-	2.00656	1.35475	2.62483	OK
25	ML000111a	-	-	ML000111a	-	-	ML0001:108031-108445	-	-	1.48892	0.205135	2.66675	OK
26	ML00022a	-	-	ML00022a	-	-	ML0002:9048-14586	-	-	4.86618	3.92579	5.766	OK
27	CUFF.21	-	-	CUFF.21	-	-	ML0002:7547-8333	-	-	21.9415	18.261	25.7619	OK
28	ML00023a	-	-	ML00023a	-	-	ML0002:14791-16731	-	-	32.5216	26.3642	33.5128	OK
29	CUFF.22	-	-	CUFF.22	-	-	ML0001:176297-182947	-	-	89.2902	82.4019	96.1787	OK
30	ML00037a	-	-	ML00037a	-	-	ML0003:39752-40184	-	-	0	0	0	OK
31	CUFF.23	-	-	CUFF.23	-	-	ML0003:72556-73075	-	-	6.88679	3.10904	7.69078	OK
32	ML000312a	-	-	ML000312a	-	-	ML0003:73476-77906	-	-	8.86058	7.14782	10.4592	OK
33	CUFF.26	-	-	CUFF.26	-	-	ML0003:29927-34072	-	-	40.5148	34.2588	41.8107	OK
34	CUFF.27	-	-	CUFF.27	-	-	ML0003:34225-34603	-	-	4.85363	1.34803	5.16744	OK
35	ML00036a	-	-	ML00036a	-	-	ML0003:35217-38953	-	-	5.00336	4.17024	5.81381	OK
36	CUFF.19	-	-	CUFF.19	-	-	ML0001:247518-252750	-	-	8.36266	7.10624	9.63082	OK
37	CUFF.20	-	-	CUFF.20	-	-	ML0001:252902-253837	-	-	2.26791	1.18079	3.26987	OK
38	CUFF.24	-	-	CUFF.24	-	-	ML0001:258628-285705	-	-	7.80224	5.57303	10.0045	OK
39	CUFF.25	-	-	CUFF.25	-	-	ML0001:282805-284718	-	-	3.07781	2.25518	3.89792	OK
40	CUFF.28	-	-	CUFF.28	-	-	ML0001:288715-290655	-	-	52.7171	48.8135	56.6853	OK
41	CUFF.29	-	-	CUFF.29	-	-	ML0001:285933-293851	-	-	11.4689	6.82331	12.3063	OK
42	CUFF.30	-	-	CUFF.30	-	-	ML0003:111834-117047	-	-	1.66999	0.933269	2.41457	OK
43	ML000313a	-	-	ML000313a	-	-	ML0003:79499-97515	-	-	326.493	233.132	263.175	OK
44	CUFF.37	-	-	CUFF.37	-	-	ML0003:119800-144640	-	-	26.8318	24.509	29.1634	OK
45	CUFF.41	-	-	CUFF.41	-	-	ML0004:10180-10734	-	-	2.30811	0.766478	3.6791	OK
46	CUFF.34	-	-	CUFF.34	-	-	ML0003:42892-72287	-	-	124.587	106.319	122.351	OK
47	ML00039a	-	-	ML00039a	-	-	ML0003:46810-47149	-	-	25.3156	8.76815	18.7889	OK

iTerm2 Shell Edit View Profiles Toolbelt Window Help

1.ssh

	tracking_id	class_code	nearest_ref_id	gene_id	gene_short_name	tss_id	locus	length	coverage	FPKM	FPKM_conf_lo	FPKM_conf_hi	FPKM_status
1	CUFF.1	-	-	CUFF.1	-	-	ML0001:1114-1859	-	-	3.40091	1.78902	4.55977	OK
2	CUFF.2	-	-	CUFF.2	-	-	ML0001:3149-5023	-	-	3.17083	2.01129	4.41652	OK
3	CUFF.3	-	-	CUFF.3	-	-	ML0001:5201-6008	-	-	3.4317	1.78902	4.94611	OK
4	CUFF.4	-	-	CUFF.4	-	-	ML0001:10567-17012	-	-	95.1855	66.665	83.1454	OK
5	CUFF.5	-	-	CUFF.5	-	-	ML0001:16842-25200	-	-	8.3306	6.73692	9.99184	OK
6	ML00014a	-	-	ML00014a	-	-	ML0001:18182-18727	-	-	0	0	0	OK
7	ML00015a	-	-	ML00015a	-	-	ML0001:20704-21437	-	-	0.310551	0	0.579303	OK
8	CUFF.6	-	-	CUFF.6	-	-	ML0001:116596-117203	-	-	84.1429	30.5411	48.7587	OK
9	CUFF.7	-	-	CUFF.7	-	-	ML0001:117529-120080	-	-	27.7801	24.287	31.2763	OK
10	CUFF.8	-	-	CUFF.8	-	-	ML0001:96723-100218	-	-	19.537	17.3757	21.7	OK
11	CUFF.9	-	-	CUFF.9	-	-	ML0001:95229-96857	-	-	0	0	2.6981	OK
12	ML000120a	-	-	ML000120a	-	-	ML0001:170230-171641	-	-	26.3709	18.6629	26.7691	OK
13	ML000117a	-	-	ML000117a	-	-	ML0001:160376-163637	-	-	17.0622	14.1038	18.4302	OK
14	ML000118a	-	-	ML000118a	-	-	ML0001:164238-166002	-	-	7.58874	5.87355	8.85847	OK
15	CUFF.12	-	-	CUFF.12	-	-	ML0001:158656-160173	-	-	99.6587	88.9378	110.403	OK
16	ML000121a	-	-	ML000121a	-	-	ML0001:171781-174164	-	-	18.474	14.6806	19.5556	OK
17	ML000122a	-	-	ML000122a	-	-	ML0001:174306-176040	-	-	17.8856	14.4971	18.8561	OK
18	CUFF.16	-	-	CUFF.16	-	-	ML0001:166667-169953	-	-	12.5275	11.1472	13.9171	OK
19	CUFF.17	-	-	CUFF.17	-	-	ML0002:1471-1804	-	-	8.86287	2.29529	7.90601	OK
20	CUFF.18	-	-	CUFF.18	-	-	ML0002:53-1326	-	-	12.3905	7.24842	13.3478	OK
21	CUFF.11	-	-	CUFF.11	-	-	ML0001:101488-113483	-	-	0	0	2.68119	OK
22	CUFF.13	-	-	CUFF.13	-	-	ML0001:113060-114810	-	-	101.412	93.2058	109.982	OK
23	CUFF.15	-	-	CUFF.15	-	-	ML0001:104169-106175	-	-	2.00656	1.35475	2.62483	OK
24	ML000111a	-	-	ML000111a	-	-	ML0001:108031-108445	-	-	1.48892	0.205135	2.66675	OK
25	ML00022a	-	-	ML00022a	-	-	ML0002:9048-14586	-	-	4.86618	3.92579	5.766	OK
26	CUFF.21	-	-	CUFF.21	-	-	ML0002:7547-8333	-	-	21.9415	18.261	25.7619	OK
27	ML00023a	-	-	ML00023a	-	-	ML0002:14791-16731	-	-	32.5216	26.3642	33.5128	OK
28	CUFF.22	-	-	CUFF.22	-	-	ML0001:176297-182947	-	-	89.2902	82.4019	96.1787	OK
29	ML00037a	-	-	ML00037a	-	-	ML0003:39752-40184	-	-	0	0	0	OK
30	CUFF.23	-	-	CUFF.23	-	-	ML0003:72556-73075	-	-	6.88679	3.10904	7.69078	OK
31	ML000312a	-	-	ML000312a	-	-	ML0003:73476-77906	-	-	8.86058	7.14782	10.4592	OK
32	CUFF.26	-	-	CUFF.26	-	-	ML0003:29927-34072	-	-	40.5148	34.2588	41.8107	OK
33	CUFF.27	-	-	CUFF.27	-	-	ML0003:34225-34603	-	-	4.85363	1.34803	5.16744	OK
34	ML00036a	-	-	ML00036a	-	-	ML0003:35217-38953	-	-	5.00336	4.17024	5.81381	OK
35	CUFF.19	-	-	CUFF.19	-	-	ML0001:247518-252750	-	-	8.36266	7.10624	9.63082	OK
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42	ML000313a	-	-	ML000313a	-	-	ML0003:79499-97515	-	-	326.493	233.132	263.175	OK
43	CUFF.37	-	-	CUFF.37	-	-	ML0003:119800-144640	-	-	26.8318	24.509	29.1634	OK
44	CUFF.41	-	-	CUFF.41	-	-	ML0004:10180-10734	-	-	2.30811	0.766478	3.6791	OK
45	CUFF.34	-	-	CUFF.34	-	-	ML0003:42892-72287	-	-	124.587	106.319	122.351	OK

File naming

NO

myabstract.docx

Joe's Filenames Use Spaces and Punctuation.xlsx

figure 1.png

fig 2.png

JW7d^(2sl@deletethisandyourcareerisoverWx2*.txt

YES

2014-06-08_abstract-for-sla.docx

joes-filenames-are-getting-better.xlsx

fig01_scatterplot-talk-length-vs-interest.png

fig02_histogram-talk-attendance.png

1986-01-28_raw-data-from-challenger-o-rings.txt

File naming

Three principles for filenames

- Machine readable (no spaces, consistency in naming, prefix, suffix, punctuation)
- Human readable (name contains info on content)
- Plays well with default ordering (numbers first, ISO 8601 standard for dates, left pad with zeroes)

File naming

Excerpt of complete file listing:

```
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_H01.csv
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_H02.csv
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_H03.csv
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_platefile.csv
2014-02-26_BRAFWTNEGASSAY_FFPEDNA-CRC-1-41_A01.csv
2014-02-26_BRAFWTNEGASSAY_FFPEDNA-CRC-1-41_A02.csv
2014-02-26_BRAFWTNEGASSAY_FFPEDNA-CRC-1-41_A03.csv
2014-02-26_BRAFWTNEGASSAY_FFPEDNA-CRC-1-41_A04.csv
```

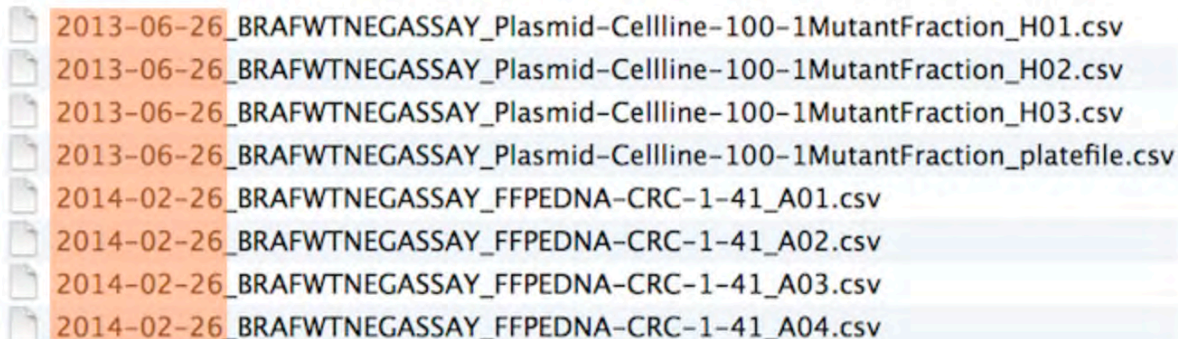
Example of globbing to narrow file listing:

```
Jennifers-MacBook-Pro-3:2014-03-21 jenny$ ls *Plasmid*
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_A01.csv
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_A02.csv
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_A03.csv
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_B01.csv
....
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_H03.csv
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_platefile.csv
```

File naming

For sorting chronologically:

Use the ISO 8601 standard for dates: YYYY-MM-DD




2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_H01.csv
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_H02.csv
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_H03.csv
2013-06-26_BRAFWTNEGASSAY_Plasmid-Cellline-100-1MutantFraction_platefile.csv
2014-02-26_BRAFWTNEGASSAY_FFPEDNA-CRC-1-41_A01.csv
2014-02-26_BRAFWTNEGASSAY_FFPEDNA-CRC-1-41_A02.csv
2014-02-26_BRAFWTNEGASSAY_FFPEDNA-CRC-1-41_A03.csv
2014-02-26_BRAFWTNEGASSAY_FFPEDNA-CRC-1-41_A04.csv

File naming

For sorting logically:

```
bio89093:delete jonbra$ ls -l
total 0
-rw----- 1 jonbra 17400 0 Nov 9 10:46 1-fileOne.txt
-rw----- 1 jonbra 17400 0 Nov 9 10:46 10-fileTen.txt
-rw----- 1 jonbra 17400 0 Nov 9 10:46 2-fileTwo.txt
-rw----- 1 jonbra 17400 0 Nov 9 10:46 3-fileThree.txt
```

Left pad with zeroes



```
bio89093:delete jonbra$ ls -l
total 0
-rw----- 1 jonbra 17400 0 Nov 9 10:46 01-fileOne.txt
-rw----- 1 jonbra 17400 0 Nov 9 10:46 02-fileTwo.txt
-rw----- 1 jonbra 17400 0 Nov 9 10:46 03-fileThree.txt
-rw----- 1 jonbra 17400 0 Nov 9 10:46 10-fileTen.txt
```

Open science (Lex snakker ikke så mye om dette)

Passer kanskje ikke her, men vil gjerne få inn noe om dette.
Kanskje senere når de har lært litt mer?

Jeg vil gjerne lære de litt "good practice" og muligheter for open science.

- Markdown?
- Raw data as read only
- Tools and programs to use (text editors and not word etc.)
- Filnavn?
- GitHub?

Quote from Wikipedia

Budskapet her er at Bioinformatikk ikke er én ting, og at det på mange måter er alt. (mao et ubrukelig begrep).