Methods and Techniques in Bioinfomatics

(From DNA bases to image databases)

Stephen Taylor

MRC WIMM Centre of Computational Biology

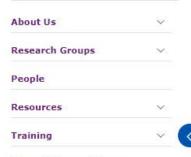


Weatherall Institute of Molecular Medicine

MRC WIMM Centre for Computational Biology

Using computational biology to help understand complex biological systems and combat diseases, from blood disorders to cancer and diabetes.

IN THIS SECTION



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Machine learning meets microscopy -Subjectivity dethroned

How can we remove subjectivity from science, whilst keeping the human in the loop?

Read more

. . . .



LATEST PUBLICATIONS

A revised model for promoter competition based on multi-way chromatin interactions at the a-globin locus.

Oudelaar AM. et al, (2019), Nat Commun, 10

A Spontaneous Ring-Opening Reaction Leads to a Repair-Resistant Thymine Oxidation Product in Genomic DNA.

Sahakyan AB. et al, (2020), Chembiochem, 21, 320 - 323

Haplotype matching in large cohorts using the Li and Stephens model.

NEWS



Role-playing computer game helps players understand how vaccines work on a global scale

8 October 2020

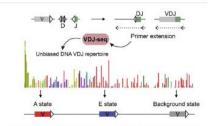


Marieke Oudelaar awarded prestigious Lise Meitner Excellence Program grant

6 October 2020

Research Groups

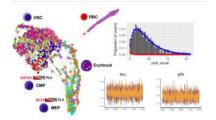




Hughes Group

Iotchkova Group: Statistical Genetics

Koohy Group: Machine Learning and Integrative Approaches in Immunology



Morrissey Group: Quantitative biology of cell fate and tissue dynamics



Sahakyan Group: Integrative Computational Biology and Machine Learning



Sims Group: Computational Genomics

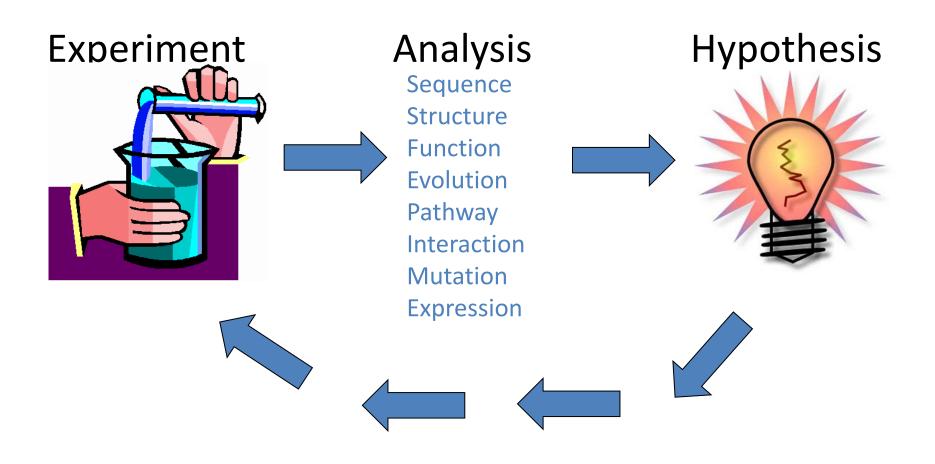


Taylor Group: Analysis, Visualisation and Informatics

Computational Biology and Bioinformatics is all about data...

- Definition
 - Bioinformatics is the computational analysis and storage of biological data
- Derivation
 - informatique French for 'data processing'
- Goal
 - To discover new biological insights using computers and biology

What is bioinformatics?



Why use bioinformatics?

Find an answer quickly

Most in silico biology is faster than in vitro

Massive amounts of data to analyse

Need to make use of all information Not possible to do analysis by hand Can't organise and store information only u

Can't organise and store information only using lab note books

Automation is key

However!

All results of computer analysis should to be verified by biologists

MRC Weatherall Institute for Molecular Medicine

Bioinformatics databases



Public databases are the most important entity in bioinformatics



Store knowledge about Sequence e.g. EMBL/Genbank HTS Experiments e.g. GEO Structure e.g. PDB Pathways e.g. KEGG, Metacore Diseases e.g. OMIM Genome Architecture: e.g. UCSC



Can be searched in a variety of ways

e.g. keyword, sequence, pattern

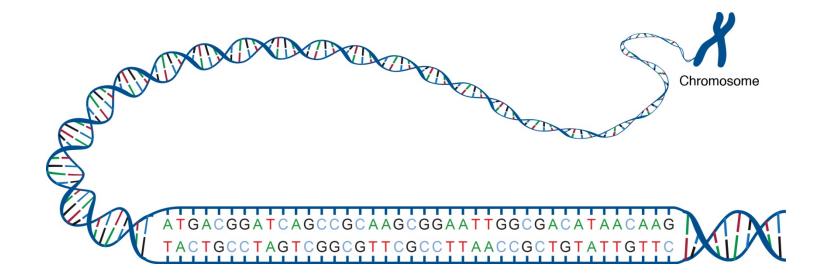
Keyword

| Search NCBI d | latabases | | | | He |
|----------------------|-------------|---|-------------------|-----------|--|
| | p | 3 | | | Search |
| About 1,673,010 | search resu | ilts for "p53" | | | |
| Literature | | | Genes | | |
| Books | 1.279 | books and reports | EST | 796 | expressed sequence tag sequences |
| MeSH | 158 | ontology used for PubMed indexing | Gene | 7,879 | collected information about gene loci |
| NLM Catalog | 108 | books, journals and more in the NLM Collections | GEO Data Sets | 8,899 | functional genomics studies |
| PubMed | 71,937 | scientific & medical abstracts/citations | GEO Profiles | 1,403,459 | gene expression and molecular abundance profiles |
| PubMed Central | 93,434 | full-text journal articles | HomoloGene | 38 | homologous gene sets for selected organisms |
| Health | | | PopSet | 94 | sequence sets from phylogenetic and population studies |
| ClinVar | 225 | human variations of clinical significance | UniGene | 414 | clusters of expressed transcripts |
| dbGaP | 22 | genotype/phenotype interaction studies | 5.43 | | |
| GTR | 110 | genetic testing registry | Proteins | | |
| MedGen | 72 | medical genetics literature and links | Conserved Domains | 120 | conserved protein domains |
| ОМІМ | 583 | online mendelian inheritance in man | Protein | 29,695 | protein sequences |
| PubMed Health | 71 | clinical effectiveness, disease and drug reports | Protein Clusters | 15 | sequence similarity-based protein clusters |
| Genomes | | | Structure | 1,082 | experimentally-determined biomolecular structures |
| | | | - Chemicals | | |
| Assembly | 1 | genomic assembly information | | | molecular pathways with links to genes, proteins and |
| BioProject | 642 | biological projects providing data to NCBI | BioSystems | 3,799 | chemicals |
| BioSample Clone | 307 0 | descriptions of biological source materials | PubChem BioAssay | 10,848 | bioactivity screening studies |
| dbVar | 1.464 | genomic and cDNA clones | PubChem | 8 | chemical information with structures, information and |
| abvar Epigenomics | 1,464 | genome structural variation studies epigenomic studies and display tools | Compound | | links |
| Genome | 5 | genome sequencing projects by organism | PubChem Substance | ə 650 | deposited substance and chemical information |
| GSS | 36 | genome sequencing projects by organism genome survey sequences | | | |
| Nucleotide | 24,181 | DNA and RNA sequences | | | |
| Probe | 3.507 | sequence-based probes and primers | | | |
| SNP | 6.592 | sequence-based probes and primers short genetic variations | | | |
| SRA | 440 | high-throughput DNA and RNA sequence read archive | | | |
| Тахопоту | 440 | taxonomic classification and nomenclature catalog | | | |

Bioinformatics Tools

- Hundreds of computer programs
- Many freely available
- Generally available on UNIX or LINUX
- Often interact with bioinformatics databases
- Many accessible via the WWW
- Some require very powerful computers to run on
- CCB provide a environment to do this

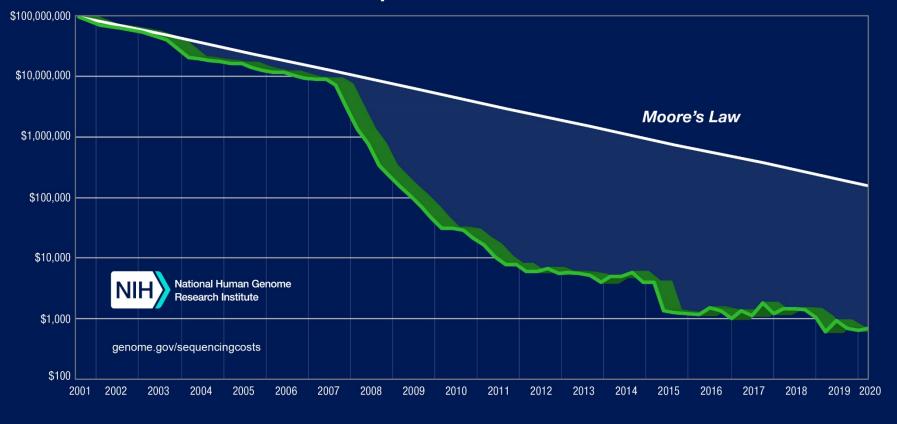
DNA



The Human Genome Project (1990-2003)

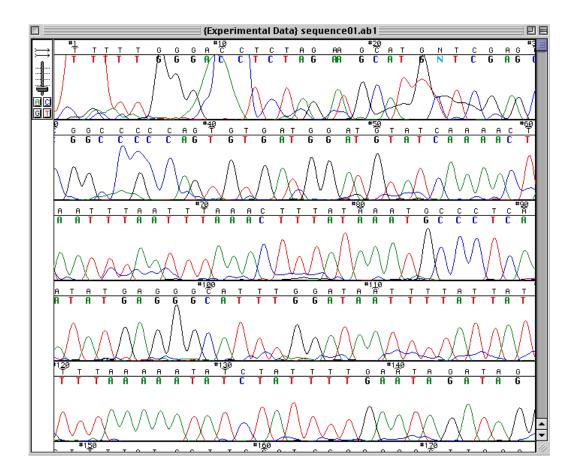
- Cost \$3 billion
- Could not have been achieved without bioinformatics
- Goals
 - *identify* all the 20,500 genes in human DNA,
 - *determine* the sequences of the 3 billion chemical base pairs that make up human DNA
 - store this information in databases
 - *improve* tools for data analysis
 - transfer related technologies to the private sector, and
 - address the ethical, legal, and social issues (ELSI) that may arise from the project.
- Need to bring together and store vast amounts of information from
 - Lab equipment and experiments
 - Computer Analysis
 - Human Analysis
 - Make visible to the world's scientists

Cost per Human Genome



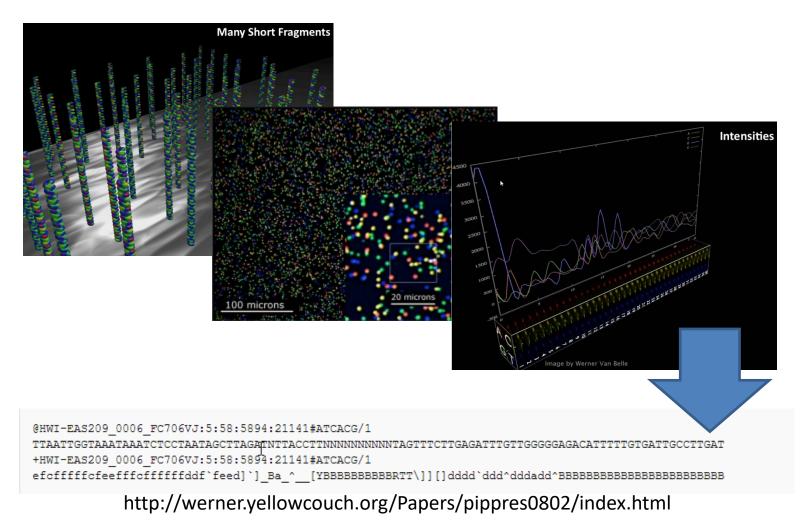
https://www.genome.gov/about-genomics/fact-sheets/Sequencing-Human-Genome-cost

Sanger Sequencing



Read length typically 500-600bp (up to 800bp)

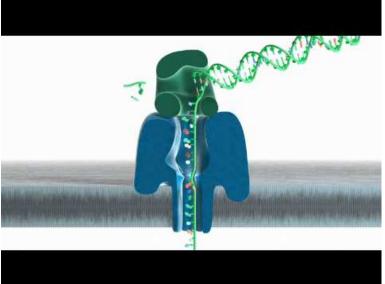
Next Generation Sequencing



Paired end reads, 50-300bp

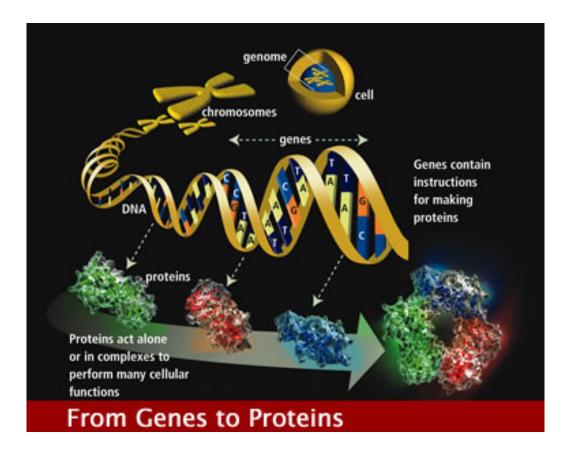
Nanopore sequencing





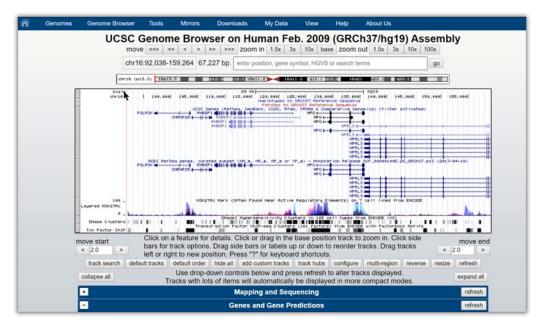
2Mb Read Lengths, genome assembly, direct RNA sequencing...

Organising Information



(http://www.ornl.gov/sci/techresources/Human_Genome/home.shtml)

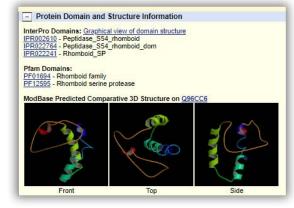
UCSC Genome Browser (http://genome.ucsc.edu/)

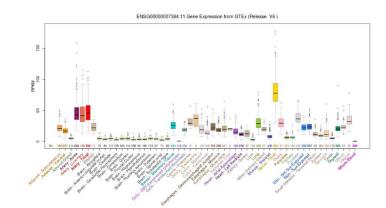


- Comments and Description Text from UniProtKB

ID: EHOP 1. HUMAN DESCRIPTION: Rechame. Full-PUNCTION: Rhomboid proteste-SUBUNIT: Homodime. or homoo SUBCELLULAR LOCATION: En TUSSUE SUFCEPEUTV Index. or or 1. ARName. Full-p1004Bpc, praking pathway and may thereby regulate sleep, cell survival, proliferation and migration sections, IERD). OF Interacts with EOF rugh the

PTM: N-plycosylated SIMILARITY: Belongs to the peptidase SS4

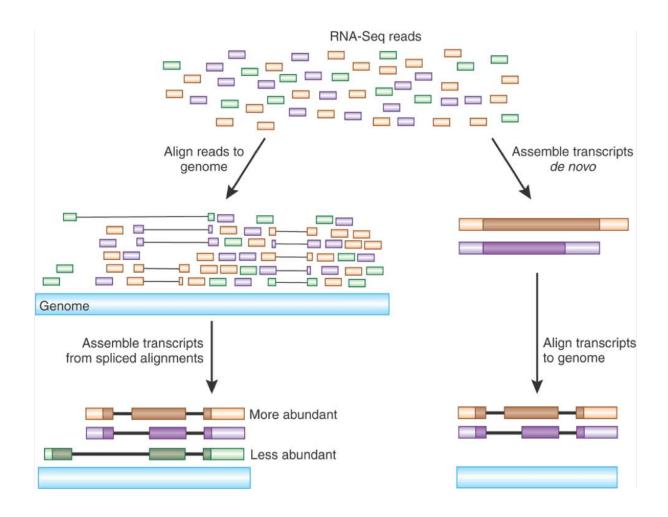




Post Genome (19 years on)

- What do all the genes do?
 - How do they interact?
 - How to cells specialise?
- Junk DNA is not junk after all...
 - 2% Genome contains genes
 - Between 80% (ENCODE) and 25% (Graur et al, 2017) genome seems to have function, usually regulation

Expression Analysis (RNA-Seq)

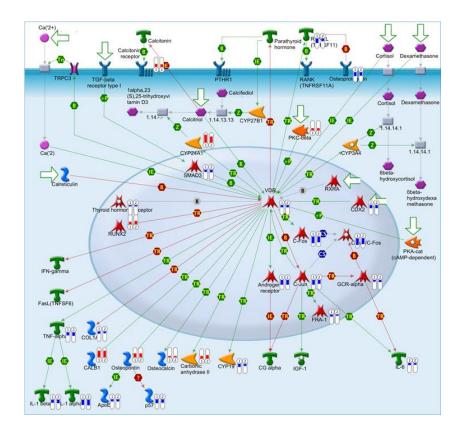


Hass and Zody, Advancing RNA-Seq analysis, Nature Biotechnology 28:421-423

Tools for Alignment

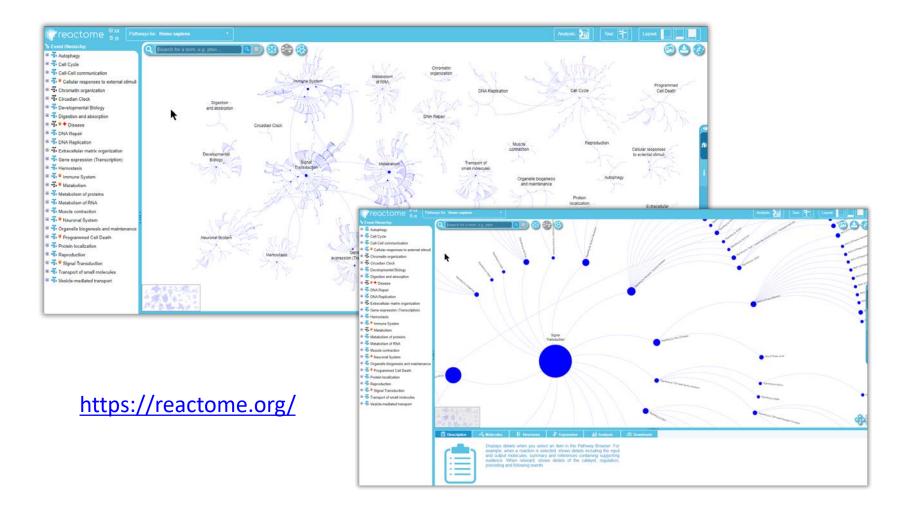
- STAR (Spliced Transcripts Alignment to a Reference). Fast, but uses a lot of memory.
 - Alexander Dobin, Carrie A. Davis, Felix Schlesinger, Jorg Drenkow, Chris Zaleski, Sonali Jha, Philippe Batut, Mark Chaisson, Thomas R. Gingeras, STAR: ultrafast universal RNA-seq aligner, *Bioinformatics*, Volume 29, Issue 1, January 2013, Pages 15–21
- Normalisation and quantification of read counts use:
 - edgeR
 - edgeR: a Bioconductor package for differential expression analysis of digital gene expression data." *Bioinformatics*, 26(1), 139-140)or DESeq2
 - DESeq2
 - Love MI, Huber W, Anders S (2014). "Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2." *Genome Biology*, **15**, 550
- **Salmon** Very fast and does quantification. Uses *quasi-mapping* but no alignments to visualise.
 - Patro, R., Duggal, G., Love, M. I., Irizarry, R. A., & Kingsford, C. (2017). Salmon provides fast and biasaware quantification of transcript expression. Nature Methods.

Functional Annotation



- Metacore
- Ingenuity

Reactome

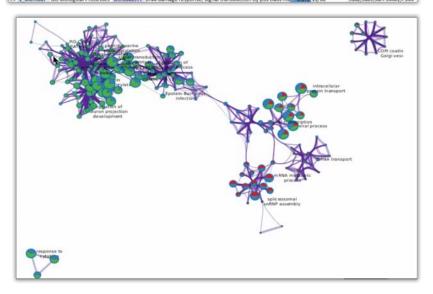


Metascape



http://metascape.org/

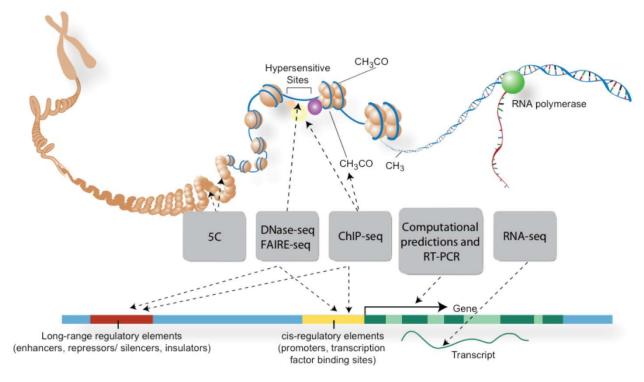
| | A | 8 | C | D | E | F | G | н |
|----|-----------|--------------------------------|------------|---|---------|---------------|--------------|----------------|
| 1 | GroupID | Category | Term | Description | LogP | InTerm_InList | Genes | Symbols |
| 2 | 1_Summary | GO Biological Processes | GO:0016032 | viral process | -18.851 | 49/771 | 156,527,2033 | ADR8K1,ATP |
| 3 | 1_Member | GO Biological Processes | GO:0016032 | viral process | -18.851 | 49/771 | 156,527,2033 | ADR8K1,ATP |
| 4 | 1_Member | GO Biological Processes | GO:0044764 | multi-organism cellular process | -18.549 | 49/784 | 156,527,2033 | ADRBK1,ATP |
| 5 | 1_Member | GO Biological Processes | GO:0044419 | interspecies interaction between organisms | -18.086 | 50/838 | 156,527,1536 | ADR8K1,ATP |
| 6 | 1 Member | GO Biological Processes | GO:0044403 | symbiosis, encompassing mutualism through parasitism | -18.086 | 50/838 | 156,527,1536 | ADR8K1,ATP |
| 7 | 1_Member | GO Biological Processes | GO:0051351 | positive regulation of ligase activity | -12.291 | 16/101 | 5347,5682,56 | PLK1,PSMA1, |
| 8 | 1_Member | GO Biological Processes | GO:0051437 | positive regulation of ubiquitin-protein ligase activity invo | -12.095 | 14/72 | 5347,5682,56 | PLK1.PSMA1. |
| 9 | 1_Member | GO Biological Processes | GO:000020 | protein-polyabigatiinstion- | | 1/212 | 331,4734,568 | XIAP, NEDD4, |
| 10 | 1_Member | GO Biological Processes | GO:005144 | positive regulation of ubiquitin-protein transferase activity | -11.468 | 5/96 | 5347,5682,56 | PLK1,PSMA1 |
| 11 | 1_Member | GO Biological Processes | GO:2000060 | positive regulation of protein ubiquitination involved in u | -10.921 | 14/87 | 5347,5682,56 | PLK1,PSMA1 |
| 12 | 1 Member | KEGG Pathway | hsa03050 | Proteasome | -10.903 | 11/44 | 5682,5683,56 | PSMA1,PSM |
| 13 | 1_Member | GO Biological Processes | GO:0051436 | negative regulation of ubiquitin-protein ligase activity invi | -10.714 | 14/90 | 5347,5682,56 | PLK1,PSMA1 |
| 14 | 1_Member | GO Biological Processes | GO:2000058 | regulation of protein ubiquitination involved in ubiquitin- | -10.322 | 14/96 | 5347,5682,56 | PLK1,PSMA1 |
| 15 | 1 Member | GO Biological Processes | GO:0051340 | regulation of ligase activity | -10.312 | 16/135 | 5347,5682,56 | PLK1,PSMA1 |
| 16 | 1_Member | GO Biological Processes | GO:0051444 | negative regulation of ubiquitin-protein transferase activit | -10.260 | 14/97 | 5347,5682,56 | PLK1,PSMA1 |
| 17 | 1_Member | GO Biological Processes | GO:0051352 | negative regulation of ligase activity | -10.198 | 14/98 | 5347,5682,56 | PLK1,PSMA1 |
| 18 | 1_Member | GO Biological Processes | GO:0051439 | regulation of ubiquitin-protein ligase activity involved in n | -10.137 | 14/99 | 5347,5682,56 | PLK1,PSMA1 |
| 19 | 1_Member | GO Biological Processes | GO:0032446 | protein modification by small protein conjugation | -9.776 | 37/821 | 331,4734,492 | XIAP, NEDD4, |
| 20 | 1_Member | GO Biological Processes | GO:0031398 | positive regulation of protein ubiquitination | -9.579 | 17/174 | 331,5347,568 | XIAP,PLK1,PS |
| 21 | 1_Member | GO Biological Processes | GO:0031145 | anaphase-promoting complex-dependent proteasomal ub | -9.565 | 14/109 | 5347,5682,56 | PLK1,PSMA1 |
| 22 | 1_Member | GO Biological Processes | GO:0002479 | antigen processing and presentation of exogenous peptide | -9.412 | 12/75 | 1536,5682,56 | CYBB, PSMA1 |
| 23 | 1_Member | GO Biological Processes | GO:0051438 | regulation of ubiquitin-protein transferase activity | -9.302 | 15/135 | 5347,5682,56 | PLK1,PSMA1 |
| 24 | 1_Member | GO Biological Processes | GO:0042590 | antigen processing and presentation of exogenous peptide | -9.140 | 12/79 | 1536,5682,56 | CYBB, PSMA1 |
| 25 | 1_Member | GO Biological Processes | GO:1903322 | positive regulation of protein modification by small protei | -9.091 | 17/187 | 331,5347,568 | XIAP, PLK1, PS |
| 26 | 1_Member | GO Biological Processes | GO:0044265 | cellular macromolecule catabolic process | -9.055 | 38/913 | 3146,3837,41 | HMGB1, KPNI |
| 27 | 1_Member | GO Biological Processes | GO:0006521 | regulation of cellular amino acid metabolic process | -9.021 | 11/64 | 5682,5683,56 | PSMA1,PSM/ |
| 28 | 1_Member | GO Biological Processes | GO:0042787 | protein ubiquitination involved in ubiquitin-dependent pr | -8.962 | 16/166 | 4734,5347,56 | NEDD4, PLK1, |
| 29 | 1 Member | GO Biological Processes | GO:0006977 | DNA damage response, signal transduction by p53 class me | -8.872 | 11/66 | 5682.5683.56 | PSMA1.PSM/ |



Gene Expression Omnibus (GEO)

| S NCBI | | | | | | | | | | SEO |
|---------------------------------------|-----------------------|--|---|--|---|-----------------|--------------|------------------|-----------------------|----------------------------------|
| L + GRO + Bannahory Innasar + Samulas | | | | | | | | GEO Publ | ications (FAQ) | MIANE Email C |
| eries Samples. Plat | forms DataSets | Summary Advanced search | | | | | | | | |
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| xession # Title | | | | Sample h | AND DESCRIPTION OF ALL | + Ch + Platform | + Series | Supplementary | Contact | • Release date |
| N952626 SPC/cRaf mouse dysplasia 65 | | | | RINA | Mue museulus | 1 @ GPL6096 | GSE38948 | ♦ CEL CHP | Kishor Bapu Londhe | Jun 23, 20 |
| M952627 SPC/cRal mouse dysplasia 67 | | | | RNA | Mus musculus | 1 🖬 GPL6096 | G5E38948 | 4 CEL CHP | Kishor Bapu Londhe | Jun 23, 20 |
| M952628 SPC/cRaf mouse dysplasia 73 | | | | RNLS, | Mus musculus | 1 🖬 GPL6095 | GSE38948 | 4 CEL OIP | Kishor Bapu Londhe | Jun 23, 20 |
| 4952629 SPC/cRaf mouse dysplasia 73 | 1.7 male 6 months | | | RNA | Mue musculue | 1 🖬 GPL6095 | GSE38948 | & CILOP | Kishor Bapu Londhe | Jun 23, 20 |
| N952630 non-transgenic mouse 65.0 m | nale 7 months | | | RNA | Mus musculus | 1 🗃 GPL6095 | GSE38948 | CEL CHP | Kishor Bapu Londhe | Jun 23, 20 |
| M952631 non-transgenic mouse 67.5 6 | emale 7 months | | | RIPLA | Mus musculus | 1 🖬 GPL6096 | B 05E38948 | A CEL CHP | Kishor Bapu Londhe | Jun 23, 20 |
| H952632 non-transgenic mouse 92.7 6 | emale 11 months | | | RINA | Mus musculus | 1 @ GPL6096 | ■ GSE38948 | + CEL CHP | Kishor Bapu Londhe | Jun 23, 20 |
| 1952633 non-transgenic m HOME | SEARCH SITE MAP | | | | | GEO Pubpo | CSE28948 | 4 CEL OP | Kishor Bapu | Jun 23, 20 |
| M952634 non-transgenic m NCBI | > GEO > Accessio | n Display 🛛 | | | | 90 | GSE38948 | A CELOP | Kishor Bapu | Jun 23, 20 |
| 1952635 SPC/oRal mouse | | | | | | 96 | COSE 38648 | A CEL OV | Londhe Kishor Bapu | Jun 23, 20 |
| M052636 SPC/c-Raf mouse | elp: Mouse over scree | n elements for information. | | | | 0.0 | G G5E38948 | A CEL CHP | Londhe Kishor Bapu | 3un 23, 20 |
| 4932637 SPC/c-Raf mouse Scope | : Self 🔻 Fo | ormat: HTML 🔻 Amou | nt: Quick 🔻 GEO access | sion: Gs | M952626 | GO | G 05E38948 | | Londhe Kishor Bapu | Jun 23, 20 |
| ACCOUNT OF A CALLER AND | | | | | | | | | Londhe | |
| | ple GSM952626 | | Query Da | taSets | for GSM952626 | 96 | GSE38948 | | Kishor Bapu Londhe | Jun 23, 20 |
| 4952639 SPC/c-Raf mouse Statu | is 🗟 | Public on Jun 23, 2014 | | | | 96 | GSE38948 | 4 CEL CHP | Kishor Bapu Londhe | 3un 23, 20 |
| 1952640 SPC/c-Raf mouse Title | | SPC/cRaf mouse dysplasi | ia 65.1 male 6 months | | | 96 | OSE38948 | 4 CEL OIP | Kishor Bagu Londhe | Jun 23, 20 |
| 1932641 SPC/c-Raf mouse Samp | ole type | RNA | | | | 96 | ■ GSE38948 | CEL CHP | Kishor Bapu Londhe | Jun 23, 20 |
| 1952642 SPC/c-Ral mouse | | | | | | 96 | GSE38948 | 4 CEL CHP | Kishor Bapu Londhe | hin 23, 20 |
| | | dysplasia male | | | | 154 | B 05E54416 | & SRA Experiment | Terri DiMaio | Auts 23, 20 |
| 11314709 ECFC_L1_2 Organ | | Mus musculus | | | | 154 | | SRA Experiment | | Jun 23, 20 |
| Chara | | age: 6 months genotype: SPC/cRaf tran tissue: lung dysplastic le Sex: male | | | | 154 | III GSE54418 | SRA Experiment | Tem DiMaio | 3un 23, 20 |
| Grow | th protocol | Four samples each for samples from healthy i micro-dissection. Lung t cryomicrotome (MICROM membrane slide (Zeiss desired cells either dysj normal) or adenocarcin were laser microdissect: | dysplastic and adenoca non-transgenic lungs w tissue slices of 10µm w GmbH, Walldorf, German GmbH) and stained w Jolastic or transgenic (mi nona or healthy non-tra ed and collected in an a tion Pressure Catapultin | ere se ere pr ny) and ith Ha croscop ansgeni idhesiv | lected for laser epared using a l fixed over PEN ematoxylin. The ically unaltered, c alveolar cells e cap using the | | | | | |
| Extra | | total RNA | | | | | | | | |
| | ction protocol | Four samples each for dysplastic and adenocarcinoma stages and 5 samples from healthy non-transgenic lungs were selected for laser micro-dissection. Lung tissue slices of 10µm were prepared using a cryomicrotome (MLCROM GmbH, Walldorf, Germany) and fixed over PEN membrane slide (Zeiss GmbH) and stained with Haematoxylin. The desired cells either dysplastic or transgenic (microscopically unaltered, normal) or adenocarcimona or healthy non-transgenic alveolar cells were laser microdissected and collected in an adhesive cap using the LMPC (Laser microdissection Pressure Catapulting) system. | | | | | | | | |
| Labe | Label biotin | | | | | | | | | |
| Label | | rRNA reduction was done using Ribominus kit (Life technologies, Invitrogen, Carlsbad, California). Single-stranded CDNA was generated from the amplified CRNA with the WT CDNA Synthesis kit (Affymetrix) and then fragmented and labeled with the WT Terminal Labeling Kit (Affymetrix). | | | | | | | | |

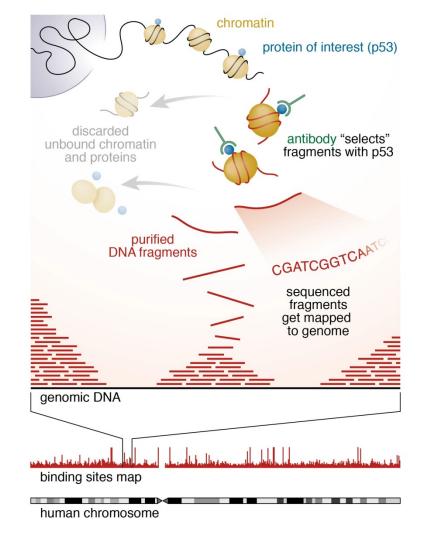
ENCODE (Encyclopedia of DNA Elements)



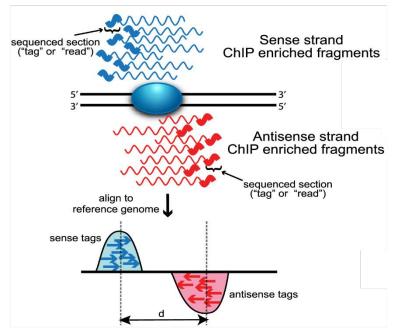
http://genome.ucsc.edu/ENCODE/

What Controls Expression?

ChIP-Seq



Tools for ChIP-Seq



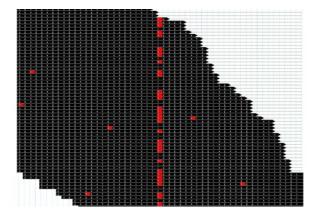
- 1. Align using Bowtie
- Peak call using Model-based Analysis of ChIPseq (MACS)
- 3. Look for motif enrichment using HOMER
- 4. Functional annotation using GREAT



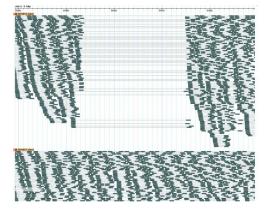
- 1. Ultrafast and memory-efficient alignment of short DNA sequences to the human genome. Genome Biol 10:R25.
- 2. Zhang, Y., Liu, T., Meyer, C.A. *et al.* Model-based Analysis of ChIP-Seq (MACS). *Genome Biol* **9**, R137 (2008) doi:10.1186/gb-2008-9-9-r137
- 3. Heinz S, Benner C, Spann N, Bertolino E et al. Simple Combinations of Lineage-Determining Transcription Factors Prime cis-Regulatory Elements Required for Macrophage and B Cell Identities. Mol Cell 2010 May 28;38(4):576-589.
- 4. Cory Y McLean, Dave Bristor, Michael Hiller, Shoa L Clarke, Bruce T Schaar, Craig B Lowe, Aaron M Wenger, and Gill Bejerano. "GREAT improves functional interpretation of *cis*-regulatory regions". *Nat. Biotechnol.* **28**(5):495-501, 2010

DNA Mutations

Single base mutation



Insertion







The Single Nucleotide Polymorphism database (**dbSNP**) is a public-domain archive for a broad collection of simple genetic polymorphisms.

(http://www.ncbi.nlm.nih.gov/SNP/)

Tools for variant calling

SAMTOOLS

A statistical framework for SNP calling, mutation discovery, association mapping and population genetical parameter estimation from sequencing data. Li H Bioinformatics. 2011 Nov 1;27(21):2987-93.

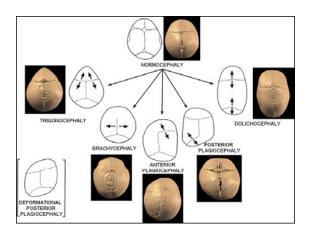


The Genome Analysis Toolkit: a MapReduce framework for analyzing next-generation DNA sequencing data McKenna A, Hanna M, Banks E, Sivachenko A, Cibulskis K, Kernytsky A, Garimella K, Altshuler D, Gabriel S, Daly M, DePristo MA, 2010 *GENOME RESEARCH 20:1297-303*



A unified haplotype-based method for accurate and comprehensive variant calling Daniel P Cooke, David C Wedge, Gerton Lunter bioRxiv 456103; doi: https://doi.org/10.1101/456103

Craniosynostosis



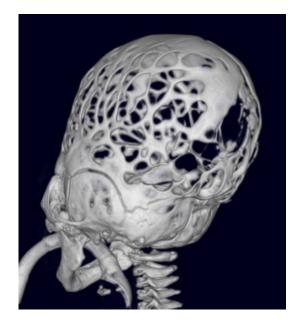






Andrew Wilkie, WIMM

Craniosynostosis

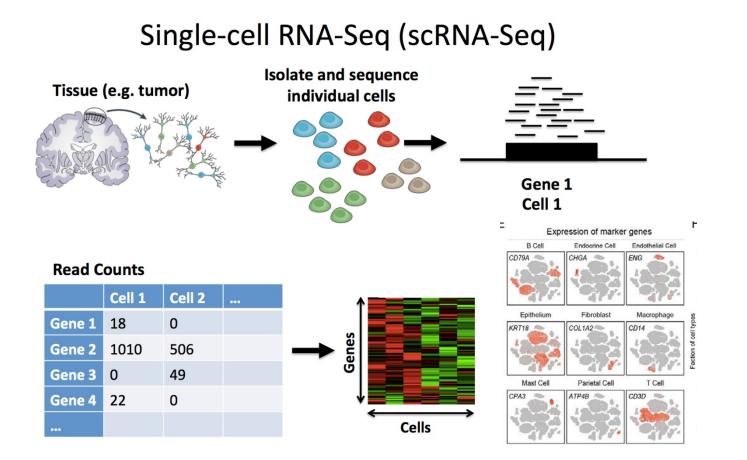






- Genomics England 100,000 Genomes Project (GEL)
- 100,000 patients with rare inherited disease, common cancers and pathogens from the NHS Whole Genome Sequencing
- <u>http://www.genomicsengland.co.uk/</u>
- Secure environment to do analysis

Single Cell Sequencing



https://learn.gencore.bio.nyu.edu/single-cell-rnaseq/



HUMAN CELL cellxgene

cell×gene tabula-muris

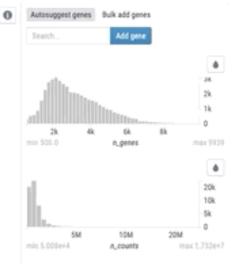
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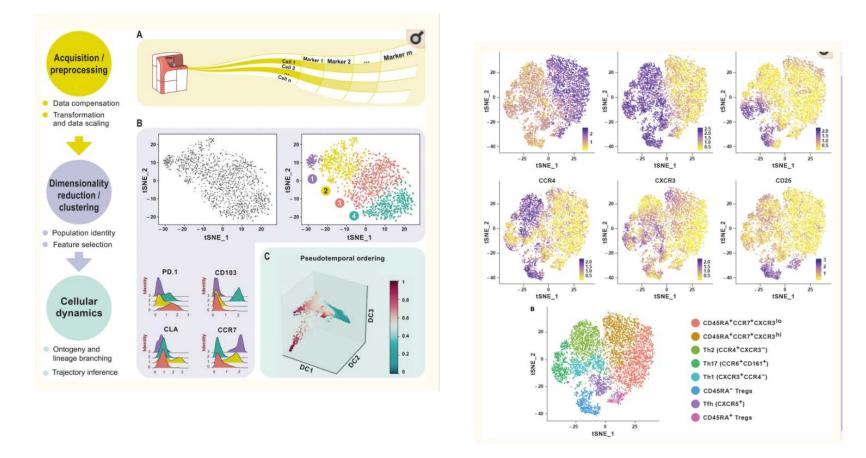
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- clusters_louvain >
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- 🛃 tissue >
- Create new category





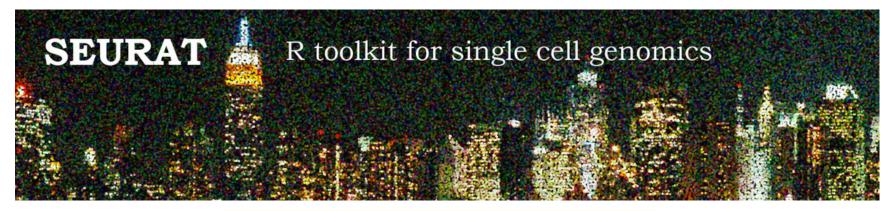
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CyTOF – Single Cell Proteomics



Palit, Subarna, Christoph Heuser, Gustavo P. de Almeida, Fabian J. Theis, and Christina E. Zielinski. 2019. "Meeting the Challenges of High-Dimensional Single-Cell Data Analysis in Immunology." Frontiers in Immunology 10 (July): 1515.

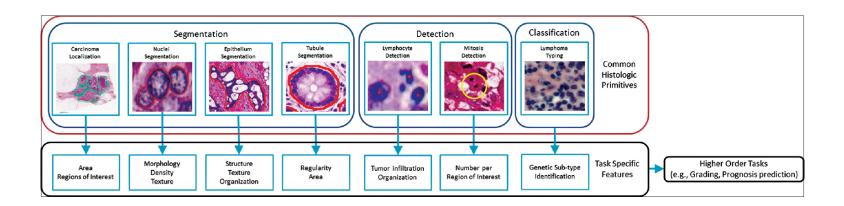
Single Cell Analysis

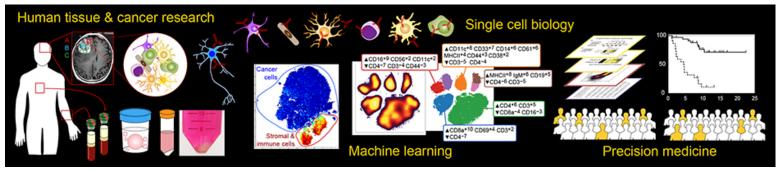


Overview of some of the established single-cell analysis methods.

| Class | Methods | Description |
|--|----------------|---|
| Linear dimensionality reduction | PCA | Cannot account for the smooth nature of single-cell data |
| Non-linear dimensionality reduction | t-SNE | More intuitive representation of high-dimensional data on a lower manifold |
| | UMAP | Scales better and improves global structure of the data compared to t-SNE(see $\underline{Box}\ 1)$ |
| | HSNE | Scales better than conventional t-SNE(see $\underline{Box 1}$) |
| | Diffusion maps | Explores continuity through progression of cell differentiation |
| Clustering methods; single- cell resolution is lost | SPADE | Hierarchical branched tree representation (see $\underline{Box 2}$) |
| | FlowSOM | Self-organizing maps trained to detect cell populations (see $\underline{Box 3}$) |

Machine Learning

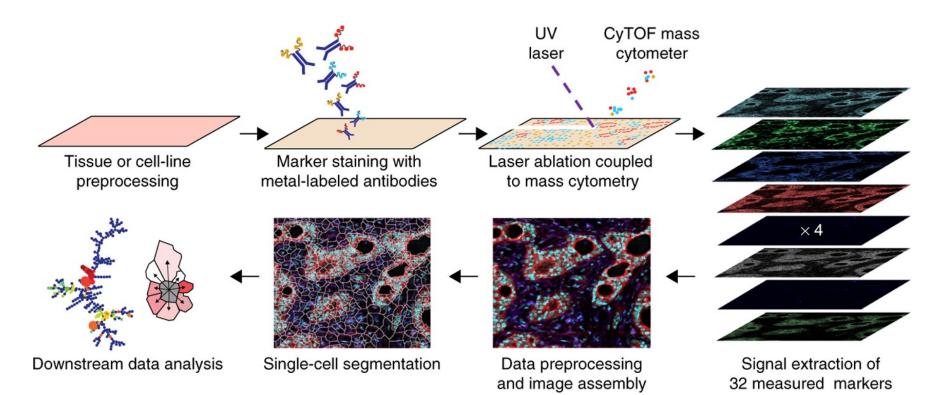




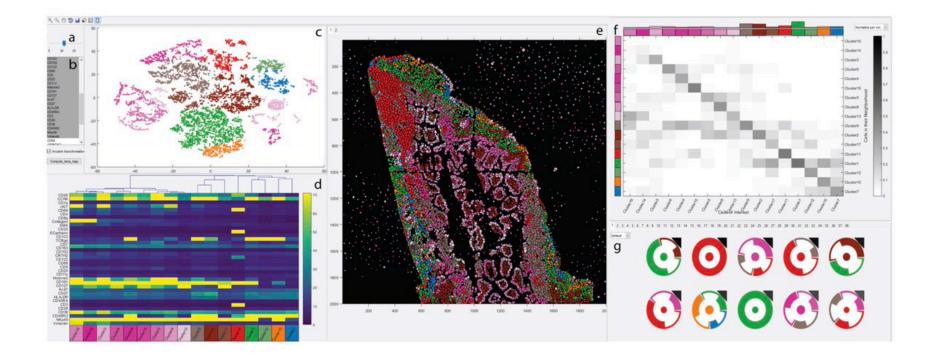
https://my.vanderbilt.edu/irishlab/

Also being applied increasing to all data types e.g. health records, DNA sequences

Spatial Proteomics

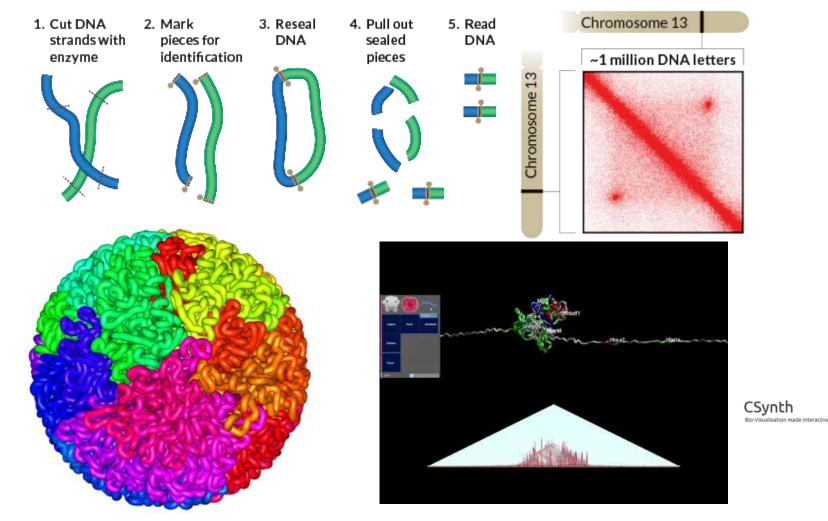


ImaCyte : Analyse Cell Microenvironment



Somarakis, Antonios, Vincent Van Unen, Frits Koning, Boudewijn P. F. Lelieveldt, and Thomas Hollt. 2019. "ImaCytE: Visual Exploration of Cellular Microenvironments for Imaging Mass Cytometry Data." *IEEE Transactions on Visualization and Computer Graphics*, July. https://doi.org/<u>10.1109/TVCG.2019.2931299.</u>

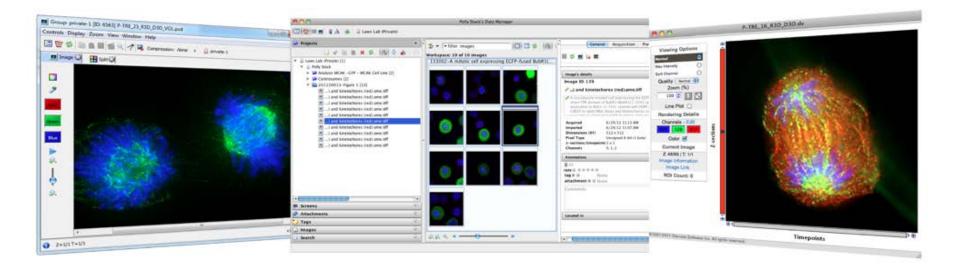
Genome Modelling



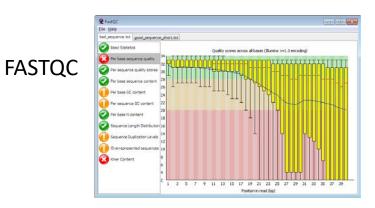
Lieberman-Aiden, Erez, Nynke L. van Berkum, Louise Williams, Maxim Imakaev, Tobias Ragoczy, Agnes Telling, Ido Amit, et al. 2009. "Comprehensive Mapping of Long-Range Interactions Reveals Folding Principles of the Human Genome." Science 326 (5950): 289–93.

Todd, Stephen, Peter Todd, Simon J. McGowan, James R. Hughes, Yasutaka Kakui, Frederic Fol Leymarie, William Latham, and Stephen Taylor. 2020. "CSynth: An Interactive Modelling and Visualisation Tool for 3D Chromatin Structure." Bioinformatics , August. https://doi.org/10.1093/bioinformatics/btaa757.

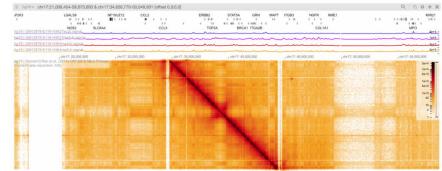
OMERO Image Database



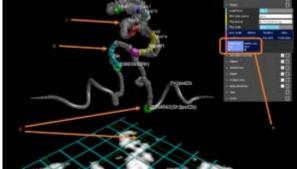
Visualisation



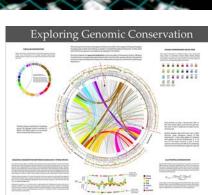
HiGlass



CSynth



Circos



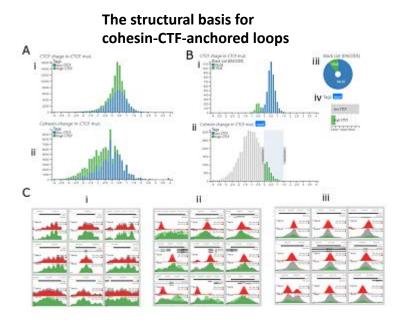
Zegami



BabelVR



Multi Locus Viewer



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Looking at ChIP-seq signals in enhancers and promoters

Multi Locus View : An Extensible Web Based Tool for th Analysis of Genomic Data

Martin J Sergeant, Jim R Hughes, Lance Hentges, Damien J Downes, Stephen Taylor doi: https://doi.org/10.1101/2020.06.15.151837

Under review "Nature Communications Biology"

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Coding in R

- Very good for statistics
- Libraries
 - CRAN (12000 packages)
 - Bioconductor (1823 packages)
- Lots of methods for communicating results
- RStudio is nice graphical environment

Coding in Python

- Most popular language in bioinformatics (and probably data science)
- Used in industry and academic settings
- Very readable
- Great 'glue' for automation
- Lots of libraries for using matrices, machine learning, plotting etc
- <u>https://biopython.org/</u>

Python vs R

| - | - | |
|--|--|---|
| Parameter | R | Python |
| Objective | Data analysis and statistics | Deployment and production |
| Primary Users | Scholar and R&D | Programmers and developers |
| Flexibility | Easy to use available library | Easy to construct new models from scratch. I.e., matrix computation and optimization |
| Learning curve | Difficult at the beginning | Linear and smooth |
| Popularity of Programming Language. Percentage change | 4.23% in 2018 | 21.69% in 2018 |
| Average Salary | \$99.000 | \$100.000 |
| Integration | Run locally | Well-integrated with app |
| Task | Easy to get primary results | Good to deploy algorithm |
| Database size | Handle huge size | Handle huge size |
| IDE | Rstudio | Spyder, Ipthon Notebook |
| Important Packages and library | tydiverse, ggplot2, caret, zoo | pandas, scipy, scikit-learn, TensorFlow, caret |
| Disadvantages | Slow High Learning curve Dependencies between library | Not as many libraries as R |
| Advantages | Graphs are made to talk. R makes it beautiful Large catalog for data analysis GitHub interface RMarkdown Shiny | Jupyter notebook: Notebooks help to share data with colleagues Mathematical computation Deployment Code Readability Speed Function in Python |

Learn both!

See review https://www.guru99.com/r-vs-python.html

CCB Training



INTRODUCTORY COURSES

Introductory short courses cover the Unix command line, programming in R and genomics workflows (ChIP-seq, RNAseq). Find out more



OXFORD BIOMEDICAL DATA SCIENCE TRAINING PROGRAMME

This unique training programme consists of 10 week secondments, first building basic data science skills and then applying them to the analysis of your own biomedical data. **Find out more**

More information

- <u>https://www.imm.ox.ac.uk/research/units-and-centres/mrc-wimm-centre-for-computational-biology</u>
- Google "WIMM CCB"
- Tech Helpdesk : <u>genmail@molbiol.ox.ac.uk</u>
- General Questions : <u>ccb@imm.ox.ac.uk</u>