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eScience and Post-Genome Biomedical Research

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Outline

- 1. General Observations about Postgenomic eScience
- Specific Case Study where Traditional Publication and Archiving Practices are Lacking
- 1. Impact of Emerging Technology on Scale of the eScience Problems





Post-Genomic eScience

• The "Post-Genome" Era

<u>3 Primary Types of Investigation</u>

- 1. Generation of New High-Throughput Data (new "Genome Projects")
- Generating New Data in the Context of Existing Results (Published or in Databases)
- No New Data Exclusive re-use of "Published" Data





Case Study: Genomic Rearrangements or Deletions/Duplications

(Dr. Krishna Rani Kalari, Mayo Clinic)

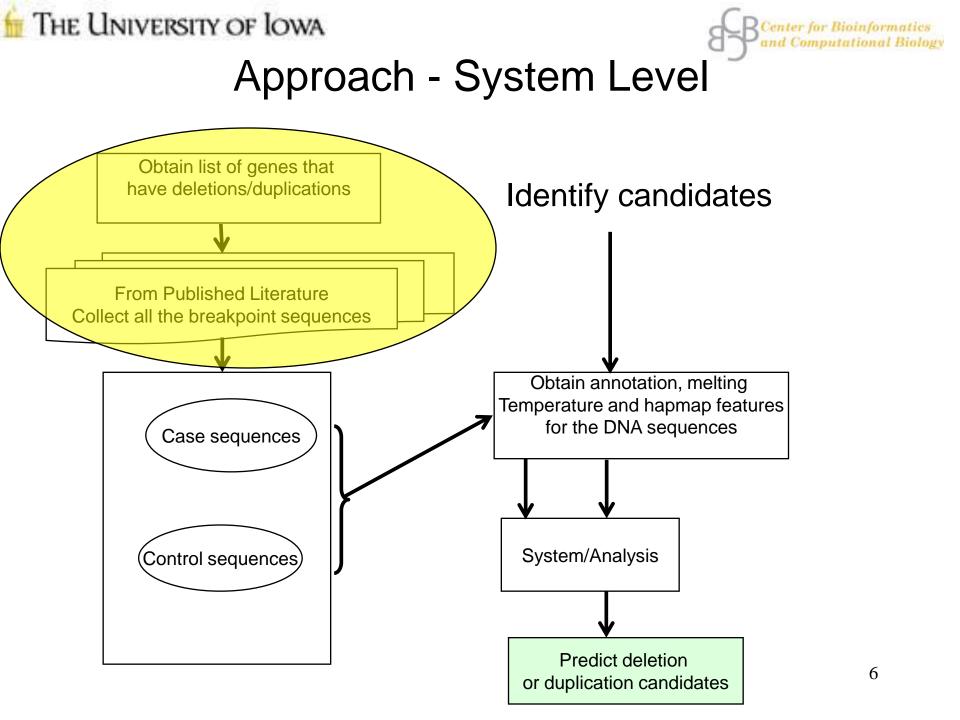
- 1. Goal: Identification of Human disease causing mutations
- 2. Observation: Assays exist to identify deletions and duplications
 - time consuming
 - laborious
 - expensive
- 3. Approach: Develop *In-silico* procedures to identify and prioritize candidate deletion/duplication sites and accelerate the finding of disease mutation discovery





Approach Details

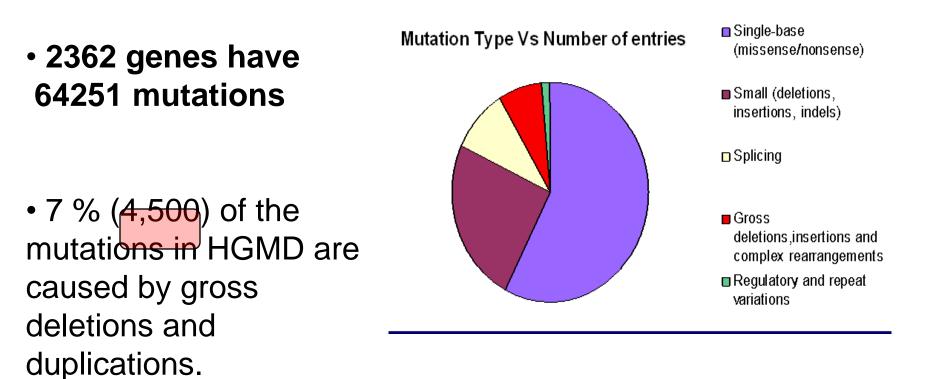
- Construct case and control data sets for all known cases of disease causing unequal recombinations
- 2. Identify and obtain informative sequencebased features to create a training set
- 3. Evaluate machine learning methods on the training set
- 4. Design and develop a computational system to identify and prioritize candidate intragene deletions and duplications







HGMD statistics







HGMD mutation classification

Mutation type	Total number of mutations
Nucleotide substitutions (missene / nonsense)	294
Nucleotide substitutions (splicing)	46
Nucleotide substitutions (regulatory)	0
Small deletions	52
Small insertions	12
Small indels	1
Gross deletions	2
Gross insertions and duplications	0
Complex rearrangements (inversions)	1
Repeat variations	0





HGMD - Gross deletions

Accession Number	Description	Phenotype	Reference
CG035110	ex. 18 (described at genomic DNA level)	Stargardt disease	<u>Yatsenko (2003)</u> <u>Hum Mutat 21,</u> <u>636</u>
CG994802	36 bp nt. 6543 (described at genomic DNA level)	Stargardt disease	<u>Lewis (1999)</u> <u>Am J Hum</u> <u>Genet 64, 422</u>





Local Deletion Database

Welcome to University of Iowa Human GrossDeletions Database

All the information is obtained from HGMD database

This database is maintained by Center for Bioinformatics and Computational Biology. The database consists of all Genes that were found in HGMD database with exonic gross deletions (>20bp). Our database consists of 1463 exonic deletions found in 441 gross deletion genes.

Reln

	GrossDeletions	Phenotype	Reference
1	148 bp incl. ex. 42 (mutation described at cDNA level)	Lissencephaly with cerebellar hypoplasia	<u>1 - Hong (2000) Nat Genet 26, 93</u>

Ghr

GrossDeletions	Phenotype	Reference
1 ex. 3 and ex. 5-6 (mutation described at cDNA level)	Laron dwarfism	<u>1 - Godowski (1989) Proc Natl Acad Sci USA</u> 86, 8083





Of the 4,500 Possible Training Cases, How Many Did We Get???

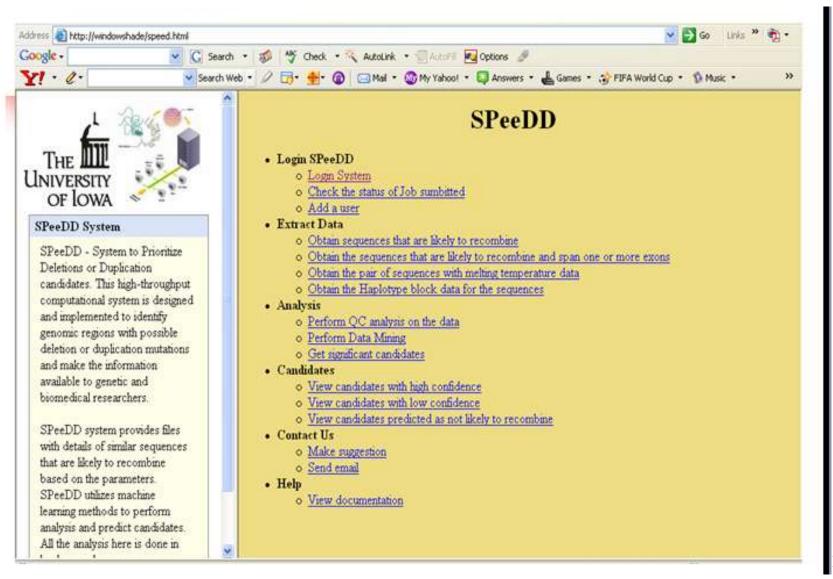
Searched for specific break point information for 1463 IDDs described in HGMD

Identified 102 fully-characterized rearrangement breakpoints (cases)

know exactly where the breakpoint occurs

Identified 2338 matching set of breakpoints for each of the positives for which IDDs have not been observed (<u>controls</u>)

THE UNIVERSITY OF IOWA SPeeDD web-interface



Center for Bioinformatics and Computational Biology





Lessons From Deletion Case Study

- Important results and "data" are buried in traditional forms of scientific publication and dissemination mechanisms (no surprise here).
- Fidelity and throughput of legacy results inadequate
- Necessary data can be requested from investigators
 - In some cases
 - Reference to a changing world of what is assumed to be "known"

• Stay tuned... the problem will only get worse...

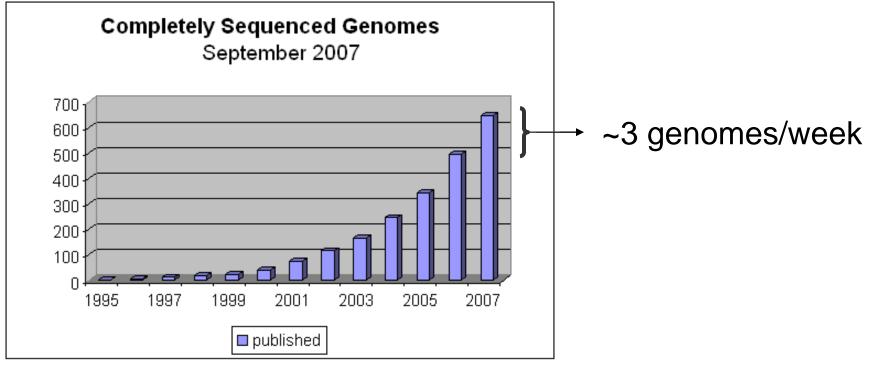




Impact of Emerging Technology on Scale of the eScience Problems

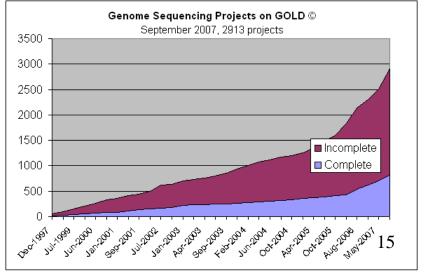
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Genomes Online Database v2.0

www.genomesonline.org







How did we get here?

- Advances in genome sequencing were driven by the Human Genome Project
 - Scale-up started in 1999
 - Resources concentrated in large genome centers
 - Increase in capacity
 - Reduction in cost
 - Economies of scale
 - Improved technology
- Sequencing infrastructure available for non-human projects





George Weinstock, Wash-U/Baylor GSCs)

- Research is Data-driven
 - Produce more data
 - Hypothesis generating > hypothesis testing
 - Community resource projects
 - Rapid data release; prepublication
 - Etiquette in use of prepublication data
 - No intellectual property contraints
- Production is Technology-enabled
 - Develop or acquire new technologies







Human disease study

- 500 cases + 500 controls
- 500 genes, 15 exons/targets per gene
- 2 reads/target
- 15 million reads to screen 1,000 subjects
- 454: 10M rds/d or Solexa: 160M rds/d
- Conclusion: this is a small experiment





Project Jim



- Whole human genome "Proof of Principle"
 - What can be learned from a single genome?
 - What biases exist in the data?
 - What analysis issues arise?
 - Not a consensus sequence but need to capture both alleles:
 6 GB not 3 GB
 - Data quality vs variation: how do you know a variant base is a mutation and not an error





Conclusions: Post-genomic eScience

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