NGS theory: An introduction to next generation sequencing

Richard H. Scheuermann, PhD Director, J. Craig Venter Institute La Jolla, CA, USA



J. Craig Venter Institute (JCVI)

- Non-profit research institute
- 200 staff total
 - 120 in La Jolla, CA
 - 80 in Rockville, MD
- By function:
 - 100 wet lab scientists
 - 70 bioinformatics scientists and software engineers
 - 30 administrative support
- Operating budget of about \$35 million/year
- 40 faculty manage >100 active sponsored projects

- Human Health: Genomes to Clinic
 - Genomic Medicine
 - Human Microbiome
 - Infectious Disease
- Environmental Sustainability and Discovery
 - Microbial & Environmental Genomics
 - Synthetic Biology
 - Microbial Fuel Cells and Bioenergy
- Platforms
 - Sequencing and Bioinformatics
 - Policy Center
 - Education



Viral Genomics

y exposed to a plethora of viruses that cause disease. Highly infectious pathogens such as influenza virus, rotavirus, enteroviruses, and respiratory syncytial virus are a persistent public health concern. Recent outbreaks caused by West Nile, Zika, yellow fever, and Middle-East Respiratory Syndrome (MERS) coronavirus remind us of the ongoing threat posed by emerging and re-remerging viruses. The virology group at the JCVI focuses on defining the genomes of viruses and characterizing virus-host interactions to better understand viral evolution, elucidate the underlying molecular mechanisms that cause disease, and develop novel vaccines and therapeutics.

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Key Staff →

Research Areas >

Research Projects ~



Protective Determinants of ZIKV NS1specific Antibodies



Whole Genome Sequencing of West Nile Virus





Uncovering Role of Vaginal Microbiome and Connection to HPV Infection and **Cervical Cancer**



Chikungunya Virus



Development of a Multivalent Rhinovirus Vaccine



Sequencing of Respiratory Syncytial Virus



Sequencing of Influenza A and Influenza **B** Viruses



Rotavirus Genome Sequencing





Sequencing and VEME

• Application of NGS to the study of Virus Evolution and Molecular Epidemiology

- Track the evolution of viruses over time
- Better understand the selective pressures that drive virus evolution
- Identify the origins (reservoirs) of outbreak strains
- Investigate transmission dynamics
- Identify molecular determinants of host range
- Identify molecular determinants of virulence
- Identify evolutionarily conserved regions for targeted vaccines
- Identify evolutionarily diverse regions for diagnostics



Outline

- A Brief History of Sequencing
- Next Generation Sequencing (NGS) Technologies
- Applications and Challenges of NGS
- Bioinformatics Methods and Resources



• What year was the first whole genome sequence reported?

a) 1969 b) 1977 c) 1981 d) 1985

• For which organism?



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Bacteriophage $\Phi X174$ (5,375 bp)

• What method was used?

article

Nature 265, 687 - 695 (24 February 1977); doi:10.1038/265687a0

Nucleotide sequence of bacteriophage $\phi X174$ DNA

F. SANGER, G. M. AIR^{*}, B. G. BARRELL, N. L. BROWN[†], A. R. COULSON, J. C. FIDDES, C. A. HUTCHISON III[‡], P. M. SLOCOMBE[§] & M. SMITH[¶]

MRC Laboratory of Molecular Biology, Hills Road, Cambridge CB2 2QH, UK

Present addresses : "John Curlin School of Medical Research, Microbiology Department, Canbedrara City ACT 2601, Australia, ¹Department of Biochemistry, University of Bristol, Bristol BS8 1TD, UK, ¹Department of Bacteriology and Immunology, University of North Carolina, Chapel Hill, North Carolina 27514, [§]Max-Planck-Institut für Molekulare Genetik, 1 Berlin 33, FRG, ¹Department of Biochemistry, University of Britsh Columbia, Vancouver BC, Canada V&T 1WS,

A DNA sequence for the genome of bacteriophage $\phi X174$ of approximately 5,375 nucleotides has been determined using the rapid and simple 'plus and minus' method. The sequence identifies many of the features responsible for the production of the proteins of the nine known genes of the organism, including initiation and termination sites for the proteins and RNAs. Two pairs of genes are coded by the same region of DNA using different reading frames.

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Haemophilus influenza (1.8 x 10⁶ bp)

• What method was used?

Whole-Genome Random Sequencing and Assembly of Haemophilus influenzae Rd

Robert D. Fleischmann, Mark D. Adams, Owen White, Rebecca A. Clayton, Ewen F. Kirkoses, Anthory R. Kerlavage, Carol J. Buit, Jaan-Francois Tomb, Brian A. Dougherty, Joseph M. Merrick, Keith McKenney, Granger Sutton, Will Fitzlugh, Chris Fields, Jeannine D. Goczyne, John Scott, Robert Shirley, Li-Ing Lu, Anna Glodek, Janny M. Kelley, Janice F. Weidman, Chevj A. Phillips, Trav Springs, Eva Heddborn, Matthew D. Cotton, Tressa R. Utlerback, Michael C. Hanna, David T. Nguyen, Deborah M. Saudek, Rhonda C. Brandon, Leah D. Fine, Janice L. Frichmann, Joyce L. Furhmann, N. S. M. Geoghagen, Cheryl L. Gnehm, Lisa A. McDonald, Keith V. Small, Claire M. Fraser, Hamilton O. Smith, J. Craig Venteri

An approach for genome analysis based on sequencing and assembly of unelected pieces of DNA from the whole chromosome has been applied to obtain the complete nucleotide sequence (1,330,137 base pairs) of the genome from the bacterium *Hae mophilus influenze* Rd. This approach eliminates the need for initial mapping efforts and is therefore applicable to the vast array of microbial species for which genome maps are unvailable. The *I*. *Influenze* Rd genome sequence (Genome Sequence DataBase accession number L4202) represents the only complete genome sequence form a freeliving organism.

SCIENCE • VOL. 269 • 28 JULY 1995

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Sanger sequencing with fluorescence

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An approach for genome analysis based on sequencing and assembly of unelected pieces of DNA from the whole chromosome has been applied to obtain the complete nucleotide sequence (1,830,137 base pairs) of the genome from the bacterium *Hae mophilus influenze* R4. This approach eliminates the need for initial mapping efforts and is therefore applicable to the vast array of microbial species for which genome maps are uravailable. The *L influenze* R4 genome sequence (Genome Sequence DataBase accession number L42023) represents the only complete genome sequence form a freeling organism.

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1st Generation Nucleic Acid Sequencing

• Maxim-Gilbert chemical method

 Maxam AM, Gilbert W (1977). "A new method for sequencing DNA". Proc. Natl. Acad. Sci. U.S.A. 74 (2): 560–4.

• Sanger chain termination method

- Sanger F, Coulson AR (1975). "A rapid method for determining sequences in DNA by primed synthesis with DNA polymerase". J. Mol. Biol. 94 (3): 441–8.
- Sanger F, Nicklen S, Coulson AR (1977). "DNA sequencing with chain-terminating inhibitors". Proc. Natl. Acad. Sci. U.S.A. 74 (12): 5463–7.



Maxim-Gilbert – chemical cleavage







Sanger - chain termination



ATGC





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Sanger 1st => Sanger 2nd





A chromatogram



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Primer Walking vs Shotgun



Next Generation Sequencing (NGS)

- Massively parallel; sequencing by synthesis or ligation
- Common attributes
 - Random fragmentation of starting DNA
 - Ligation with custom linkers/adapters
 - Library amplification on a solid surface (either bead or glass slide)
 - Direct step-by-step detection of each nucleotide base incorporated during the sequencing reaction
 - Hundreds of thousands to hundreds of millions of reactions imaged per instrument run = "massively parallel sequencing"
 - Shorter read lengths than capillary sequencers











Illumina HiSeq 2000

Illumina MiSeq



1st Generation vs Next Generation





Change in Output



ER Mardis. Nature 470, 198-203 (2011) doi:10.1038/nature09796

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Cost per Human Genome



Wetterstrand KA. DNA Sequencing Costs: Data from the NHGRI Genome Sequencing Program (GSP) Available at: <u>http://www.genome.gov/sequencingcosts/</u>. Accessed 15JUL2019.

JCVI Joint Technology Core



New JCVI Sequencing Core

Illumina NextSeq/MiSeq 800 million reads/runs





Oxford Nanopore MinION



General Workflow



Major NGS Technologies (2nd Generation)

- 454 Life Sciences/Roche
- Ion Torrent/PGM/Life Technologies
- Illumina HiSeq/MiSeq/NovaSeq
- SOLiD/ABI



454 Sequencing



Emulsion PCR



"Break micro-reactors" Isolate DNA containing beads

Perform emulsion PCR



Picotiter Plates





Pyrosequencing



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Shotgun Sequencing by 454









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If that nucleotide is incorporated, enzymes turn PPi by-products into light:







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If that nucleotide is incorporated, enzymes turn PPi by-products into light:



The real power of this method is that it can take place in millions of tiny wells in a single plate at once.







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Ion Torrent/PGM/Life Technologies

- Ion semiconductor sequencing detection of released hydrogen ions using ion-sensitive field-effect transistor technology
- Also uses emulsion PCR amplification step





Detection of pH Change

"Nanowell" solid-state detection



Rothberg et al. Nature (2011)



Illumina/Solexa

- Adapter ligation
- DNA cluster amplification
- Sequencing by synthesis with reversible dye termination



https://www.youtube.com/watch?v=fCd6B5HRaZ8

Illumina Benchtop Sequencers

	iSeq 100	MiniSeq	MiSeq Series O	NextSeq 550 Series O	NextSeq 1000 & 2000
Popular Applications & Methods	Key Application	Key Application	Key Application	Key Application	Key Application
 Large Whole-Genome Sequencing (human, plant, animal) 					
Small Whole-Genome Sequencing (microbe, virus)	•	•	•	•	•
Exome & Large Panel Sequencing (enrichment-based)				•	•
Targeted Gene Sequencing (amplicon- based, gene panel)	•	•	•	•	•
 Single-Cell Profiling (scRNA-Seq, scDNA-Seq, oligo tagging assays) 				•	•
 Transcriptome Sequencing (total RNA- Seq, mRNA-Seq, gene expression profiling) 				•	•
Targeted Gene Expression Profiling	•	•	•	•	•
miRNA & Small RNA Analysis	•	٠	•	•	•
DNA-Protein Interaction Analysis (ChIP- Seq)			•	•	•
Methylation Sequencing				•	•
* 16S Metagenomic Sequencing		•	•	•	•
Metagenomic Profiling (shotgun metagenomics, metatranscriptomics)				•	•
Run Time	9.5-19 hrs	4-24 hours	4–55 hours	12-30 hours	11-48 hours
Maximum Output	1.2 Gb	7.5 Gb	15 Gb	120 Gb	360 Gb*
Maximum Reads Per Run	4 million	25 million	25 million [†]	400 million	1.2 billion*
Maximum Read Length	2 × 150 bp	2 × 150 bp	2 × 300 bp	2 × 150 bp	2 × 150 bp

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Illumina Production-scale Sequencers

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NextSeq 550 Series O



NovaSeq 6000

NextSeq 1000 & 2000

Popular Applications & Methods Key Application Key Application Key Application Large Whole-Genome Sequencing (human, plant, animal) Small Whole-Genome Sequencing (microbe, virus) Exome & Large Panel Sequencing (enrichment-based) Targeted Gene Sequencing (amplicon-based, gene panel) Single-Cell Profiling (scRNA-Seq, scDNA-Seq, oligo tagging assays) Transcriptome Sequencing (total RNA-Seq, mRNA-Seq, gene expression profiling) Chromatin Analysis (ATAC-Seq, ChIP-Seq) Methylation Sequencing Metagenomic Profiling (shotgun metagenomics, metatranscriptomics) Cell-Free Sequencing & Liquid Biopsy Analysis ~13 - 38 hours (dual SP flow cells) ~13-25 hours (dual S1 flow cells) Run Time 12-30 hours 11-48 hours ~16-36 hours (dual S2 flow cells) ~44 hours (dual S4 flow cells) Maximum Output 120 Gb 360 Gb* 6000 Gb Maximum Reads Per Run 400 million 1.2 billion* 20 billion Maximum Read Length 2 × 150 bp 2 × 150 bp 2 × 250**



Third Generation

- Single molecule sequencing
- Helicos, PacBio, Oxford Nanopore



PacBio SMRT



Oxford Nanopore sequencing

- DNA pushed through a nanopore in a lipid membrane
- Speed control provided by a Phi29 DNA polymerase
- Measure changes in the ionic current of an applied electric field







Base calling and accuracy



Magi et al. Briefings in Biotechnology, 2017, 1–17.

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Accuracy improvements





Karst SM, et al. Nature Biotechnology volume 36, pages 190–195 (2018)

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Oxford produces high quality reads >50 kb; longest >800 kb



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Long Read Technology Comparison

- Advantages
 - Full length transcriptomes, including splice variants
 - Resolution of long repeat regions in genomes
 - Genomic structural variants
 - Haplotype phasing
- Disadvantages
 - Higher error rates
 - Lower throughput



NGS applications



https://www.technologynetworks.com/genomics/articles/an-overview-of-next-generation-sequencing-346532

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NGS Challenges

- Requires infrastructure to transfer, manage, store and process large amounts of data
- Some technologies are error prone and generate systematic biases and so data quality is an issue
- Assembling complete genomes from short sequence reads can be very difficult, especially in repeat regions and for shotgun sequences without reference genomes
- Difficult to establish phase of variants in diploid genomes
- Requires skills in bioinformatics that may not be readily accessible to many research labs



BIOINFORMATICS METHODS



• How many base pairs (bp) are there in a human genome?



- How many base pairs (bp) are there in a human genome?
 - \sim 3 billion (haploid)
- When was the first human genome sequence completed?

a) 1990 b) 1995 c) 2000 d) 2003



Science

AAAS

The Sequence of the Human Genome

Craig Venter

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J. Craig Venter et al.

Science **291**, 1304 (2001); DOI: 10.1126/science.1058040

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articles

Initial sequencing and analysis of the human genome

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NATURE | VOL 409 | 15 FEBRUARY 2001 | www.nature.com

International Human Genome Sequencing Consortium

860

• How many base pairs (bp) are there in a human genome?

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- How much did it cost to sequence the first human genome? a) \$2 million b) \$20 million c) \$200 million d) \$2 billion
- How long did it take to sequence the first human genome? a) 1 year b) 4 years c) 7 years d) 13 years



The Sequence of the Human Genome J. Craig Venter et al. Science 291, 1304 (2001); DOI: 10.1126/science.1058040

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The Sequence of the Human Genome J. Craig Venter *et al. Science* **291**, 1304 (2001); DOI: 10.1126/science.1058040

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- Whose genome was it?

several people (public); Craig Venter (private)



But was it?

- The <u>Human Genome Project</u>, completed in 2003, covered about 92% of the total human genome sequence. The technologies to decipher the gaps that remained didn't exist at the time.
- Since then, researchers have developed better laboratory tools, computational methods, and strategic approaches. The final, complete human genome sequence was described in a set of six papers in the April 1, 2022, issue of *Science*.
 - Telomere to Telomere (T2T) consortium led by researchers at NIH's National Human Genome Research Institute (NHGRI), the University of California, Santa Cruz, and the University of Washington, Seattle.
 - Long read sequencing technology used:
 - PacBio HiFi 20,000 letters with nearly perfect accuracy
 - Oxford Nanopore up to 1 million DNA letters at a time—with modest accuracy.
 - Added nearly 200 million nucleotides (8% of the genome) newly added sequences were mainly in the centromeres and telomeres
 - https://www.nih.gov/news-events/nih-research-matters/first-complete-sequence-human-genome



BIOINFORMATICS METHODS



NGS Processing Workflows



NGS Processing Workflows



FASTQ File Format

- The FASTQ format allows the storage of both sequence and quality information for each read.
- This is a compact text-based format that has become the *de facto standard for* storing data from next generation sequencing experiments.



FASTA File Format

>gb:0N369979|0rganism:Influenza A virus|Strain Name:A/Alaska/04/2022|Segment:8|Subtype:H3N2|Host:Human GTGACAAAGACATAATGGATTCCAACACTGTCAAGTTTCCAGGTAGATTGCTTCCTTGGCACATCCG GAAACAAGGCATAATGGATTGCAACCACAAAACAGGACACCCATCTCTGCAGAGTCGCCC CTAGGGGGAAGAGGCAATACTCTCGGTCTAGACATCAAATCAGCCACCCATGTTGGAAAGCAAATCGTAG AAAAGATTCTGAAAGGAGAATCTGATGAGGACTTAAAATGACCATGGTCTCAACACCTGCTCGCGAATCAGGGGA GACCTCTTTGCATCAGAATGGATCAGGCACTTAAAATGACCATGGTCCAAGCACCTGCTTCGCCGATGA CATAACTGACATGACTGATGGAGATTGTCAAGAACTGGTTCATGCCCAAGCACGAGAGGGGGAA GGACCTCTTTGCATCAGAATGGATCAGGCAATCAGGAGAAAAACATCATGTTTAAAAGCGAATTTCAATG TGATTTTTGGCACGGCTAGAGGCCATAGTATTGCTAAGGGCTTTCACCGAAGAGGGGGAGCAATTGTTGGGGCCCTCATG GGAGGACTTCTCTTTTCCAGGACATACTAGTGAGAAAAACATCATGTTTAAAAGCGAATTGTGGGGCACTCCAG AATCTCACCATTGCCTTCTTTTCCAGGACATACTATGGAGAAAAACATCATGCTGAGGGAGCCA ATCTCACCATTGCATGGAAAGAACAACATCATGCGAAAAATCTACAGAGGATGGGGGACCAC GGAGGACTTCTTGAATGGAAAGAATGAATCTACAGAGAATCGCATTGGGGTCCTCATG GGAGGACTCCCACTTCACTGCAAAACAGAAACGAGAAATCGCAATGGCGAGAACAGCA GTCATGAGGAATGGGGGACCTCCACTTGCAAAACAGGAAAACGAGAAATGGCGAGAACAGCTAGGTCAGA AGTTTGAAGGGAATAGGAAGCACTGCAACAACGAGAAACGAGAAACGAGAAACGAGAACAGCTAGGCAAGCA GTCATGAGAGAATAAGATGGCGGGACCTCCACTTGCAACAACGAGAAACGAGAAACGAGAAACGGGGAACAGCTAGGTCAGA AGTTTGAAGAGAATAAGAGGCCTGATTGAAAACGAGGAAACGAGAAACGAGAAACGGGGAACAGCTAGGTCAGA ACAAATAACACTTCCAGGCAACTACCACTTGCAACACGTGAGACAGGGGAACAGGCAACTTTCCACTTG ACAAATAACATTCATGCAAAAAAACAC



FastQ Format



: = phred of 25 ; = phred of 26

the member of a pair, /1 or /2 (paired-end or							
mate-pair reads only)							

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'v'-coordinate of the cluster within the tile

index number for a multiplexed sample (0 for

1973

#0

/1

no indexing)

Assessing Quality: Phred scores

• Phred quality scores were originally produced by the Phred base calling program using a statistical analysis of Sanger chromatogram trace files in support of the Human Genome Project. Subsequently adapted to NGS technologies for judging qualities of sequences.

 $Q = -10 \log_{10} P_e$ $P_e = error probability$ of a given base call

Phred Quality Score	Probability of incorrect base call	Base call accuracy
10	1 in 10	90 %
20	1 in 100	99 %
30 🔶	1 in 1000	99.9 %
40	1 in 10000	99.99 %
50	1 in 100000	99.999 %

Phred quality scores are logarithmically linked to error probabilities



Trimming



Trimming objectives and methods

- Objectives
 - Remove primers/adapters
 - Remove low quality positions and reads
- Methods
 - Trimmomatic
 - <u>https://github.com/timflutre/trimmomatic</u>
 - Cutadapt
 - <u>https://github.com/marcelm/cutadapt</u>
 - TrimGalore (Cutadapt + FastQC)
 - <u>https://github.com/FelixKrueger/TrimGalore</u>
 - Fastp
 - https://github.com/OpenGene/fastp

Single	index		P5	SP1	Insert	SP2	i7 index	P7		
Unique	e dual index	P5	i5 index	SP1	Insert	SP2	i7 index	P7		
Dual in	dex UMI	P5	i5 index	SP1	Insert	SP2	i7 index	UMI A	P7	
	Flow cell binding sequence: Platform-specific sequences for library binding to instrument									
	Sequencing primer sites: Binding sites for general sequencing primers									
	Sample indexes: Short sequences specific to a given sample library									
	Molecular index/barcode: Short sequence used to uniquely tag each molecule in a given sample library									

Insert: Target DNA or RNA fragment from a given sample library


Alignment/Assembly



What is a sequence assembly?

- "An assembly is a hierarchical data structure that maps the sequence data to a putative reconstruction of the target"
- Miller JR, Koren S and Sutton G. 2010. Assembly algorithms for next-generation sequencing data. *Genomics* 95:315-32





Two Classes of Assembly

- <u>Alignment-based mapping and assembly (aka reference-based</u> <u>alignment)</u> refers to reconstruction of the underlying sequence facilitated by alignments to a previously resolved reference sequence.
- <u>*De novo* assembly</u> refers to reconstruction of the underlying sequence without a previously resolved reference sequence.



Alignment Mapping and Assembly of Short Reads

- *Strategy:* When a suitable reference sequence is available, index the reference genome sequence and search it efficiently
- For this purpose, map-alignment sequence assembly approaches generally use a computing strategy called Burrows–Wheeler transformation and indexing to notably reduce compute time and memory usage
 - **BWA Burrows-Wheeler Aligner** <u>http://bio-bwa.sourceforge.net</u>
 - Bowtie An ultrafast memory-efficient short read aligner Ben Langmead and Cole Trapnell, University of Maryland <u>http://bowtie-bio.sourceforge.net/</u>
 - MAQ Mapping and Assembly with Quality Heng Li, Sanger Centre <u>http://maq.sourceforge.net/maq-man.shtml</u>
 - SOAPaligner/soap2 (Short Oligonucleotide Analysis Package) <u>http://soap.genomics.org.cn/soapaligner.html</u>
 - Also see <u>https://en.wikibooks.org/wiki/Next_Generation_Sequencing_(NGS)/Alignment</u>



Current List of Alignment Mappers



DNA mappers are plotted in blue, RNA mappers in red, miRNA mappers in green, and bisulfite mappers in purple

bisulfite mappers in purple.

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https://www.ebi.ac.uk/~nf/hts_mappers/

Considerations

- The short reads do not come with position information, that is, we do not know what part of the genome they came from; we need to use the sequence of the read itself to find the corresponding region in the reference sequence.
- The reference sequence can be quite long (~3 billion bases for human), making it a daunting task to find a matching region.
- Since our reads are short, there may be several, equally likely places in the reference sequence from which they could have been read. This is especially true for repetitive regions.
- If we were only looking for perfect matches to the reference, we would never see any variation. Therefore, we need to allow some mismatches and small structural variation (InDels) in our reads.
- Any sequencing technology produces errors. Similar to the "real" variations, we need to tolerate a low level of sequencing errors in our reads and separate them from the "real" variations later.

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• We need to do that for each of the millions of reads in our sequencing data

Indexing and searching reference

- Short reads with long reference
- Preprocess reference
- Book index => topics (in alphabetical order) and locations
- Instead of words, we'll use k-mer substrings
- Offset for position
- Query => index hits
- Extend by verification => match

Reference:
TACCTTCCCAGGTA
K-mers $(k=5)$:
TACCT 1
ACCTT 2
CCTTC 3
CTTCC 4
TTCCC 5
TCCCA 6
CCCAG 7
CCAGG 8
CAGGT 9
AGGTA 10

- K-mer index: ACCTT 2 AGGTA 10 CAGGT 9 CCAGG 8 CCCAG 7 CCTTC 3 CTTCC 4 TACCT 1 TCCCA 6 TTCCC 5
- Read: CCAGGTA
- Match: 8



De novo Assembly of Short Reads

- Strategy: When a suitable reference sequence is not available, assemble reads de novo
- For this purpose, assembly approaches generally use a computing strategy called de Bruijn graph of k-mers to notably reduce compute time and memory usage, but de Bruijn graphs are inherently very large due to the observed number of distinct k-mers in the target sequences and hence require significant computer memory to hold the constructed graph

• SPAdes

- Max A. Alekseyev and Pavel Pezner
- <u>https://github.com/ablab/spades</u>

• Velvet

- Daniel Zerbino and Ewan Birney, EMBL-EBI
- <u>http://www.ebi.ac.uk/~zerbino/velvet/</u>

• <u>ABySS</u>

- Inanç Birol, Shaun Jackman, Steve Jones and others, GSC
- <u>http://www.bcgsc.ca/platform/bioinfo/software/abyss</u>

• <u>ALLPATHS-LG</u>

- Jaffe et al CRD, Broad Institute
- <u>http://www.broadinstitute.org/software/allpaths-lg/blog/</u>
- **SOAPdenovo**
 - Li et al. Beijing Genome Institute
 - <u>http://soap.genomics.org.cn/soapdenovo.html</u>
- Additional software listed in the Earl DA et al. 2011. Assemblathon 1: A competitive assessment of de novo short read assembly methods. <u>http://genome.cshlp.org/content/early/2011/09/16/gr.126599.111</u> J. Craig Venter^{**}

INSTITUTE

Current Methods

Name ¢	Description / e Methodology	Technologies +	Author 🕈	Presented / Last updated	Licence* ¢	Homepage ¢
ABySS	parallel, paired-end sequence assembler designed for large genome assembly of short reads (genomic and transcriptomic), employ a Bloom filter to De Bruijn graph	Illumina	[8][9]	2009 / 2017	OS	linkമ
DISCOVAR	paired-end PCR-free reads (successor of ALLPATHS-LG)	Illumina (MiSeq or HiSeq 2500)	[10]	2014	OS	linkଙ୍କ
DNA Baser Sequence Assembler	DNA sequence assembly with automatic end trimming & ambiguity correction. Includes a base caller.	Sanger, Illumina	Heracle BioSoft SRL	2018.09	C (\$69)	NA
DNASTAR Lasergene Genomics	(large) genomes, exomes, transcriptomes, metagenomes, ESTs	Illumina, ABI SOLiD, Roche 454, Ion Torrent, Solexa, Sanger	DNASTAR	2007 / 2016	с	link ଜ
Newbler	genomes, ESTs	454, Sanger	454 Life Sciences	2004/2012	С	linkജ
Phrap	genomes	Sanger, 454, Solexa	Green, P.	1994 / 2008	C / NC-A	link ଜ
Plass	Protein-level assembler: assembles six-frame-translated sequencing reads into protein sequences	Illumina	[11]	2018 / 2019	OS	linkଜ
Ray	a suite of assemblers including de novo, metagenomic, ontology and taxonomic profiling; uses a De Bruijn graph		[12]	2010	OS	linkമ
SPAdes	(small) genomes, single-cell	Illumina, Solexa, Sanger, 454, Ion Torrent, PacBio, Oxford Nanopore	[13]	2012 / 2021	OS	linkജ
Velvet	(small) genomes	Sanger, 454, Solexa, SOLiD	[14]	2007 / 2011	OS	linkଙ
HGAP	Genomes up to 130 MB	PacBio reads	[15]	2011 / 2015	OS	link 🖗
Falcon	Diploid genomes	PacBio reads	[16]	2014 / 2017	OS	link 🖗
Canu	Small and large, haploid/diploid genomes	PacBio/Oxford Nanopore reads	[17]	2001 / 2018	OS	link ଜ
MaSuRCA	Any size, haploid/diploid genomes	Illumina and PacBio/Oxford Nanopore data, legacy 454 and Sanger data	[18]	2011 / 2018	OS	linkള
Hinge	Small microbial genomes	PacBio/Oxford Nanopore reads	[19]	2016 / 2018	OS	link 🖗
Trinity	transcriptome assemblies by de Bruijn graph	Illumina RNA-seq	[20]	2011		link ଜ
***************************************	- Course C. Commercial C /NC A. Commercial but free for one commercial and condemice					

*Licences: OS = Open Source; C = Commercial; C / NC-A = Commercial but free for non-commercial and academics

https://en.wikipedia.org/wiki/Sequence_assembly https://en.wikipedia.org/wiki/De_novo_sequence_assemblers



Hybrid Approach

- Align reads to reference if you can
- *De novo* assemble remaining reads for identification of novel regions/genomes



SAM/BAM File Format

- Sequence Alignment Map (SAM) format
- BAM compressed binary version

Example Header Lines

		í.
@HD	VN:1.0 S0:coordinate	į.
@SQ	SN:1 LN:249250621 AS:NCBI37 UR:file:/data/local/ref/GATK/human_g1k_v37.fasta M5:1b22b98cdeb4a9304cb5d48026a85128	í.
@SQ	SN:2 LN:243199373 AS:NCBI37 UR:file:/data/local/ref/GATK/human_g1k_v37.fasta M5:a0d9851da00400dec1098a9255ac712e	L
@SQ	SN:3 LN:198022430 AS:NCBI37 UR:file:/data/local/ref/GATK/human_glk_v37.fasta M5:fdfd811849cc2fadebc929bb925902e5	į.
@RG	ID:UM0098:1 PL:ILLUMINA PU:HWUSI-EAS1707-615LHAAXX-L001 LB:80 DT:2010-05-05T20:00:00-0400 SM:SD37743 CN:UMCORE	į.
@RG	ID:UM0098:2 PL:ILLUMINA PU:HWUSI-EAS1707-615LHAAXX-L002 LB:80 DT:2010-05-05T20:00:00-0400 SM:SD37743 CN:UMCORE	ł.
@PG	ID:bwa VN:0.5.4	i.
@PG	ID:GATK TableRecalibration VN:1.0.3471 CL:Covariates=[ReadGroupCovariate, QualityScoreCovariate, CycleCovariate, DinucCovariate, TileCovariate], default_read_group=null, default	i

In the alignment examples below, you will see that the 2nd alignment maps back to the RG line with ID UM0098.1, and all of the alignments point back to the SQ line with SN:1 because their RNAME is 1.

Example Alignments

This is what the alignment section of a SAM file looks like:

											1.1
1:497:R:-272+13M17D24M 113	1	497	37	37M 15	100338662	0	CGGGTCTGACCTGAGGAGAACTGTGCTCCGCCTTCAG	0;==-==9;>>>>>=>>>>>>>>>>>>>>>>>>>>>>>>>>	XT:A:U	NM:i:0	SM
19:20389:F:275+18M2D19M 99	1	17644	0	37M =	17919 314	TATGACT	rgctaataatacctacacatgttagaaccat >>>>>	<pre>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>></pre>	M0098:1	XT:A:R	NM
19:20389:F:275+18M2D19M 147	1	17919	0	18M2D19M	= 17644	-314	GTAGTACCAACTGTAAGTCCTTATCTTCATACTTTGT	;44999;499<8<8<<<8<<>><< <;<<< ><<	XT:A:R	NM:i:2	SN
9:21597+10M2I25M:R:-209 83	1	21678	0	8M2I27M =	21469 -244	CACCACI	ATCACATATACCAAGCCTGGCTGTGTCTTCT <;9<<5	><<<<>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>	NM:i:2	SM:i:0	AN

In this example, the fields are:

Field	Alignment 1	Alignment 2	Alignment 3	Alignment 4
QNAME	1:497:R:-272+13M17D24M	19:20389:F:275+18M2D19M	19:20389:F:275+18M2D19M	9:21597+10M2I25M:R:-209
FLAG	113	99	147	83
RNAME	1	1	1	1
POS	497	17644	17919	21678
MAPQ	37	0	0	0
CIGAR	37M	37M	18M2D19M	8M2I27M
MRNM/RNEXT	15	=	=	=
MPOS/PNEXT	100338662	17919	17644	21469
ISIZE/TLEN	0	314		
SEQ	CGGGTCTGACCTGAGGAGAACTGTGCTCCGCCTTCAG	TATGACTGCTAATAATACCTACACATGTTAGAACCAT	GTAGTACCAACTGTAAGTCCTTATCTTCATACTTTGT	CACCACATCACATATACCAAGCCTGGCTGTGTCTTCT
QUAL	0;==-==9;>>>>>=>>>>>>>>>>>>>>>>>>>>>>>>>>	>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>	;44999;499<8<8<<<8<><< ?<;<<>><<	<;9<<5><<>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>>
TAGs	XT:A:U NM:i:0 SM:i:37 AM:i:0 X0:i:1 X1:i:0 XM:i:0 XO:i:0 XG:i:0 MD:Z:37	RG:Z:UM0098:1 XT:A:R NM:i:0 SM:i:0 AM:i:0 X0:i:4 X1:i:0 XM:i:0 X0:i:0 XG:i:0 MD:Z:37	XT:A:R NM:i:2 SM:i:0 AM:i:0 X0:i:4 X1:i:0 XM:i:0 XO:i:1 XG:i:2 MD:Z:18^CA19	XT:A:R NM:i:2 SM:i:0 AM:i:0 X0:i:5 X1:i:0 XM:i:0 XO:i:1 XG:i:2 MD:Z:35

http://genome.sph.umich.edu/wiki/SAM https://en.wikipedia.org/wiki/SAM_(file_format)



Visualization – IGV Pile Up



J. Craig Venter

ONT base calling, quality control, and methylation/modification profiles

- Changes in current as a molecule is pulled through a pore is changed into "squiggle plots" by the *Minknow* software - this can also be used to determine methylation or BrdU incorporation (fast5 files)
- *Guppy* converts the fast5 files into fastq files, i.e., standard base calls with quality scores in either real time or post run
 - Real time Guppy combined with K-mer analysis allows for "adaptive sequencing" where pores will recognize unwanted DNA in real time, reverse polarity, and reject the untargeted sequences
- *Nanofilt* is used for quality trimming of fastq files
- The fastq and fast5 are used in combination to determine methylation profiles using *Nanopolish*



What can you do with these data?

- High quality genome assembly *Flye*
- Viral variant detection- Variabel
- De-novo bulk transcript assembly with splice variants- *Pinfish*
- 10X single cell transcript assembly-*Sockeye*
- Differential methylation analysispycoMeth

ARTICLE

tps://doi.org/10.1038/s41467-022-28852-1 OPEN

Check for updates

Rescuing low frequency variants within intra-host viral populations directly from Oxford Nanopore sequencing data

Yunxi Liu^{1,3}, Joshua Kearney^{1,3}, Medhat Mahmoud $^{\odot}_{\circ}^2$, Bryce Kille¹, Fritz J. Sedlazeck $^{\odot}_{\circ}^{1,2}$ & Todd J. Treangen $^{\odot}_{\circ}^{183}$

Article Published: 01 April 2019

Assembly of long, error-prone reads using repeat graphs

Mikhail Kolmogorov, Jeffrey Yuan, Yu Lin & Pavel A. Pevzner 🖂

 Nature Biotechnology
 37, 540–546 (2019)
 Cite this article

 22k
 Accesses
 874
 Citations
 164
 Altmetric
 Metrics



NGS Processing Workflows



Variant Calling

- *Strategy:* Determine the presence of sequence variations within a sample (alleles in a diploid organism, quasi-species in population samples, somatic mutation heterogeneity in cancer tissues) and sequence variations between sets of samples or between a sample and a reference
- Single nucleotide variants (SNVs), copy number variants (CNVs), insertions and deletions (Indels), rearrangements
 - GATK Broad Institute <u>https://www.broadinstitute.org/gatk/</u>
 - SAMtools Wellcome Trust Sanger Institute https://github.com/samtools/samtools
 - Atlas-Indel2 Baylor College of Medicine <u>http://sourceforge.net/p/atlas2/wiki/Atlas-Indel/</u>
 - Bambino National Cancer Institute https://github.com/NCIP/cgr-bambino



Variant detection through NGS



Figure 3 | Types of genome alterations that can be detected by second-generation sequencing. Sequenced

Meyerson et al. NRG 2010



VCF File Format

- Developed for the 1000 Genomes Project
- Store only the variant information SNVs and Indels

```
##fileformat=VCFv4.0
##fileDate=20110705
##reference=1000GenomesPilot-NCBI37
##phasing=partial
##INFO=<ID=NS,Number=1,Type=Integer,Description="Number of Samples With Data">
##INFO=<ID=DP,Number=1,Type=Integer,Description="Total Depth">
##INFO=<ID=AF,Number=.,Type=Float,Description="Allele Frequency">
##INFO=<ID=AA,Number=1,Type=String,Description="Ancestral Allele">
##INFO=<ID=DB,Number=0,Type=Flag,Description="dbSNP membership, build 129">
##INFO=<ID=H2,Number=0,Type=Flag,Description="HapMap2 membership">
##FILTER=<ID=q10, Description="Quality below 10">
##FILTER=<ID=s50,Description="Less than 50% of samples have data">
##FORMAT=<ID=GQ,Number=1,Type=Integer,Description="Genotype Quality">
##FORMAT=<ID=GT, Number=1, Type=String, Description="Genotype">
##FORMAT=<ID=DP,Number=1,Type=Integer,Description="Read Depth">
##FORMAT=<ID=HQ,Number=2,Type=Integer,Description="Haplotype Quality">
#CHROM POS
                       REF ALT
                                    QUAL FILTER INFO
                                                                                  FORMAT
                                                                                                             Sample2
                                                                                                                            Sample3
             ID
                                                                                              Sample1
2
       4370
            rs6057
                            А
                                                NS=2;DP=13;AF=0.5;DB;H2
                                                                                  GT:GQ:DP:HQ 0 0:48:1:52,51 1 0:48:8:51,51 1/1:43:5:...
                       G
                                    29
                                        .
2
       7330
                       т
                            A
                                    3
                                         q10
                                                NS=5;DP=12;AF=0.017
                                                                                  GT:GQ:DP:HQ 0 0:46:3:58,50 0 1:3:5:65,3
                                                                                                                            0/0:41:3
2
       110696 rs6055
                       Α
                            G,T
                                    67
                                         PASS
                                                NS=2;DP=10;AF=0.333,0.667;AA=T;DB GT:GQ:DP:HQ 1 2:21:6:23,27 2 1:2:0:18,2 2/2:35:4
2
       130237 .
                       т
                                    47
                                         .
                                                NS=2;DP=16;AA=T
                                                                                  GT:GQ:DP:HQ 0 0:54:7:56,60 0 0:48:4:56,51 0/0:61:2
2
       134567 microsat1 GTCT G,GTACT 50
                                        PASS NS=2;DP=9;AA=G
                                                                                              0/1:35:4
                                                                                                             0/2:17:2
                                                                                                                            1/1:40:3
                                                                                  GT:GQ:DP
```

https://vcftools.github.io/specs.html

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Visualization – Circos Plots



M. Stratton et al. Nature, 458 (2009)



NGS Processing Workflows



Variant Interpretation

- *Strategy:* Compare against databases of known variants with functional information and predict functional effects of novel variants
- Databases of known variants dbSNP, ClinVar, OMIM, Cosmic, Ensembl Variation
- Functional predictions SIFT and PolyPhen-2
 - SG-Adviser Scripps Research Institute, <u>https://genomics.scripps.edu/ADVISER/</u>
 - Annovar University of Pennsylvania, <u>http://www.openbioinformatics.org/annovar/</u>
 - SnpEff Wayne State University, <u>http://snpeff.sourceforge.net/</u>
 - SuRFR University of Edinburgh, <u>http://www.cgem.ed.ac.uk/resources/SuRFR/SuRFR_Sweave_v1.pdf</u>
 - Phen-Gen Genome Institute of Singapore, <u>http://phen-gen.org/</u>
 - o Promethease SNPedia, <u>http://www.snpedia.com/index.php/Promethease/</u>



Interpretation of genetic variants

- Substitutions translated bioinformatically
- SIFT probability that a substitution is tolerated
 - \circ < 0.05 is deleterious.
- PolyPhen categorical definitions
 "benign", "possibly damaging" and "probably damaging"
- Protein structural mapping



J. Craig Venter[™]

Parson et al., Science, (2008)

SG-Adviser Example

Chromosom	Begin	End	VarType	Reference	Allele	Gene_Type	Location	Coding_Impact	Protein_Impact_Prediction(Polyphen)	Protein_Impact_Prediction(SIFT)	COSMIC_Gene~NumSamples	Gene_Symbol
chr3	140419772	140419773	snp	G	т	Protein_Co	ing Exon_5	Nonsynonymous	probably damaging	INTOLERANT	carcinoma\$NS\$malignant_me	TRIM42
chr5	79030319	79030320	snp	С	G	Protein_Cod	ing Exon_2	Nonsynonymous	benign	TOLERANT	NS\$carcinoma~41	CMYA5
chr11	5011043	5011044	snp	Α	G	Protein_Co	ing Exon_2	Nonsynonymous	benign	TOLERANT	carcinoma~6	MMP26
chr14	102514936	102514937	snp	G	Α	Protein_Cod	ing Exon_74	Nonsynonymous	benign	TOLERANT	glioma\$carcinoid-endocrine_t	DYNC1H1
chr16	3255108	3255109	snp	т	Α	Protein_Co	ing Exon_1	Nonsynonymous	probably damaging	INTOLERANT	carcinoma~2	OR1F1
chr20	39991131	39991132	snp	C	Α	Protein_Cod	ing Exon_4	Nonsynonymous	probably damaging	INTOLERANT	carcinoma~3	EMILIN3
chr19	54697485	54697486	snp	С	т	Protein_Coo	ing/ Exon_6/	Nonsynonymous	possibly damaging///-///-	INTOLERANT///-///-	carcinoma~1///carcinoma~1/	TSEN34
chr17	4689235	4689236	snp	С	G	Protein_Co	ing/Exon_2/	Nonsynonymous	probably damaging///-///probably dat	INTOLERANT///-///INTOLERANT,	carcinoma~5///carcinoma~5/	VM01
chr7	144077014	144077015	snp	т	C	Protein_Cod	ing/ Exon_15	Nonsynonymous	probably damaging///probably damag	INTOLERANT///TOLERANT	carcinoma\$NS\$malignant_me	ARHGEF5
chr15	31220826	31220827	snp	G	Α	Protein_Cod	ing/Exon_11	Nonsynonymous	possibly damaging///possibly damaging	TOLERANT///INTOLERANT	-~-///-~-	MTMR15
chr18	28588004	28588005	snp	Α	т	Protein_Co	ing/Exon_11	Nonsynonymous	possibly damaging///possibly damaging	INTOLERANT///INTOLERANT	carcinoma\$malignant_meland	DSC3
chr6	47650081	47650082	snp	т	C	Protein_Cod	ing/ Exon_6/	Nonsynonymous	probably damaging///probably damag	INTOLERANT///INTOLERANT///-	malignant_melanoma\$carcing	GPR111
chr12	88514807	88514808	snp	Α	G	Protein_Coo	ing/ Exon_14	Nonsynonymous	benign///benign///-	N/A///TOLERANT///-	NS\$carcinoma~18///NS\$carci	CEP290
chr6	97726730	97726731	snp	С	Α	Protein_Coo	ing/ Exon_3/	Nonsynonymous	probably damaging///probably damag	INTOLERANT///INTOLERANT///-,	-~-///-~-///-~-///-~-	C6orf167
chr2	241702740	241702741	snp	С	Α	Protein_Cod	ing/ Exon_20	Nonsynonymous	benign///benign///benign	INTOLERANT///TOLERANT///INT	NS\$carcinoma\$lymphoid_nec	KIF1A
chr7	43351425	43351426	snp	G	Α	Protein_Coo	ing/ Exon_4/	Nonsynonymous	benign///benign///benign	TOLERANT///TOLERANT///TOLE	primitive_neuroectodermal_t	KIAA0322///HE
chr5	176520167	176520168	snp	С	т	Protein_Coo	ing/ Exon_9/	Nonsynonymous	benign///benign///benign///-///-	TOLERANT///TOLERANT///TOLE	malignant_melanoma\$NS\$ca	FGFR4
chr1	103548371	103548372	snp	Т	С	Protein_Co	ing/Exon_2/	/ Nonsynonymous	unknown///unknown///unknown///u	r N/A///N/A///N/A///N/A	NS\$carcinoma\$glioma~45///N	COL11A1
chr1	180063725	180063726	snp	С	т	Protein_Cod	ing/Exon_34	Nonsynonymous	probably damaging///probably damag	INTOLERANT///INTOLERANT///II	NS\$carcinoma\$lymphoid_nec	CEP350
chr6	76023775	76023776	snp	С	Т	Protein_Co	ing/Exon_5/	Nonsynonymous	probably damaging///probably damag	INTOLERANT///INTOLERANT///II	carcinoma\$malignant_meland	FILIP1

Gene_Symbol	DrugBank	Reactome_Pathway	Gene_Onotology	Disease_Ontology	ACMG_Score_	Research~Disease	_Entry~E	xplanation										
TRIM42	-	-		-	.~.~.													
CMYA5	-	-	•	-	.~.~.													
MMP26	DB00786		GO:0006508~proteolysis	DOID:5616~intraepithelial	.~.~.													
VINC1H1	-	Cell Cycle~Cell Cycle,	GO:0006810~transport\$	(-	4~Charcot-Ma	arie-Tooth disease,	axonal(D	YNC1H1)\$M	Aental retard	ation(DYNC1	H1)~Rare, Ar	nino Acid Cha	nge Predicted	Neutral				
JR1F1	-	GPCR downstream sig	GO:0007165~signal trans	-	.~.~.													
EMILIN3	-	-		DOID:162~cancer	.~.~.													
SEN34	-	-	GO:0000379~tRNA-type	-	3~Pontocereb	ellar hypoplasia(T	SEN34)~R	are, Amino	Acid Change	Predicted Da	amaging with	Low Confider	ice					
VM01	-	-	GO:0030704~vitelline m	-	.~.~.													
ARHGEF5	-		GO:0035023~regulation	DOID:684~hepatocellular c	.~.~.													
MTMR15	-	-		-	.~.~.													
DSC3	-		GO:0007155~cell adhesid	DOID:9256~colorectal cano	3~Hypotricho	sis & recurrent skir	n vesicles	(DSC3)~Rar	e, Amino Acio	Change Pre	dicted Dama	ging with Low	Confidence					
GPR111	-		GO:0007186~G-protein d	-	.~.~.													
CEP290	-	Cell Cycle~Cell Cycle,	GO:0015031~protein tra	DOID:14791~Leber congen	4~Joubert syn	drome(CEP290)\$J	oubert sy	ndrome, Se	nior-Loken ty	pe(CEP290)	Senior-Loker	syndrome(Cl	P290)\$Mecke	l syndron				
C6orf167	-	•	-	-	.~.~.													
KIF1A	DB03431~	-	GO:0007018~microtubul	-	3~Intellectual	disability, nonsyna	dromic(Kl	F1A)\$Spast	ic paraparesi	(KIF1A)~Rar	e, Amino Acio	d Change Pred	icted Damagin	g with Lo				
KIAA0322///HI	E-///-	-///-	-///GO:0006464~protein	-///-	.~.~.													
FGFR4	DB00039	Downstream signalin	GO:0006468~protein pho	DOID:10008~malignant nee	3~Cancer, acc	elerated progressi	on, associ	iation with(FGFR4)~Rare	, Amino Acid	Change Pred	icted Damagi	ng with Low Co	onfidence				
COL11A1	-		GO:0007155~cell adhesid	DOID:657~adenoma\$DOID	4~Fibrochond	rogenesis(COL11A	1)\$Marsh	all / Stickle	r syndrome(0	OL11A1)\$O	teoarthritis,	early-onset ?(COL11A1)\$Lum	nbar disc	(rai	a V	onf	tor
CEP350	-	-	-	-	.~.~.										. 1 a i	yν	CIII	121
FILIP1	-	-	-		.~.~.											_		
				1										1 11	SΤ	ΙТ	U	ΤE