



Benchmarking of Variant Callers & Analysis Tools

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Illumina Sequencing





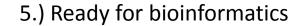
1.) Cells (e.g cancerous or matched normal)

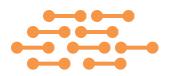




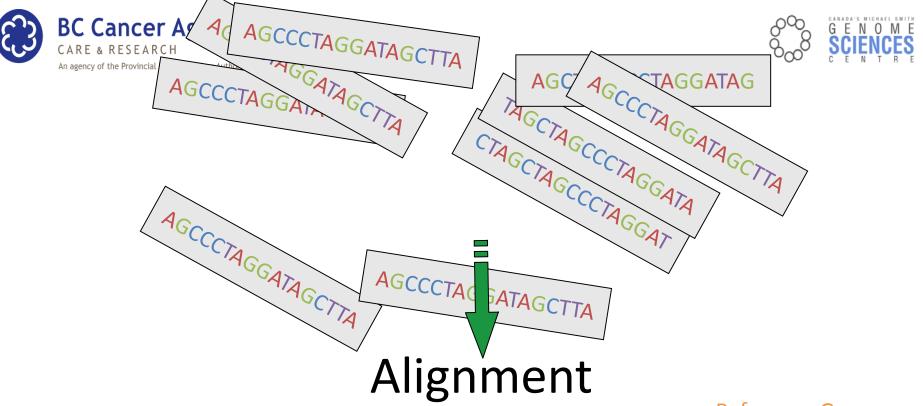


2.) Isolate DNA





3.) Sheared DNA, with sequencing adapters



Reference Genome

ACTCGCTAGCTAGCCCTAGGATAGCTTAGAGACCCTCGCGAAATAGACCCTCGAT

AGCCCTAGGATAGCTTA

AGCTAGCCCTAGGATAG

TAGCTAGCCCTAGGATA

CTAGCTAGCCCTAGGAT

<u>GCTAGCTAGCCCTAGGA</u>

GCTAGCTAGCCCTAGGA

<u>GCTAGCTAGCCCTAGGA</u>

GCTAGCTAGCCCTAGGA

ACCCTCGCGAAATAGAC

ACCCTCGCGAAATAGAC

GACCCTCGCGAAATAGA

AGACCCTCGCGAAATAG

TAGAGACCCTCGCGAAA

CTTAGAGACCCTCGCGA

CTTAGAGACCCTCGCGA

3





After alignment...

- Binary alignment file (BAM file)
- Binary file reports where in reference genome reads are aligned to
- Cancer BAM vs Matched Normal BAM

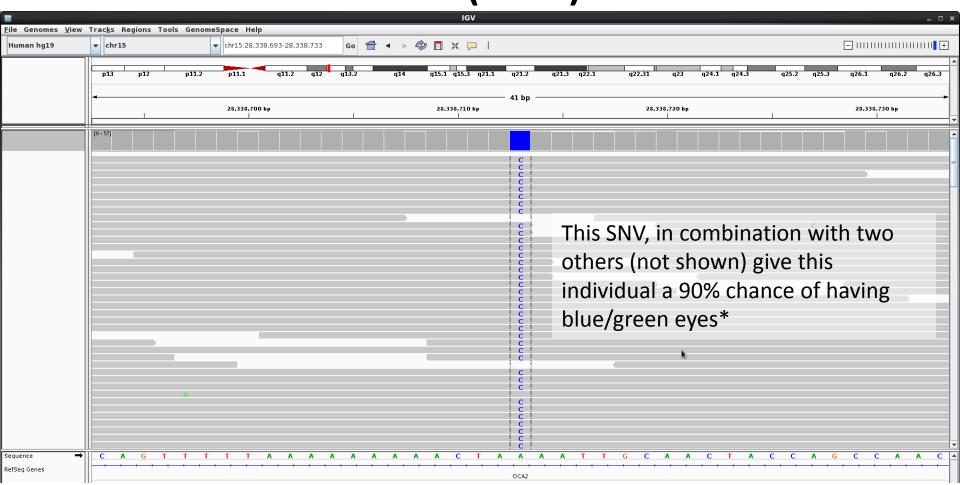


1.) Cells (e.g cancerous or matched normal)



Viewing Alignments: Single Nucleotide Variants (SNVs)









What are somatic variants?

- Variation in DNA that occurs after conception
- Not in germ cells and thus not passed on to future generations
- Somatic variants may act as cancer drivers
 - e.g. KRAS G12D gain of function mutation in colorectal cancer
- Variant in tumour, not matched normal





Challenges of calling somatic variants

- Sample purity (e.g. biopsy with low tumour content)
- Sequencing biases and errors
- Alignment ambiguities
- Differences in variant calling algorithms

How do we estimate the accuracy of a somatic variant calling pipeline?





Construct a ground truth data set to estimate somatic variant prediction accuracy

Goals:

- 1. For a cancer sample, **COLLECT** independent somatic variant data sets from different organizations
- 2. CURATE a ground truth set of somatic SNVs and indels
- **3. ESTIMATE** accuracy of somatic variants from paired somatic analysis pipeline







Melanoma cell line

- Isolated from 45 year old Caucasian male
- COLO-829BL is the matched normal made from peripheral blood





Goal #1: Four Independent COLO-829 Somatic Variant Data Sets

Wellcome Trust Sanger Institute

- Pleasance E et al, 2010 (EDP)
- 75 bp reads, tumour/normal: ~40X/~32X
- Sanger validated 497 somatic SNVs and 62 somatic indels

Translational Genomics Research Institute (TGEN)

- Craig DW et al, 2016
- 112 bp reads, tumour/normal: ~80X each

Ground Truth

Complete Genomics – BGI (CG)

- Unpublished data
- Proprietary sequencing technology (DNA Nanoball Arrays)

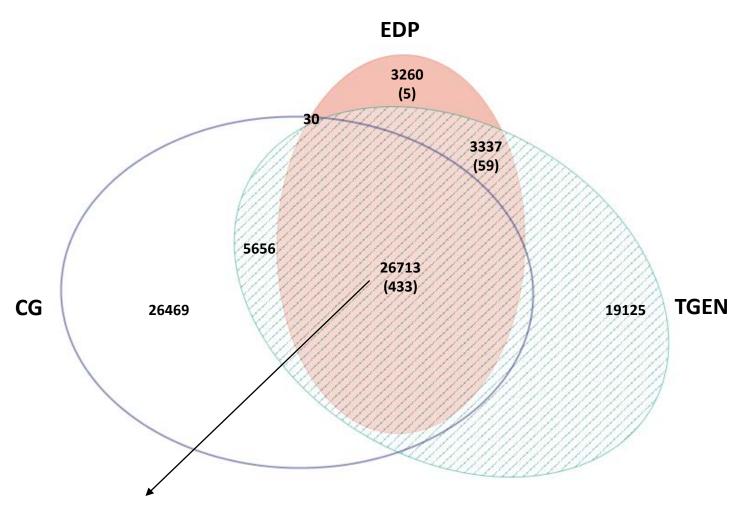
In-house (BCGSC)

- Craig DW et al, 2016
- 125 bp reads, tumour/normal: ~100X each





COLO-829 Somatic SNV Data Sets

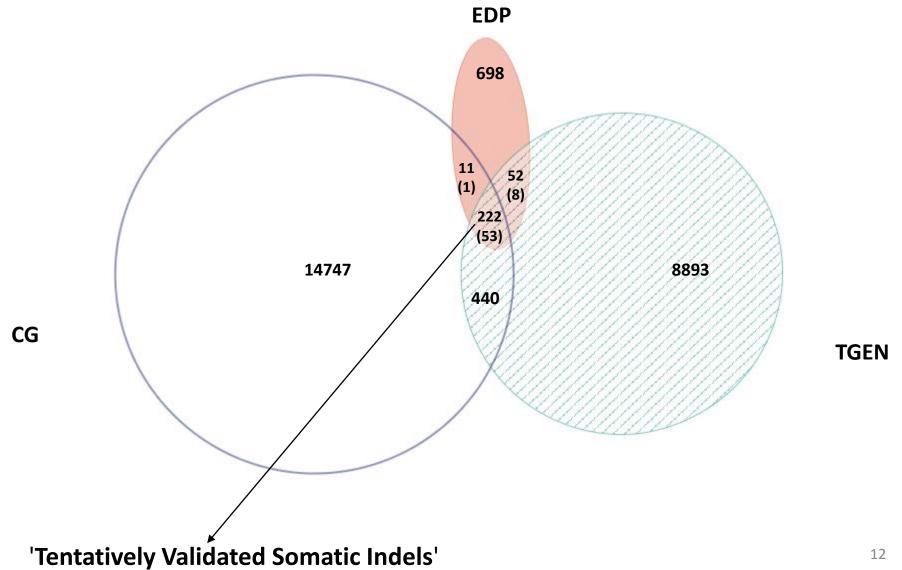


'Tentatively Validated Somatic SNVs'





COLO-829 Somatic Indel Data Sets







Goal #2: Curated ground truth variants

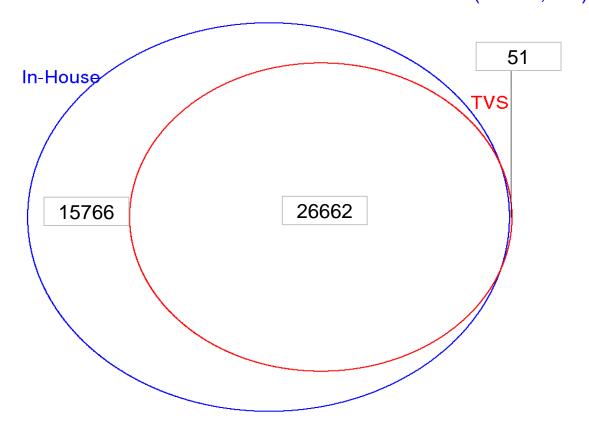
- Tentatively Validated Somatic SNVs/Indels
 - These variants should be called
 - Any variant that is missed is a potential false negative
- Union Set of Somatic SNVs/Indels
 - These represent all possible variants that could be called
 - Any extra variant is a potential false positive
- **Next step:** Compare our in-house somatic variants to these two data sets.





Somatic SNV Sensitivity

How many of the Tentatively Validated SNVs (TVS, n = 26,713) were called in our In-House Somatic SNVs (n = 42,428)?

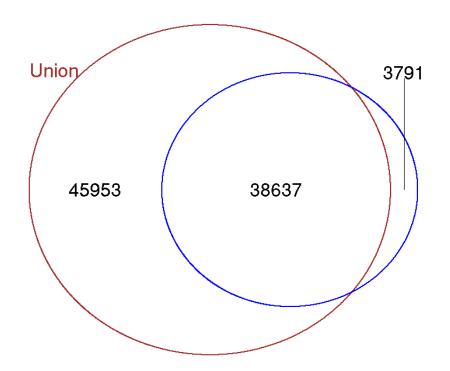








How many of the In-House Somatic SNVs (n = 42,428) were not seen in the Union SNVs (n = 84,590)?



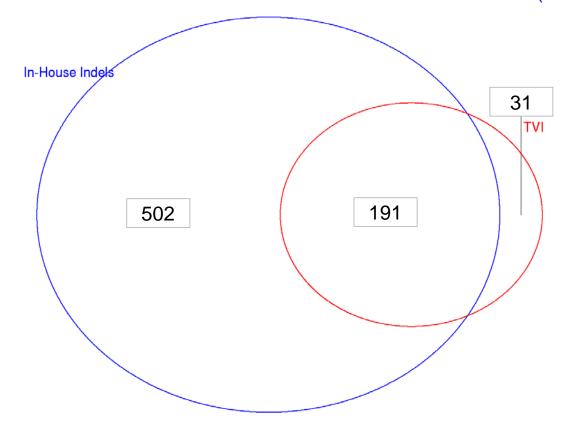
Specificity Estimate: 91.1% (3791/42428 ~ 8.9%)







How many of the Tentatively Validated Indels (TVI, n = 222) were called in our In-House Somatic SNVs (n = 693)?



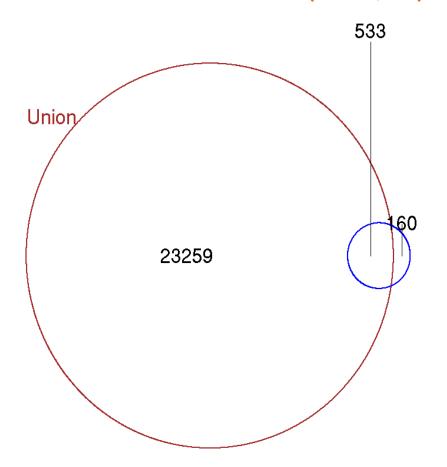
Sensitivity Estimate: 191/222 ~ 86.0% of TVI were called







How many of the In-House Somatic Indels (n = 693) were not seen in the Union Indels (n = 23,792)?



Specificity Estimate: 76.9% (160/693 ~ 23.1%)





Review

- Ground truth data set constructed for COLO-829 somatic SNVs and indels and can be used to estimate sensitivity/specificity of somatic predictions
- This data set can be used to benchmark publically available somatic SNV/indel prediction algorithms



SNV Prediction Callers



- Queried BioStars/PubMed/Google for somatic callers
 - Identified 28 tools
- Requirements Tool is...
 - published in a peer reviewed journal
 - maintained (e.g. bug fixes, updates, etc)
 - supported (e.g. authors respond to questions on BioStarts, etc)
 - outputs Variant Call Format (VCF) file
- 11/28 tools passed these requirements





Benchmarking Tools

- Strategy to evaluate candidate tools for inclusion in production pipelines:
 - Literature search
 - Local installation of tools
 - Construct ground truth set
 - Compare results with respect to evaluation criteria
 - Choose tool(s) to use
- Once every few years, revaluate and, if necessary, update tools/version





Software	Year Published	Organization
LoFreq 2.1.1	2012	Genome Institute of Singapore
MuTect 1.1.4	2013	Broad
Shimmer 20150410	2013	NHGRI/NIH
FreeBayes 0.9.21	2012	Boston College
Platypus 20150421	2014	The Wellcome Trust Centre for Human Genetics
SAMTools 1.2	2009	Sanger/Broad





Software	Year Published	Organization
SomaticSniper 20150411	2011	WUSTL
VarScan2 2.3.7	2012	WUSTL
RTG Somatic 3.4.3	2015	Real Time Genomics, Inc
Strelka 1.0.14	2012	Illumina, Inc
MutationSeq 4.3.5	2011	BC Cancer Research Centre





Run each somatic caller on our in-house tumor/normal COLO-829 sample

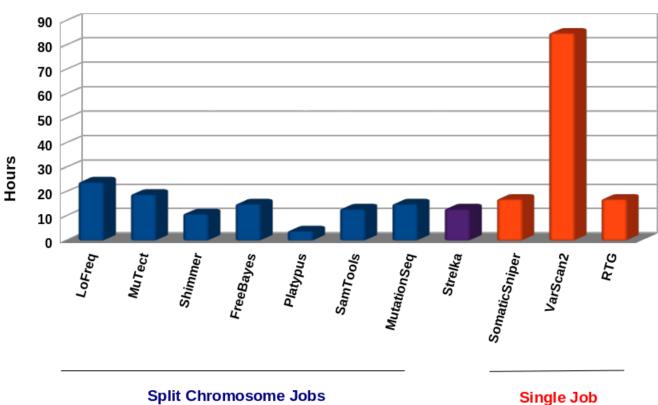
For each somatic caller, report:

- Wall clock run times
- Number of somatic SNVs called
- Estimate sensitivity/specificity





Wall Clock Run Times



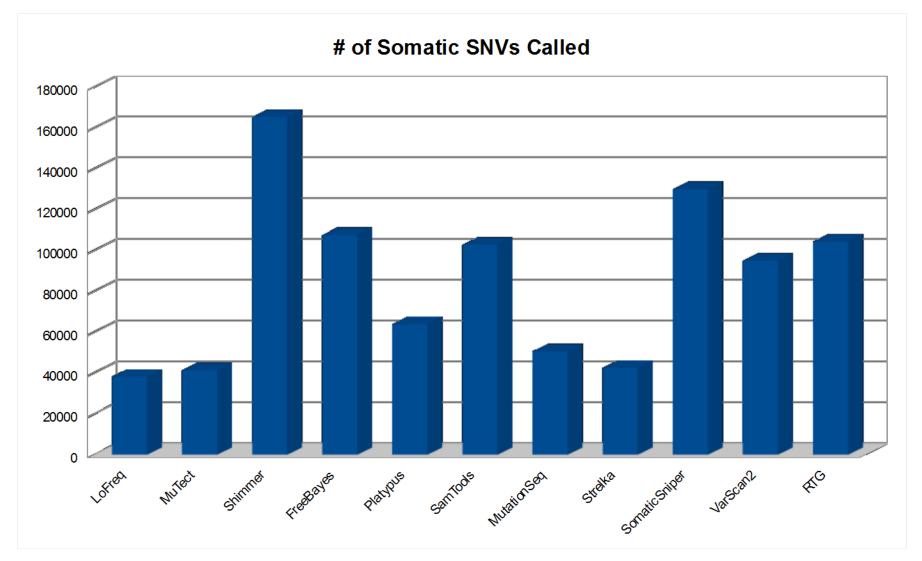
Split Chromosome Jobs

^{*} VarScan2 requires pileup files as input, pileup generation time not included

^{*} Single job was submitted for Strelka, but Strelka algorithm automatically splits jobs on chromosomes



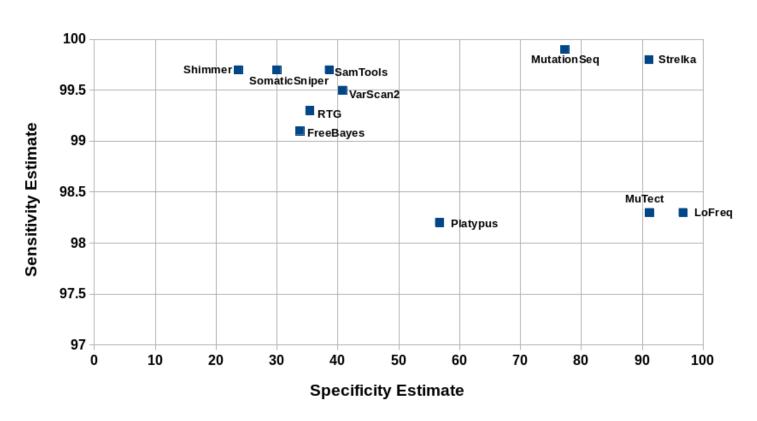








Sensitivity vs Specificity of Somatic SNV Callers



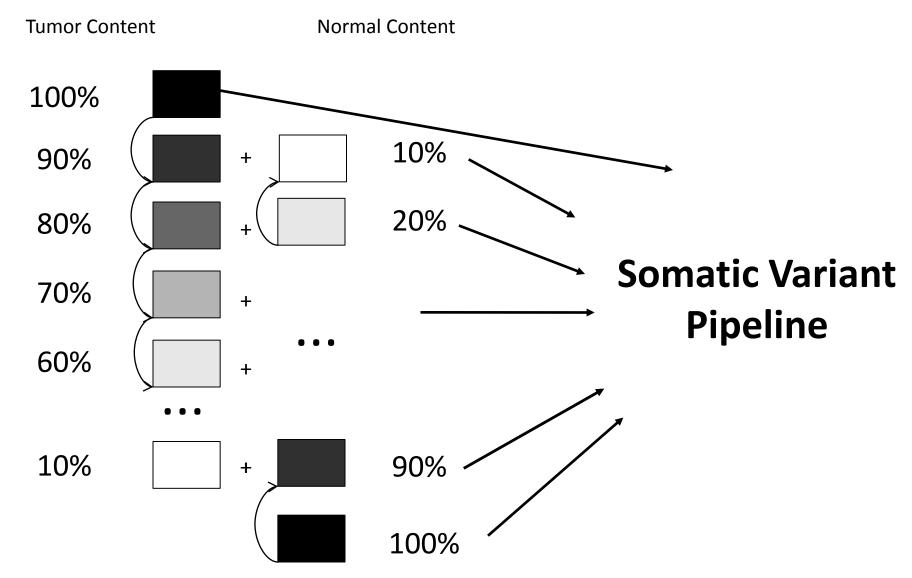




- How does a biopsy's tumour content impact the sensitivity/specificity of predicting somatic SNVs/indels?
 - Low tumour content → miss somatic SNVs/indels
- Performed a bioinformatics titration of our in-house COLO-829 tumour/normal sample
 - COLO-829 BAM vs COLO-829BL BAM

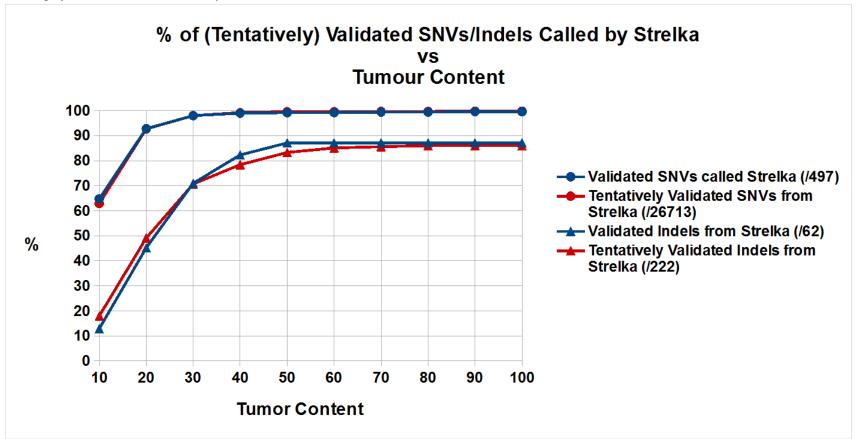












What is the minimum tumor content in which 95% of (tentatively) validated SNVs are called?

Tumor content of ~25%

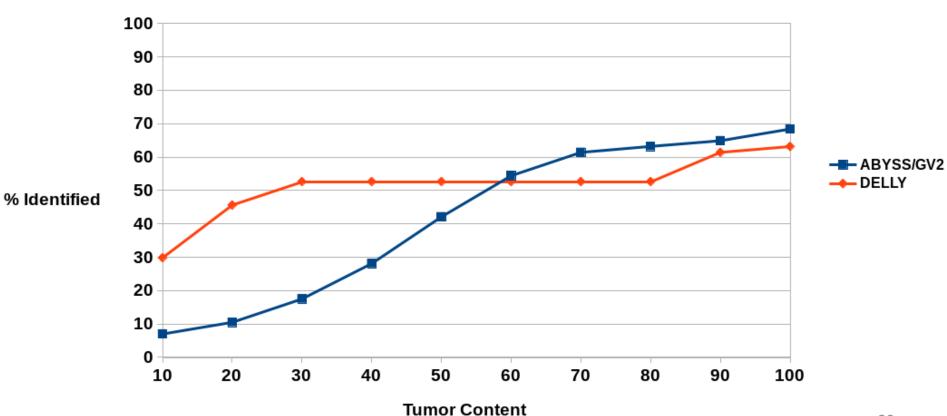
Given a sample with tumor content of 40%, what percent of (tentatively) validated SNVs are called? ≥ 99%





Structural Variant Sensitivity

(Translocations, Deletions, Duplications, Inversions)







Sci Rep. 2016 Apr 20;6:24607. doi: 10.1038/srep24607.

A somatic reference standard for cancer genome sequencing.

Craig DW¹, Nasser S¹, Corbett R², Chan SK², Murray L³, Legendre C¹, Tembe W¹, Adkins J¹, Kim N⁴, Wong S¹, Baker A¹, Enriquez D¹, Pond S⁴, Pleasance E², Mungall AJ², Moore RA², McDaniel T⁴, Ma Y², Jones SJ², Marra MA², Carpten JD¹, Liang WS¹.

Author information

Abstract

Large-scale multiplexed identification of somatic alterations in cancer has become feasible with next generation sequencing (NGS). However, calibration of NGS somatic analysis tools has been hampered by a lack of tumor/normal reference standards. We thus performed paired PCR-free whole genome sequencing of a matched metastatic melanoma cell line (COLO829) and normal across three lineages and across separate institutions, with independent library preparations, sequencing, and analysis. We generated mean mapped coverages of 99X for COLO829 and 103X for the paired normal across three institutions. Results were combined with previously generated data allowing for comparison to a fourth lineage on earlier NGS technology. Aggregate variant detection led to the identification of consensus variants, including key events that represent hallmark mutation types including amplified BRAF V600E, a CDK2NA small deletion, a 12 kb PTEN deletion, and a dinucleotide TERT promoter substitution. Overall, common events include >35,000 point mutations, 446 small insertion/deletions, and >6,000 genes affected by copy number changes. We present this reference to the community as an initial standard for enabling quantitative evaluation of somatic mutation pipelines across institutions.

PMID: 27094764 PMCID: PMC4837349 DOI: 10.1038/srep24607





Tool Evaluations

- Strategy to evaluate candidate tools for inclusion in production pipelines:
 - Literature search
 - Local installation of tools
 - Construct ground truth set
 - Compare results with respect to evaluation criteria
 - Choose tool(s) to use



Key Tools in Production Pipelines



- Alignment
 - DNA: bwa-mem
 - RNA: JAGuar, switching to STAR
 - Bisulfite: Novoalign
- Single sample SNV / indel: samtools mpileup
- CNV: cnaseq
- LOH: APOLLOH
- SVs for genomes: ABySS / DELLY / Manta
- SVs for transcriptomes: Trans-ABySS / DeFuse (chimerascan is being evaluated)
- ChIP: FindPeaks (MACS2 is being evaluated)



Conclusions



- Constructed a ground truth set of somatic SNVs/indels in COLO-829.
- Estimated sensitivity and specificity of in-house somatic SNV/indel pipeline.
- Investigated somatic SNV callers and compared their run times, number of somatic SNVs called, and accuracies.
- Even at low tumour contents (~20%), a significant number of somatic SNVs/indels were predicted accurately.







BCGSC:

Marco Marra

Steven Jones

Yussanne Ma

Richard Moore

Andrew Mungall

Richard Corbett

Karen Mungall

Tina Wong

Erin Pleasance

TGEN:

John Carpten

Daniel Craig

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Sara Nassar

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Christophe Legendre

Jonathan Adkins

Shukmei Wong

Angela Baker

Daniel Enriquez

Illumina:

Lisa Murray

Stephanie Pond

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Extra Slides





- Sanger sequencing is an effective way to verify somatic mutations
 - Draw backs:
 - Laborious and \$ to validate all somatic mutations identified in an NGS experiment
 - Not suitable for low frequency variants

BC Cancer Agend Software H		Organization	Notes IENCES C E N O M E
LoFreq	2012	Genome Institute of Singapore	Bernoulli trial, assume each base is independent with sequnce error (quality score), Posisson- bionmial distribution
MuTect	2013	Broad	Bayesian classifier
Shimmer	2013	NHGRI/NIH	Fisher's exact test comparing ref/alt alleles in tumor/normal with multiple testing
FreeBayes	2012	Boston College	Bayesian statistics
Platypus	2014	The Wellcome Trust Centre for Human Genetics	Local assembly, haplotype-based, multi- sample variant caller using Bayesian statistics
SAMTools	2009	Sanger/Broad	Calculate genotype from Bayesian prior probability

BC Cancer Agend Sofstware An agency of the Provincial Health Services Auth	Year Published	Organization	GENOME SENCES CENTRE
SomaticSniper	2011	WUSTL	Build genotype likelihood model of MAQ, calculates probablitity of genotype differences
VarScan2	2012	WUSTL	Fisher's exact test comparing ref/alt alleles in tumor/normal
RTG Somatic	2015	Real Time Genomics, Inc	Bayesian statistics
Strelka	2012	Illumina, Inc	Bayesian statistics
MutationSeq	2011	BC Cancer Research Centre	Feature based classifiers





Sequencers at the Genome Sciences Centre

	Bases Per Second	# Machines	Total Bases / Sec.
HiSeq X	8,700,000	5	43.5 million
HiSeq 2500	3,100,000	4	12.4 million
NextSeq	1,300,000	2	2.6 million
MiSeq	50,000	3	150 thousand

~55 million bases per second





How much sequence is that?

- Human Genome: 3,000,000,000 bases (approx.)
- At the Genome Sciences Centre, we can sequence 1 human genome every:
 - 3 billion bases / 55 million bases per sec = 54.5 sec
- The first human genome draft sequence took roughly 10 years to sequence and assemble





How do we extract meaning from the sequence data?

2,000,000,000 reads per sample

150 bases per read

3,000,000,000 base reference genome

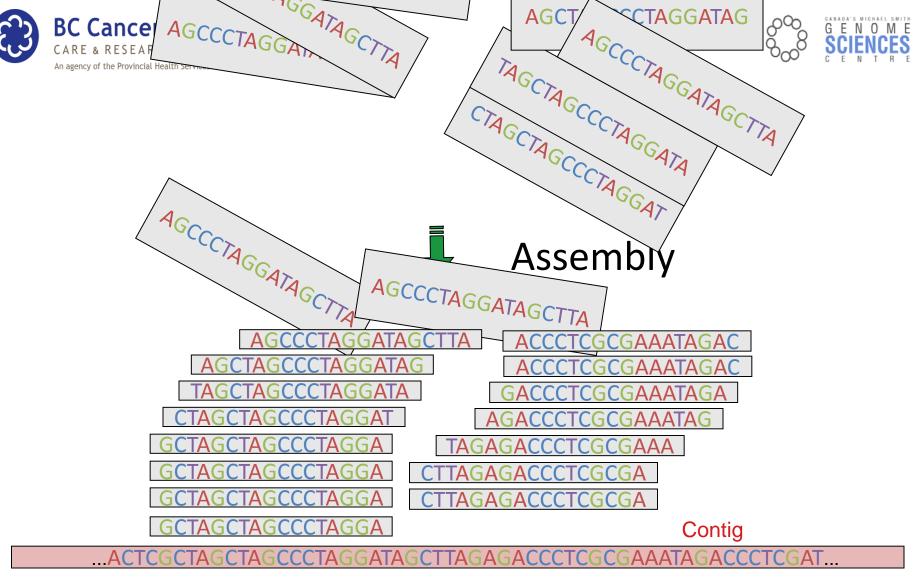




Data Interpretation

 For efficiency and to help interpretation, we often describe a sample by how it differs from a reference sample

- To compare samples, we
 - align sequence reads for a sample to a reference genome
 - find locations where our sample differs from the reference



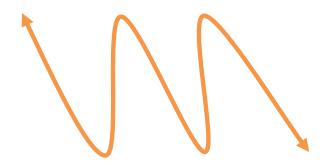


Reference Genome



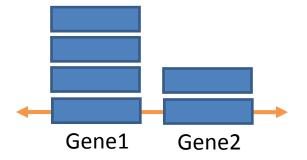
Genome and Transcriptome





Genome sequencing allow us to find:

- SNVs (single nucleotide variants)
 CCCTTTTGGGGAA
- CNVs (copy number variants)
- SVs (structural variants)



The transcriptome can be sequenced to find:

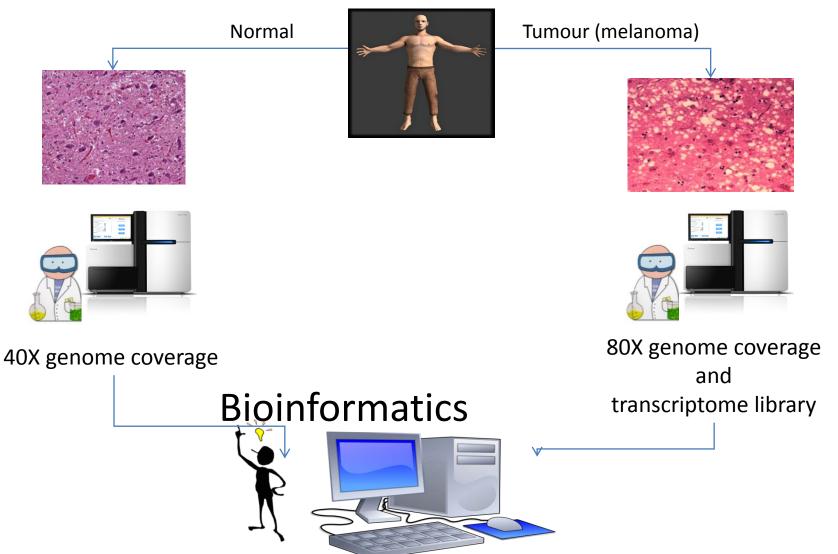
- Gene expression estimates
- Gene fusions

Gene1a Gene2b





Personalized Medicine

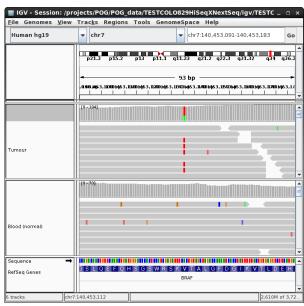




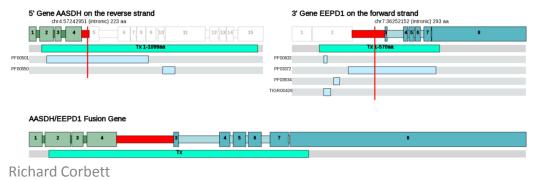
Personalized Medicine Intermediate Results



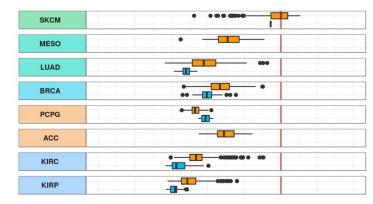
Somatic SNV calling



Gene fusion analysis



RNA expression Correlation



Somatic Copy Number

