

CNV Annotations: a crucial step in your variant analysis

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20 Most Promising Biotech Technology Providers



Hype Cycle for Life sciences



Top 10 Analytics Solution Providers

NIH Grant Funding Acknowledgments



- Research reported in this publication was supported by the National Institute Of General Medical Sciences of the National Institutes of Health under:
 - Award Number R43GM128485-01
 - Award Number R43GM128485-02
 - Award Number 2R44 GM125432-01
 - Award Number 2R44 GM125432-02
 - Montana SMIR/STTR Matching Funds Program Grant Agreement Number 19-51-RCSBIR-005
- PI is Dr. Andreas Scherer, CEO Golden Helix.
- The content is solely the responsibility of the authors and does not necessarily represent the official views of the National Institutes of Health.

Who Are We?



Golden Helix is a global bioinformatics company founded in 1998



Filtering and Annotation

ACMG Guidelines

Clinical Reports

CNV Analysis

Pipeline: Run Workflows



Variant Warehouse

Centralized Annotations

Hosted Reports

Sharing and Integration



CNV Analysis

GWAS | Genomic Prediction

Large-N Population Studies

RNA-Seq

Large-N CNV-Analysis

Cited in 1,000s of Peer-Reviewed Publications





















Over 400 Customers Globally



When you choose Golden Helix, you receive more than just the software





SOFTWARE IS VETTED

- o 20,000+ users at 400+ organizations
- o Quality & feedback



DEEPLY ENGRAINED IN SCIENTIFIC COMMUNITY

- o Give back to the community
- o Contribute content and support



SIMPLE, SUBSCRIPTION-BASED BUSINESS MODEL

- o Yearly fee
- o Unlimited training & support



INNOVATIVE SOFTWARE SOLUTIONS

o Cited in 1,000s of publications

	GENE PANEL E	XOME GENOME		Gol
	SEQ	UENCER		
PRODUCTS	BIOINFORM	ATICS PIPELINE	FUNCTION	
 DNASEQ (Sentieon) TNSEQ (Sentieon) VS-CNV 	F	ASTQ BAM VCF	 Single nucleotide variation Copy number variation & loss of hetero Chromosomal aberration 	zygosity
Annotations	Anno	tated VCF	Public & commercial annotations to enr genomic data sets	ich
VarSeqVSReportsVSPipeline	Clinic	al Report	 Annotate & filter Visually inspect alignments Variant prioritization Clinical assessment 	
VSClinical	Automated /	ACMG Guidelines	Clinical variant interpretation in coordir with ACMG Guidelines & AMP Guidelin	ation es
VSWarehouse	Data W Web-Ena + Powerful A TSV, CS	/arehousing bled Interface APQ: JSON, XML, V, SQL, FHIR	 Clinical assessment catalog Advanced data querying Versioning Interoperability Compliance with HIPPA, CLIA & CAP data discovery 	



CNVs in Clinical Testing

- Critical evidence needed for many genetic tests
- Common driver specific cancers, causal hereditary variation
 - EGFR Exon 19 deletion common in lung cancer
 - PIK3CA Amplification in breast cancer
- Large events used heavily in diagnostics
 - Chromosome 13 deletion common in melanoma
 - Autism Spectrum Disorder (ASD)
 - Developmental Delay (DD)
 - Intellectual Delay (ID)





Power of NGS CNV Detection



	Detectable events			Supported Data types			
	Small: 150b+	Medium: 1–10Kb	Large: 10Kb+	Gene panel	Whole exome	Whole genome	
MLPA	\checkmark			\checkmark			
CMA			\checkmark			\checkmark	
VS-CNV	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	

- ✓ One single testing paradigm
- True simplification of clinical workflow
- ✓ Saves time and money all on site

Addressing Issues - CNV Detection via NGS



- CNVs detected from coverage data in BAM
- Challenges
 - Coverage varies between samples
 - Coverage fluctuates between targets
 - *Systematic biases impact coverage
- Solutions
 - Data Normalization
 - Reference Sample Comparison
 - Algorithm works without case/control data
- Requirements
 - ≥ 30 ref samples
 - From same library prep method
 - Ideally ≥100X coverage



Principle Approach to CNV Calling





CNV Detection: Ratio, Z-score, and VAF



Metrics

- Ratio: sample coverage divided by reference sample mean
- Z-score: standard deviations from reference sample mean
- VAF: Variant Allele Frequency
- For Gene Panels and Exomes
 - Probabilistic model used to call CNVs
 - Segmentation identifies large cytogenetic events
- For Whole Genome Data
 - Targets segmented using Z-scores
 - Events called based on Z-score and Ratio thresholds



Optimizing CNV Detection - Segmentation



- Metrics are noisy over large regions
- Outliers cause large events to be called as many small events
- Solved using segmentation:
 - Regions containing many events are segmented
 - Small events sharing a segmented region are merged



VAF Provides Supporting Evidence



- Values other than 0 or 1 are evidence against het deletions
- Values of 2/3 and 1/3 are evidence for duplications



Advanced Optimization for CNV Detection - LOH



- Issue Large chromosomal deletions and duplications can skew the mean coverage of a sample
 - The previous approaches don't account for this
- Solution Detection of LOH areas to optimize definition of normal regions
 - Prior to running CNV caller
 - Probabilistic model based on VAF (Whole Exome/Genome)
 - Identifies and excludes non-diploid regions from normalization
 - Quality control step to improve CNV caller



CNV Confidence: P-Values



P-Values

- Introspective ability to define confidence in the CNV event
- Confidence threshold definable in your workflow/protocol
- You can modify the threshold at any point

 \checkmark

• Lower the p-value/probability the higher the confidence the event is real

Typical p-values

- <0.05*
- <0.01**
- <0.001***

				CNV Info				
		Region	Туре	# Targets	# Samples	Span	CNV State	^ p-value
		11:108160310	Loss	21	1	39676	Het Deletion	4.68818626586653e-07
		11:108128189	Gain	9	1	22167	Duplicate	8.93228134373203e-05
✓ p-value (Current) < 0.01	ų –	17:29683459	Gain	6	1	4283	Duplicate	0.00131093873642385
	- kdu	8:90965453-9	Gain	5	1	17353	Duplicate	0.0050437287427485
0.01	+	3:37053483-3	Gain	5	1	14036	Duplicate	0.00561221409589052
	-	17:56769986	Gain	4	1	10725	Duplicate	0.00574351288378239
Less than 0.01	2	11:64573088	Gain	4	1	1624	Duplicate	0.0100980764254928
Equal to 0.01	0	17:41219606	Gain	4	1	9046	Duplicate	0.0126859014853835
Greater than 0.01	0	17:41201106	Gain	4	1	14305	Duplicate	0.0156013108789921
Missing	0	11:108151711	Loss	4	1	6752	Het Deletion	0.0184763930737972
linooning	Ŭ	11:108121409	Gain	3	1	2251	Duplicate	0.0187129583209753
	2	3:37045873-3	Loss	4	1	7501	Het Deletion	0.0200324188917875

CNV Workflow and Annotations



- Workflow Issue: How to screen through many CNV events?
 - Especially relevant with WES
 - What steps are necessary to find events relevant to the patient?
- First step
 - Prioritize CNVs specific to sample
 - High quality
 - High confidence
- Second step
 - DGV CNVs Exclude CNVs in known healthy individuals
 - Genomic Sup Dups Exclude CNVs in known duplication regions
 - ExAC + 1kG Phase3 Eliminate common CNVs
 - ClinGen + ClinVar Eliminate known benign CNVs
- Third step
 - Prioritize individual sample phenotype + gene list







Project Demonstration

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We're headed to ESHG 2019 in June!