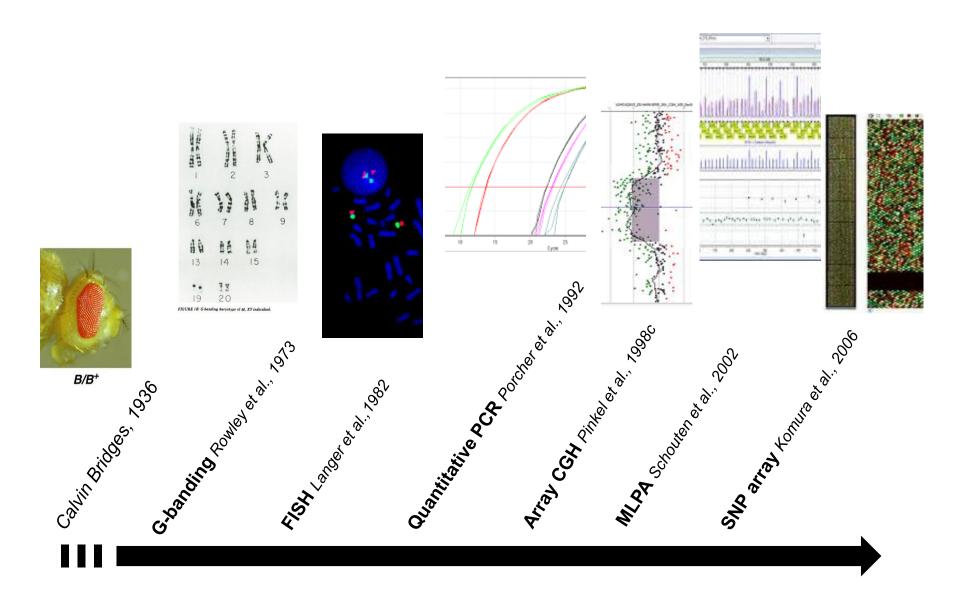
CNV detection

from targeted next-generation sequencing data

Anna Benet-Pagès

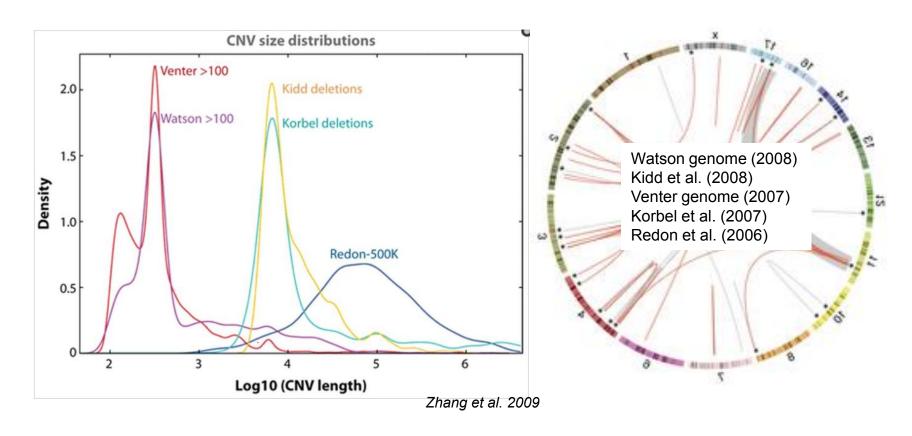
Johor 29.08.2018

Detection of structural variants and human disease



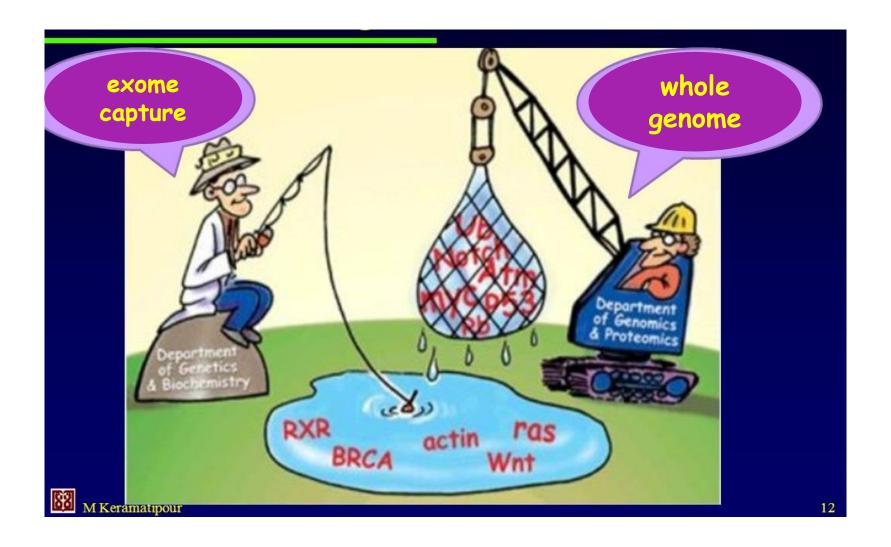
High resolution technologies reveal small-size CNVs

Size distribution of copy number variations (CNVs) larger than 100 bp



Smaller structural variants are the most frequent

On the meanwhile...



CNV detection from targeted-capture data

Challenges:

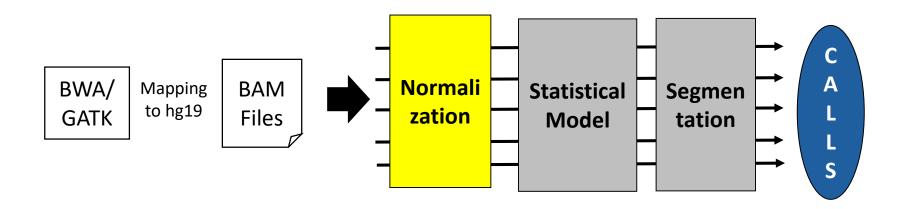
- CNV detection from exon capture approaches depends solely on read depth data
- Enrichment efficiency introduces a systematic noise in read depth data
- Coverage bias between sequencing runs and within samples of the same run
- Single exon events are extremely difficult to detect
- Control individuals are difficult to obtain (reference set / validation)
- Validation is expensive

CNV detection methods general considerations

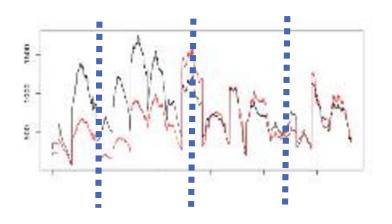
Which tool should I choose?

- applicable to capture data
- calling of rare CNVs
- easy to integrate (take bam files as input)
- easy handling (installation / running time)
- multi-sample usage (possibility to normalize against reference set)
- Tools should use different statistic models

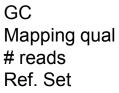
CNV Pipeline Structure

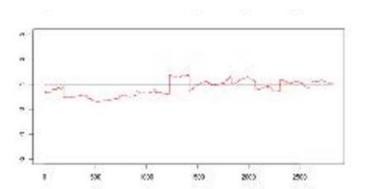


breakdown of the target region exons/ windows



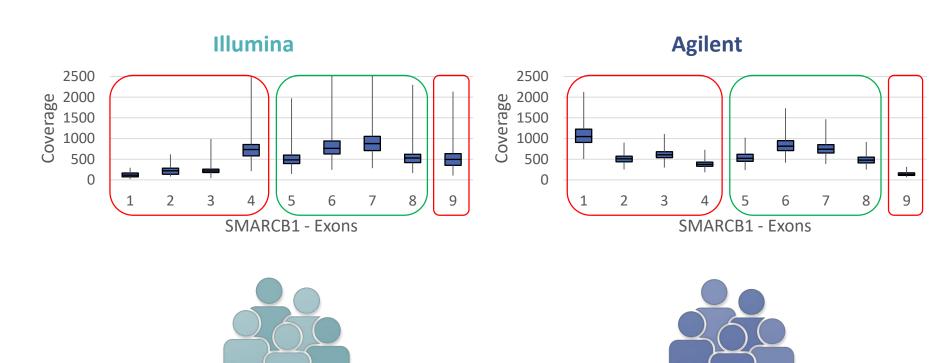




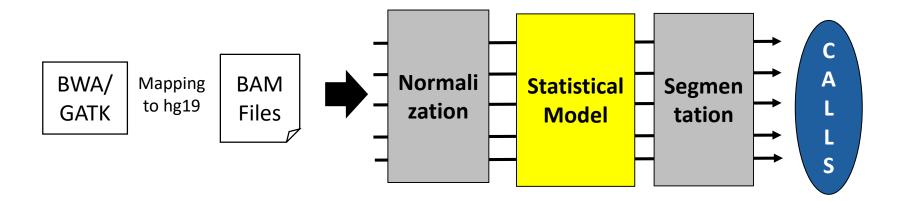


Reference Sets and data normalization

- different reference sets for different kits / enrichment methods
- > normalization against samples from the same sequencing run to improve robustness against workflow conditions



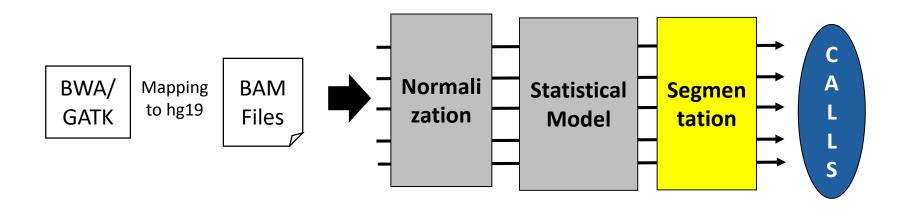
CNV Pipeline Structure



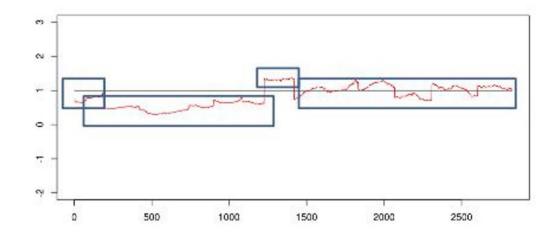
Poison distribution Beta binomial Negative binomial Normal distribution

frequency	<i>≥</i> . I
24	iii ₹i
95	- Pagle Pagle
82	80
21	O 1 2 3 Copy Number
	24 95 82

CNV Pipeline Structure



grouping areas (exon/window) with the same prediction (gain / loss / normal)



CNV detection methods

> Use a combination of several detection tools

AGE, BicSeq, BreakDancer, Breakpointer, Breakseq, Canoes, Clamms, Clever, ClipCrop, Cn.MOPS, CNAnorm, CNAseg, CND, CNV_TV, Cnvator, CNVer, CNVer, HugeSEQ, hydra, inGAP_sv, JointS ,,meta-CNV-caller" rcanavar, Patchwork, pemer ,ReadDepth, SW_seq, segseq, seqcbs, CNVer, cnvHiTSeq, cnvrd, CNV-seq, conserting, CONTROL_FREEC, cops, copySeq, crest, ERDS, codex EWT_RDXplorer, GasvPRO, GENSENG, XHMM

Meta-Tool CNV Detection Pipeline

ExomeDepth

extremely sensitive and robust against samples that do not correlate with the reference

Canoes

has a high sensitivity for small deletions, high performance in low coverage regions and with few reference samples

Clamms

corrects for GC content and mappability, divides large exons into smaller regions and calls also common CNVs

Codex

corrects for GC content and mappability, calls also common CNVs, uses no HMM for segmentation (all other tools use HMMs)

Inhouse method

is well adapted on inhouse data, screens for heterozygosity, corrects for GC content, exon score depends on previous analyses

Performance of single tools

> Training set: true set of 146 CNV calls detected via MLPA.

	Exome Depth	Clamms	Canoes	Codex	In-house
Precision	45.63%	68.57%	96.77%	64.75%	40.82%
Sensitivity	90.38%	46.15%	57.69%	63.46%	76.92%

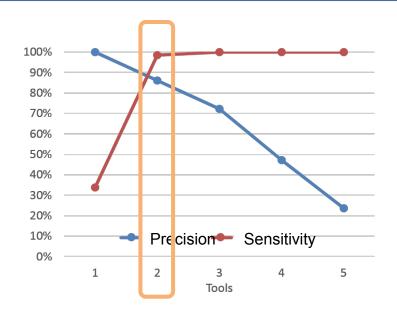
Performance of tool combinations

What is the best number of tools required to call a variant?

stringency	

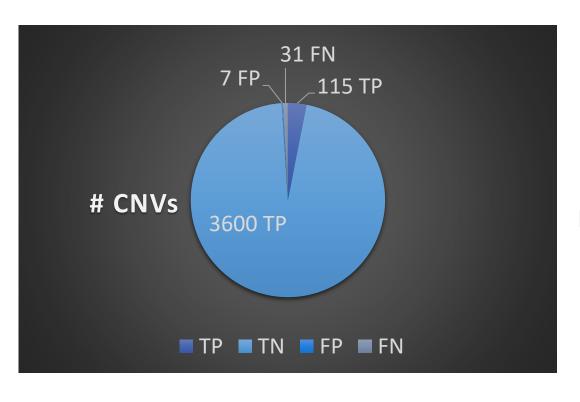
		TP	FN	FP	TN	Sensitivity (TPR)	Specificity (TNR)	Precision (PPV)	NPV
	2 out of 5	115	7	8	3600	94.26%	99.78%	93.50%	99.81%
	3 out of 5	72	50	1	3600	59.02%	99.97%	98.63%	98.63%
	4 out of 5	28	94	1	3600	22.95%	99.97% 96.55%		97.46%
•	5 out of 5	4	118	0	3600	3.28%	100.00%	100.00%	96.83%

Using two out of five concordant tool predictions shows the best balance between sensitivity and specificity



CNV Pipeline Evaluation

- >3700 MLPAs were performed in ~90 genes
- 146 CNVs (85 deletions / 61 gains)



Sensitivity: 88.60%

Specificity: 98.88%

Precision: 71.40%

Pseudogenes excluded:

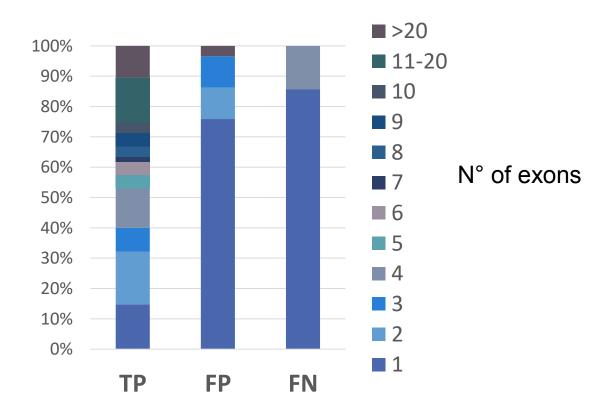
Sensitivity: 94.26%

Specificity: 99.78%

Precision: 93.50%

TP, FP, FN versus CNV size

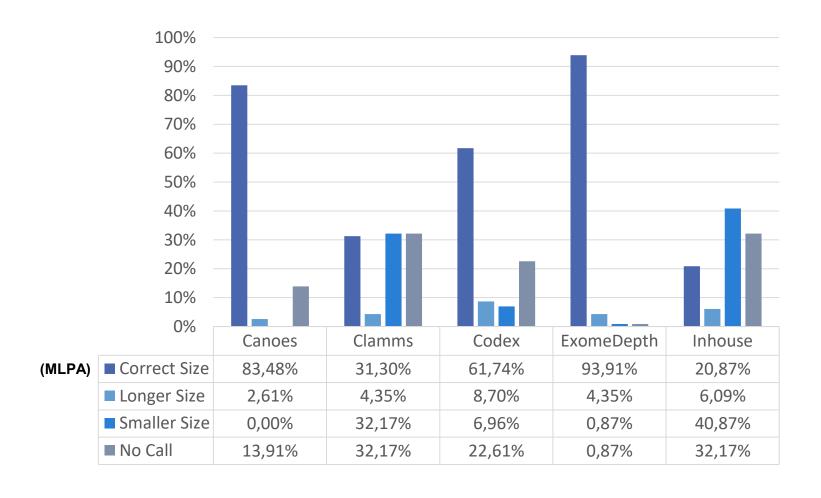
- Comparison of CNV sizes of TP, FP and FN calls detected by the combined CNV pipeline on the validation set.
- FP mainly cosiand FN calls consist mainly of single exon events.



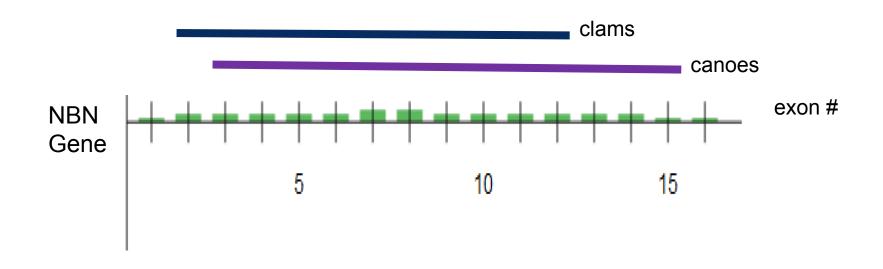
definition of special quality thresholds for single exon events to minimize false negatives

Discrepancies in CNV size detection between tools

- ➤ Size of CNV calls were compared to MLPA
- Size is given as the number of exons within the call



Meta-CNV-caller: multi calls for one event



Copy Number Variants in NBN

Calls 2

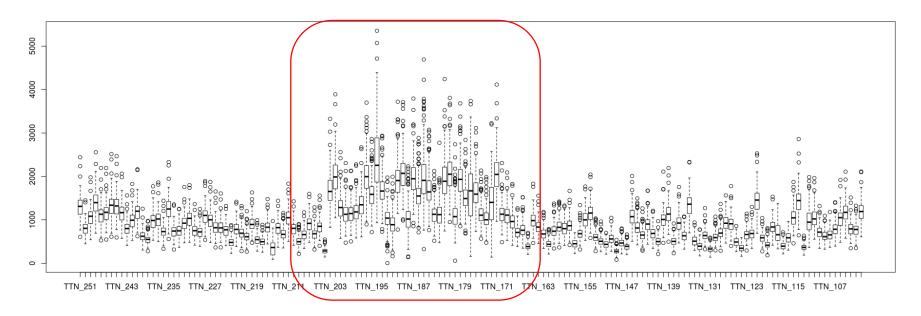
	Exons	Туре	CN	Sample	Pool	Region	PPL	•	Overlap (equal)		•	Methods
1	E2 – E12	+	3	121258	SP-666	chr8: 90,955,481 - 90,996,789		1		9		clamms CN3, exomedepth CN3
2	E3 – E15	+	3	121258	SP-666	chr8: 90,947,810 - 90,995,083		1		9		canoes CN3, exomedepth

Challenges

- Non-uniform of coverage
- CNV calling in homologous genomic regions (pseudogenes...)
- Clinical interpretation

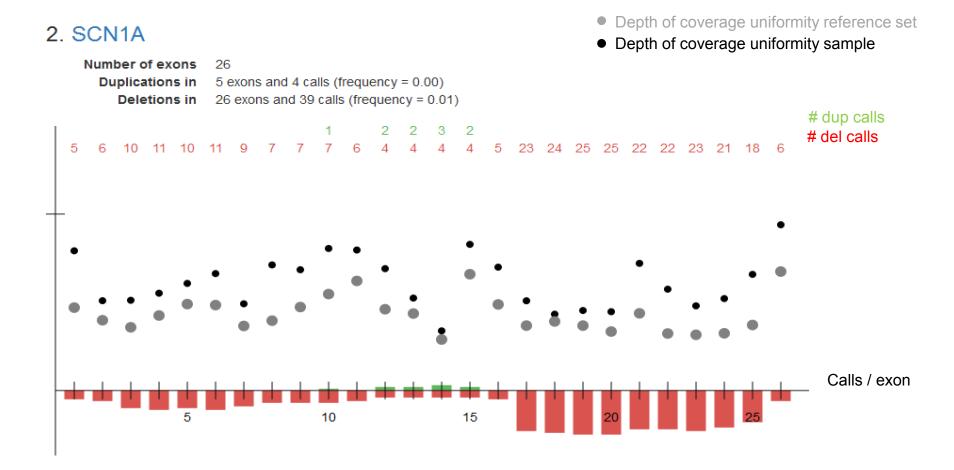
non-uniform coverage = capture bias

identification of reliable regions by assessment of capture efficiency to minimize false positives

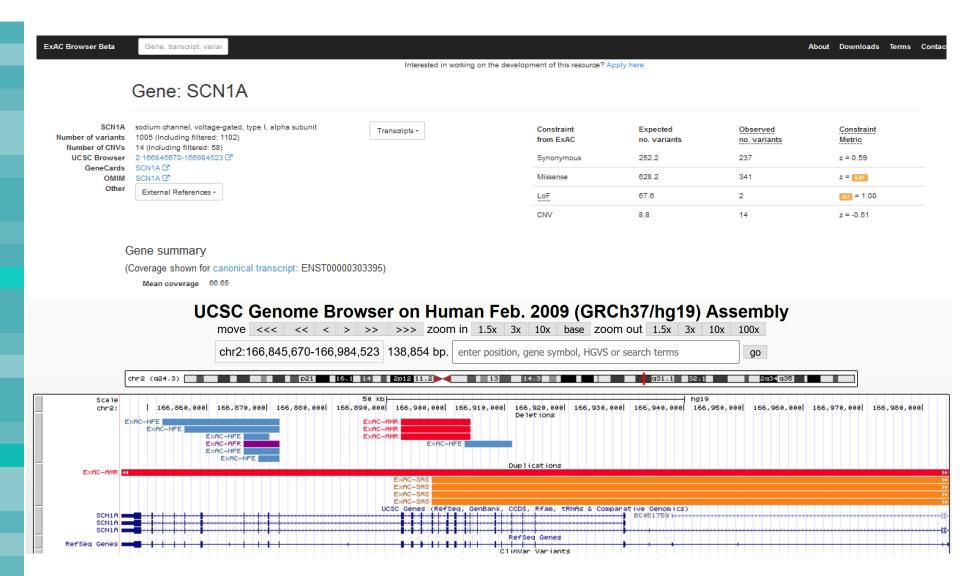


can not be analyzed

CNV calling artifacts



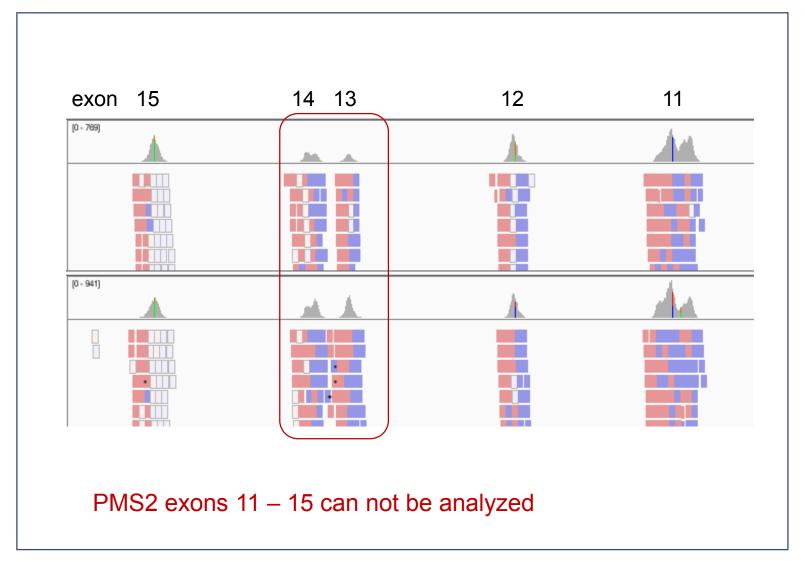
CNV calling artifacts



CNV calling in Pseudogenes

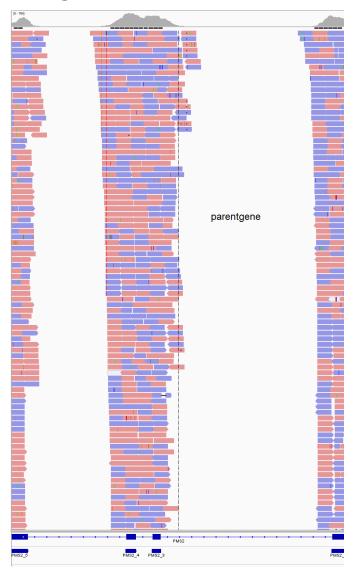
Forward read / unique mapping
Reverse read / unique mapping
Non-unique mapping

1. PMS2

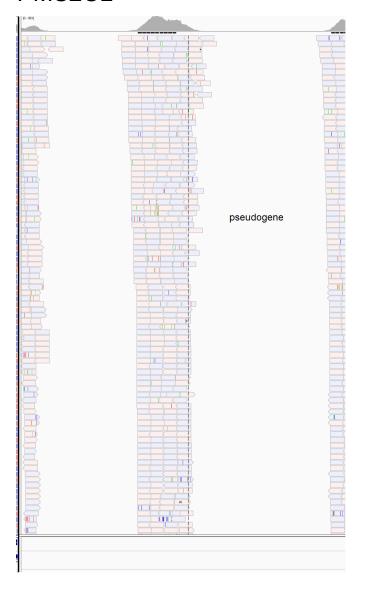


CNV calling in Pseudogenes

PMS2

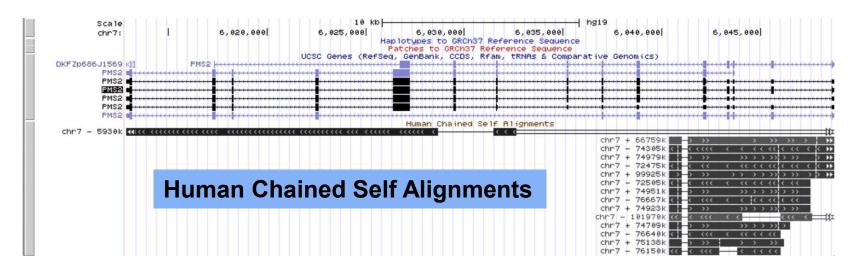


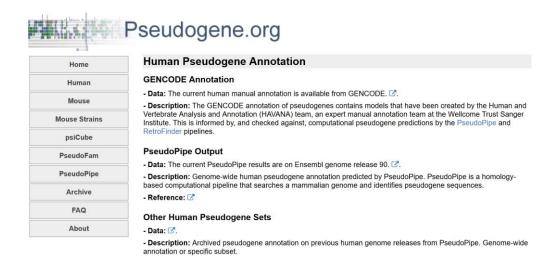
PMS2CL



How to identify regions affected by pseudogenes

alignments of the human genome with itself using blastz





Interpretation of CNV calls – population DB

- Deletions and duplications called based on read depth using XHMM; Fromer et al.
- Z score for the deviation of observed counts from the expected number



Positive Z scores indicate that the gene had fewer variants than expected. Negative Z scores are given to genes that had a more variants than expected.

Interpretation of CNV calls – clinical DB



ClinGen Dosage Sensitivity Curation Page

> DGV

DECIPHER

ClinVar

ClinGen

FBN1

Curation Status: Complete

id: ISCA-30689

Date last evaluated: 2014-06-04 Issue Type: ClinGen Gene Curation

Gene type: protein-coding

Entrez Gene: https://www.ncbi.nlm.nih.gov/gene/2200

OMIM: https://omim.org/entry/134797

Gene Reviews: https://www.ncbi.nlm.nih.gov/books/NBK1116/?term=FBN1%

5Bgenesymbol%5D

ClinGen Haploinsufficiency Score: 3 ClinGen Triplosensitivity Score: 0

ExAC pLI score: 1.0

Location Information

15a21.1

GRCh37/hg19 chr15: 48,700,503-48,937,985

View: NCBI | Ensembl | UCSC

GRCh38/hg38 chr15: 48,408,306-48,645,788

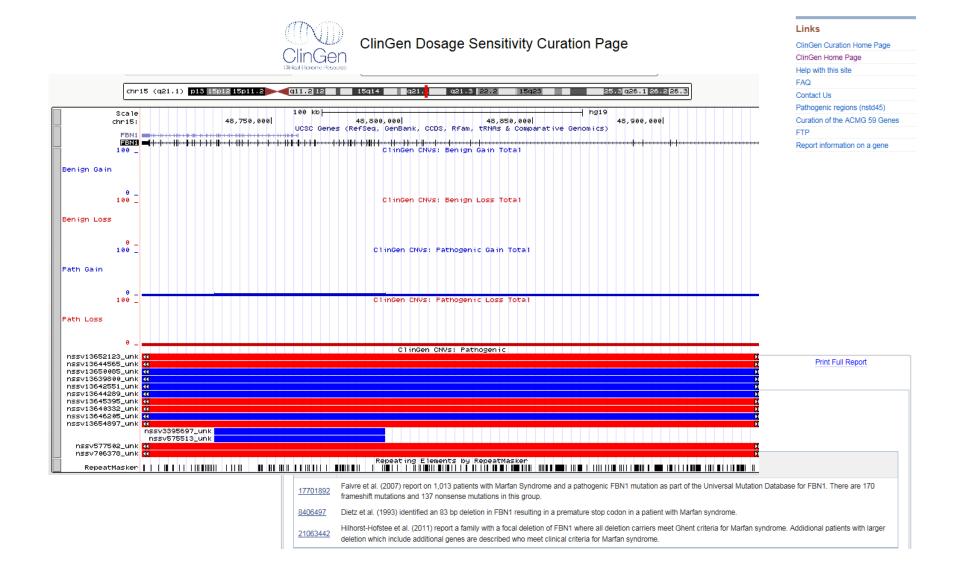
View: NCBI | Ensembl | UCSC

Links ClinGen Curation Home Page ClinGen Home Page Help with this site FAQ Contact Us Pathogenic regions (nstd45) Curation of the ACMG 59 Genes

Report information on a gene

Print Full Report Evidence for Haploinsufficiency Phenotypes Evidence for Triplosensitive Phenotypes Genome View Haploinsufficiency score: 3 Strength of Evidence (disclaimer): Sufficient evidence for dosage pathogenicity Haploinsufficiency Phenotype: MARFAN SYNDROME; MFS Evidence for haploinsufficiency phenotype PubMed Faivre et al. (2007) report on 1,013 patients with Marfan Syndrome and a pathogenic FBN1 mutation as part of the Universal Mutation Database for FBN1. There are 170 17701892 frameshift mutations and 137 nonsense mutations in this group. 8406497 Dietz et al. (1993) identified an 83 bp deletion in FBN1 resulting in a premature stop codon in a patient with Marfan syndrome. Hilhorst-Hofstee et al. (2011) report a family with a focal deletion of FBN1 where all deletion carriers meet Ghent criteria for Marfan syndrome. Addidional patients with larger 21063442 deletion which include additional genes are described who meet clinical criteria for Marfan syndrome.

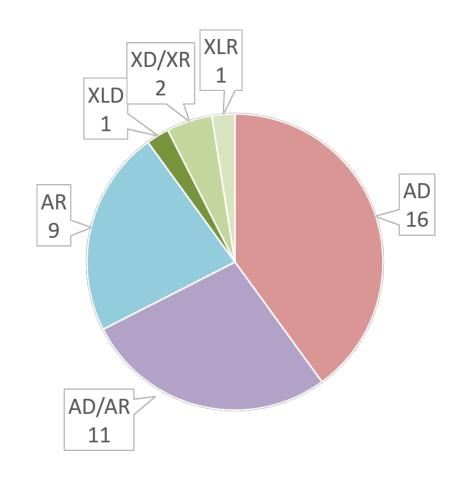
Interpretation of CNV calls – clinical DB



CNV analysis on 1600 individuals within the routine Dx

CNV clarified the underlying phenotype in 8 % of the cases

Increase of the diagnostic yield in 3%



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