MB-GATK-SGE pipeline

Classic Unified Genotyper workflow Common per-sample processing New Haplotype Caller workflow BaseRecal.sh End of sample level Per sample level workflow BAM files merged using Picard PrintReads.sh Q scores for each base are threading used to off-load workflow START recalibrated using machine [de]compression/IO. shell script Merge BAM files Base Q Score Recal learning. Two stages i) build Input is Raw .fastq.gz takes path/*.bam as input from **Raw Reads** ideally reads should not Merae BAM BOSR sample lvl model ii) apply it and "print" a new command line FAST0 be trimmed set of reads Reads are realigned around Variants called in new genomic BWA.sh HC.sh Haplotype Caller per VCF mode at sample level, this is 2nd realignment Alian Reads BWA gives SAM output. indels, two stages: -R used to encode read i) Realignment Target Creation, auick. GATK 3.x uses AVX **BWA MEM** of all reads Merged RTC.sh sample groups in SAM header accelerated PairHMM on new ii) Indel Realignment Realn merged_IDR.sh HC sample lvl **CPUs** SamToSortedBam.sh Q scores for each base are Base Q Score Recal SAM to sorted SAM converted to End of sample level workflow recalibrated using machine sorted index BAM in BQSR_merged index BAM GenotypeGVCFs.sh learning. Two stages i) build one step using Merged_BaseRecal.sh SamToSortedBam model ii) apply it and "print" a Picard: SortSam Individual gVCF files are fused Meraed PrintReads.sh Genotype and fuse new set of reads here to make a single VCF, gVFCs to a single VCF downstream analysis is as before GenotypeGVCFs MarkDuplicates.sh Find PCR Duplicates marked not Variants called on all samples removed, using Picard: duplicates Unified Genotyper UG_snps.sh simultaneously, using Unified MarkDuplicates, this is MarkDuplicates UG_indels.sh Genotyper, calls SNPs and UG_merged more effective at finding indels separately owing to size duplicates than SAMtools VOSR_snps_HC.sh Gaussian mixture model trained of unified dataset. VOSE_indels_HC.sh using 1000G, HapMap, dbSNP and Omni array data. Gaussian mixture model Varient Q Score Variant Q Score Sample level Recalibrated variants are filtered trained using 1000G, HapMap, Recalibration Recalibration realignment at a desired truth level, SNPs and dbSNP and Omni array data. VOSR_UG VQSR_snps_UG.sh VOSR_HC indels should not undergo VQSR 1stRealn Recalibrated variants are VOSE_indels_UG.sh RTC.sh Reads are realigned around indels, two stages: toaether filtered at a desired truth level. ApplyRecalibration_snps_HC.sh IDR.sh i) Realignment Target Creation, SNPs and indels should not ApplyRecalibration_snps_UG.sh ApplyRecalibration_indels_HC.sh ii) Indel Realignment undergo VQSR together ApplyRecalibration_indels_UG.sh Somatic variant calling Varient filtering stage Recalibrated variant filtering Recalibrated variants can be filtered for those MuTect and MuTect2 somatic variant calling SelectRecaledVariants_snps.sh passing recalibration via the PASS flag. This can SelectRecaledVariants_indels.sh be further filtered via the VQSlod log odds ratio which is the likelihood of being a true variant Select variants: versus being false under the trained Gaussian passing recalibration, MT.sh mixture model. VQSlod >= 0 tends to be a better Call tumor / normal pairs MuTect subtracts the normal VQSlod >= 0.MT2.sh subset with >= 3 being even better in terms of (germline) variants from the tumor using MuTect (somatic) variants. MuTect2 can VQSlod >= 3variant quality MuTect call somatic indels and SNPs, Filt_Recaled_VCF MuTect2 MuTect1 only calls SNPs

MuTect handles heterogeneous and impure tumour samples.

MuTect jobs are submitted from a list of paired normal/tumour sample read groups in the automated pipe-line

HardFilt_both_HC.sh HardFilt_indels_UG.sh

HardFilt_snps_UG.sh

Should recalibration fail (owing to lack of bad variants) then hard (i.e. preset) filters can be applied to both the SNPs and indels