



**Figure 2: Workflow Solution for Somatic Variant Detection**—\*Tumor-Normal WES subtraction analysis to identify somatic variation requires third-party software (ie, VarScan; [www.varscan.sourceforge.net](http://www.varscan.sourceforge.net)) as this functionality is not currently supported by BaseSpace Core Apps.

## A Targeted Approach With Whole-Exome Sequencing

Many cancer-causing variants have been found in exonic (protein-coding) regions. Whole-exome sequencing (WES) enables researchers to target cost-effectively 1–2% of the genome that codes for proteins while increasing sensitivity (with high read depth) for somatic variant detection in tumor and normal samples. For a given budget, compared to WGS, the higher read depth of WES enables the detection of low allele fraction variants for accurate identification of somatic mutations within heterogeneous tumor samples. The Nextera® Rapid Capture Exome kits are all-in-one methods for library preparation, enabling rapid identification of coding variants up to 70% faster than other methods (Figure 2). With low DNA input of 50 ng per sample, these kits maintain high coverage uniformity, enrichment rates, and on-target specificity, providing comprehensive coverage of the exome.

## Complete Transcriptome Analysis With RNA Sequencing

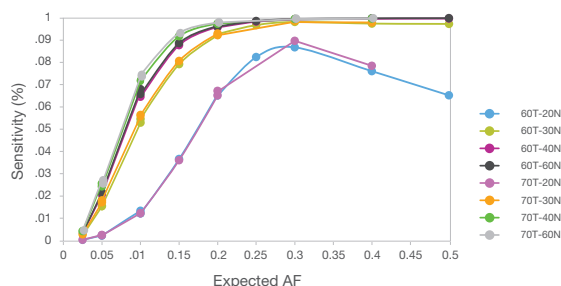
RNA sequencing enables researchers to understand the functional effects of DNA mutations through the highly sensitive detection of changes in transcript abundance, discovery of gene fusions, and novel transcript isomers. Further, a broad suite of available library preparation kits addresses the full range of experimental needs, including robust, cost-efficient preparation of low-quality (FFPE) samples and limited available input amounts. Illumina TruSeq RNA Library Prep Kits enable cost-effective, accurate, and complete sequencing of the transcriptome with high-coverage and uniformity. The simple and flexible workflows support quick and robust interrogation of standard (0.1–4 µg), low-input (20 ng), and FFPE samples (Figure 2). These kits provide a comprehensive view of the transcriptome to elicit the functional effects of somatic variants associated with tumorigenesis and progression.

For more details on Illumina library preparation kits for optimal somatic variant detection, see Table 1.

## High-Accuracy Somatic Variant Sequencing

Because of tumor heterogeneity and normal cell contamination, sequencing of a tumor sample often identifies a mix of DNA signatures representing the constituent cell types of that particular tumor sample. Thus, detection of true somatic mutations requires deep sequencing to detect driver mutations at low allelic fractions within the sample. Deep sequencing refers to sequencing a genome region multiple times with overlapping short reads to obtain high confidence that identified mutations within that region are accurately identified. Figure 3 shows the impact of read depth on the ability to call somatic variants of various allelic fractions. The required read depth depends on several factors including the purity (ie, normal cell contamination) and heterogeneity (ie, clonal architecture) of the tumor sample and desired experimental sensitivity. Thus, any increase in raw sequencing error will limit the ability to detect true variants within these heterogeneous samples. This is important for somatic variant detection as some somatic driver events are present at low allelic level within a given sample. Illumina NGS technology not only delivers the highest-quality raw data on a genome-wide level, but allows access to more of the genome, successfully sequencing difficult regions such as GC-rich and homopolymers. As a result, significantly more variants can be robustly and accurately identified.

Table 2 provides guidelines for ideal sequencing target depths for optimal somatic variant detection, providing sufficient confidence to identify mutations within tumor samples while limiting false positive and negative rates. Leveraging Illumina technology, the HiSeq® and NextSeq® Systems deliver industry-leading accuracy (> 80% sequenced bases over Q30) and high coverage capability, addressing the challenges and increasing sensitivity in detecting somatic variants.



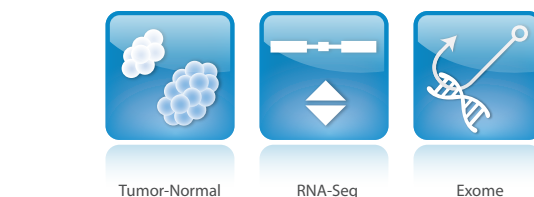
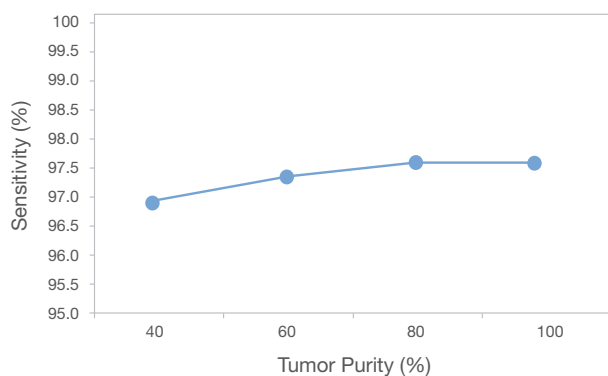
**Figure 3: Impact of Read Depth on Sensitivity for Somatic Variant Detection**—With increased read depth (coverage), the sensitivity (ability to call variants) for detecting somatic variants at low allelic levels increases for tumor-normal data sets.

## Fast and Accurate Bioinformatics Tools

### Tumor-Normal for WGS Analysis

Tumor biopsies are often highly heterogeneous and are likely to contain multiple subclones of cancer cells, as well as normal cell contamination. The Illumina Bayesian combined-calling method for somatic variant detection is designed to address this intrinsic heterogeneity.

Sequencing reads from both normal and tumor samples are mapped to the reference sequence using Illumina Isaac Alignment<sup>7</sup> software. The somatic variant caller combines tumor and normal sample data sets together (combined calling) to model varying levels of tumor purities present with superior accuracy and sensitivity compared to other subtractive analysis methods.<sup>8</sup> The combined analysis takes into account the somatic variation and noise that can occur at any allele frequency ratio, and reports somatic SNVs and small indels. Similarly, structural variants (SVs) are obtained by comparing the SVs identified in the tumor and matched normal samples. The combined-calling method from Illumina identifies somatic and true variants with superior sensitivity > 97.0% and specificity > 98.5% for tumor purity as low as 40% (Figure 5).

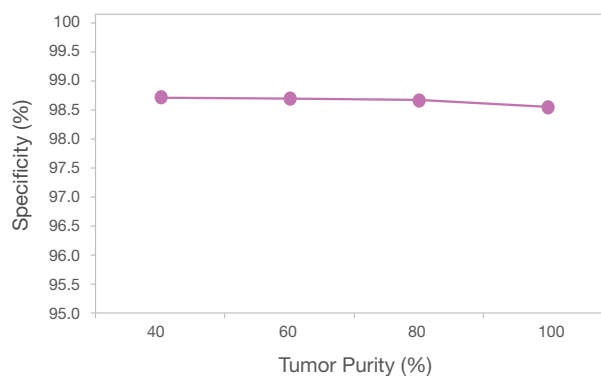


**Figure 4: BaseSpace Apps Empower Somatic Variant Discovery**—A suite of BaseSpace Apps is available to detect somatic variants and perform transcriptome analysis in tumors. Somatic variant detection from whole-genome tumor-normal sequence data can be achieved using the Tumor-Normal App. The RNA-Seq apps can be used to analyze gene expression, transcripts, and fusion genes, while the exome apps detect mutations. **Note:** Currently, tumor-normal subtraction is not enabled in BaseSpace for exome data. For somatic variant detection using subtractive analysis of tumor-normal samples after whole-exome sequencing, Illumina recommends use of a third-party software such as VarScan from Washington University ([www.varscan.sourceforge.net](http://www.varscan.sourceforge.net)).

The tumor-normal WGS workflow described here is available as part of a suite of cancer data analysis apps in the Illumina BaseSpace® computing environment (Figure 4). BaseSpace is available to users with access to an Illumina sequencing system or a third-party service provider. Raw data from Illumina sequencing instruments can be uploaded to and analyzed directly in BaseSpace. Analysis results for all apps, including annotated variants and performance statistics, are presented in intuitive, easy-to-interpret reports.

## RNA-Seq Analysis

New Illumina RNA-Seq Apps in BaseSpace (Figure 4) provide expert-preferred data analysis tools (TopHat<sup>9</sup> and Cufflinks<sup>10</sup>) packaged in an intuitive, click-and-go user interface designed for informatics novices. These apps deliver preconfigured workflows that support a range of common transcriptome data analysis needs, including high-confidence alignment for abundance measurement and detection of splice junctions, gene fusions, and cSNPs. Simple-to-follow prompts guide users through the entire process, starting with file selection generated by the sequencer to filtering and visualizing analyzed data and results. RNA-Seq Apps software generates output files that can be directly imported into a broad range of available downstream analysis tools.



**Figure 5: Combined-Calling Method Enhances the Ability to Call True Somatic Variants**—A melanoma cell line was analyzed using a combined-calling method. Tumor sample data was mixed with varying fractions of a normal sample data ranging from 40%–100% to simulate different tumor purities. Results indicate high sensitivity (ability to call variants) and specificity (ability to call true variants) even at low tumor purity.

For Research Use Only. Not for use in diagnostic procedures.

Application	Whole-Genome Sequencing		Whole-Exome Sequencing		RNA Sequencing	
Recommended Kit	TruSeq DNA PCR-Free	TruSeq Nano DNA	Nextera Rapid Capture Exome	Nextera Rapid Capture Expanded Exome	TruSeq Stranded Total RNA	TruSeq Stranded mRNA
Sample Input DNA/RNA (µg/sample)	1–2	0.1–0.2	0.5	0.5	0.1	0.1
Total Prep Time	5 hours	6 hours	30 hours for up to 96 samples	30 hours for up to 96 samples	8 hours	9 hours
Hands-On Prep Time	4 hours	5 hours	< 5 hours	< 5 hours	5.5 hours	5.5 hours
Kit Configurations (samples/kit)	24, 96	24, 96	24, 48, 96	24, 48, 96	48, 96	48, 96
Pre-Enrichment Sample Pooling	N/A	N/A	Up to 12	Up to 12	N/A	N/A
Recommended Read Length	2 × 100 bp	2 × 100 bp	2 × 75 bp – 2 × 100 bp	2 × 75 bp – 2 × 100 bp	2 × 75 bp	2 × 75 bp

Much progress has been made in the discovery of driver mutations in cancer, but recent reports suggest that many thousands of variants remain undiscovered.<sup>11</sup> Therefore, some groups are working to combine their basic research discovery efforts with clinical research by associating detected somatic variants with clinical phenotype data. Still others are collecting their NGS data sets in a CLIA-certified setting for both research purposes and clinical decision making.<sup>12</sup> While whole-genome and whole-exome approaches provide the most opportunity for discovery and detection, many groups have opted to focus only on a handful of confirmed driver mutations with established clinical significance. They have chosen to bypass discovery efforts in favor of a focus on clinical implementation. While this highly focused approach has some benefits, the emerging genetic complexity of cancer suggests significant advantages in investing in a translational data set that is complete. Not only will the new data set be available for research, but also for reanalysis as discoveries are made and the spectrum of available targeted therapies increases.<sup>13</sup>

Cancer is a diverse and deadly disease of the genome. Due to the molecular complexity, cancer has historically been a challenge to identify driver mutations and understand the biology of tumorigenesis and progression. However, by combining DNA sequencing from both the tumor sample and normal tissue with tumor RNA sequencing, researchers can obtain an optimal data set for somatic variant discovery. Furthermore, rapid and comprehensive library preparation and enrichment kits, industry-leading sequencing accuracy, and streamlined data analysis workflows from Illumina provide efficient and precise interrogation of the molecular complexity of cancer. These workflows now enable all researchers to identify a complete picture of somatic variation in tumor samples, which until now, have been accessible only to international consortiums such as TCGA and ICGC.

Application	Normal Sample	Tumor Sample
Whole-Genome Sequencing	40x	80x
Whole-Exome Sequencing	100x	130x
Total RNA Sequencing	N/A	> 50M reads
mRNA Sequencing	N/A	25M reads

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