

SINGULAR GENOMICS

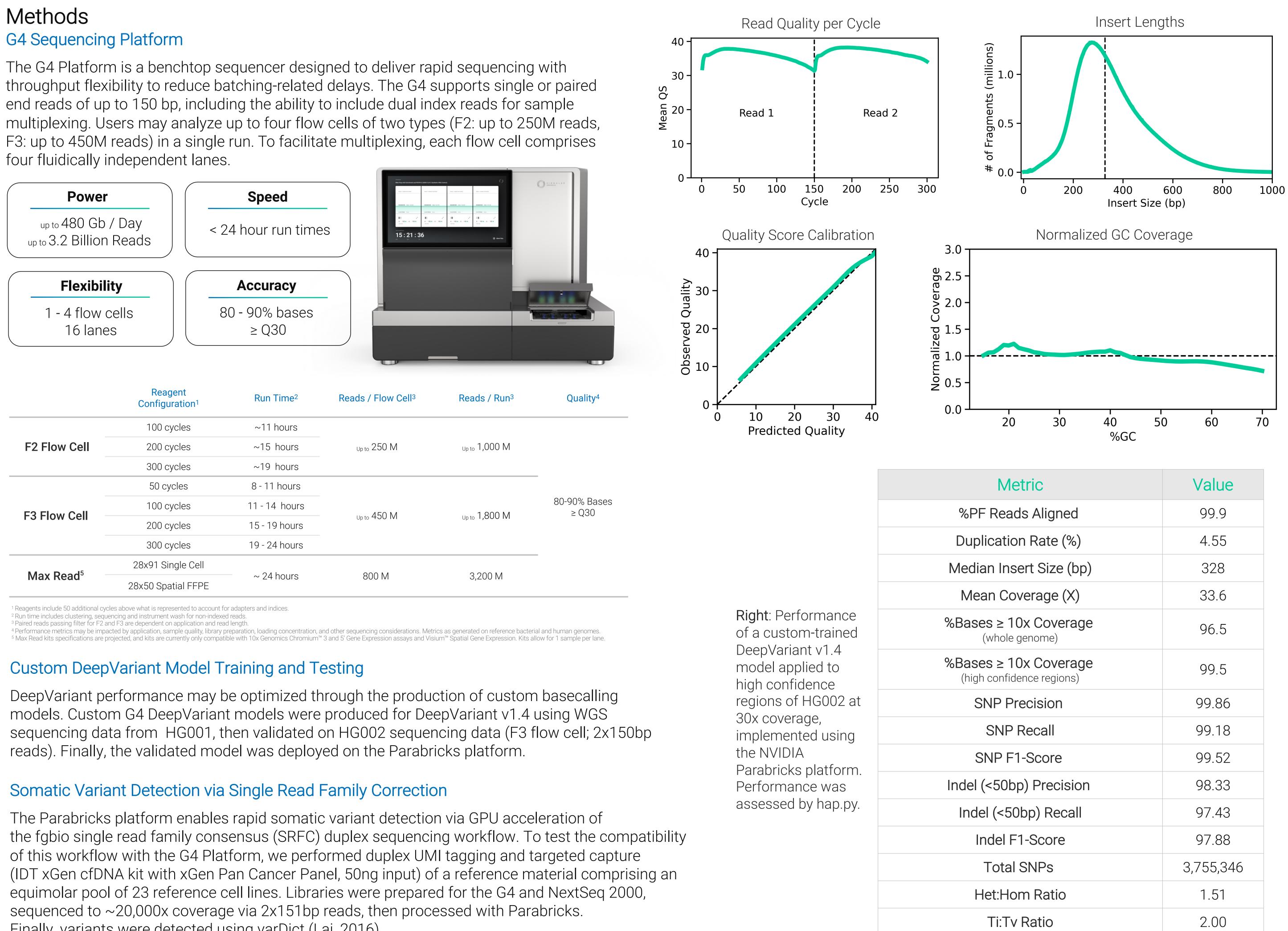
Rapid Somatic and Germline Variant Detection Using the G4TM Sequencing Platform

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Introduction

Next-generation sequencing (NGS) has become an indispensable tool for the diagnosis of genetic disease, though there remains a need to reduce turnaround for time-sensitive applications. This requires faster sequencing and accelerated data analysis. The G4™ Platform leverages a 4-color rapid sequencing by synthesis (SBS) chemistry and, in combination with advanced optics and fluidics engineering, delivers results of four human whole genomes at ~30x coverage in under 24 hours using the F3 flow cell (up to 450M reads per flow cell). We present accelerated pipelines for whole genome and targeted somatic variant detection on the G4 that leverage the NVIDIA Clara Parabricks platform.

four fluidically independent lanes.



		Reagent Configuration ¹	Run Time ²	Reads / Flow Cell ³	Reads
-	F2 Flow Cell	100 cycles	~11 hours		_{Up to} 1,(
		200 cycles	~15 hours	_{Up to} 250 M	
		300 cycles	~19 hours		
	F3 Flow Cell	50 cycles	8 - 11 hours		_{Up to} 1,8
		100 cycles	11-14 hours		
		200 cycles	15 - 19 hours	_{Up to} 450 M	
		300 cycles	19 - 24 hours		
	Max Read⁵	28x91 Single Cell	0.4 h	000 M	0.00
		28x50 Spatial FFPE	~ 24 hours 800 M		3,20

Finally, variants were detected using varDict (Lai, 2016).

Results

Whole Genome Sequencing with the F3 Flow Cell

Sequencing via a single F3 flow cell with 2x150bp reads format yielded a total of 414M read-pairs, for a mean coverage of 33.6x of the HG002 genome when discounting duplicates (4.6%), ambiguously mapped reads (5.4%), low quality base calls (0.4%), and overlapping bases (7.6%) as reported by Picard.¹ Read quality and accuracy were high (88.6% and 92.6% of base calls \geq Q30; mean single-pass accuracies of 99.87% and 99.92%, Read 1 and Read 2 respectively. Insert lengths were varied, with a median of 328bp. Base quality scores were well calibrated and there was minimal GC related coverage bias.

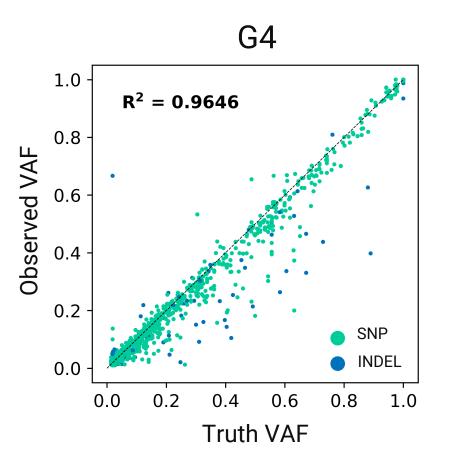
etric	Value		
ids Aligned	99.9		
on Rate (%)	4.55		
sert Size (bp)	328		
overage (X)	33.6		
l Ox Coverage e genome)	96.5		
I Ox Coverage dence regions)	99.5		
Precision	99.86		
Recall	99.18		
1-Score	99.52		
op) Precision	98.33		
0bp) Recall	97.43		
=1-Score	97.88		
I SNPs	3,755,346		
om Ratio	1.51		
v Ratio	2.00		

Results Somatic Variant Detection

	Metrics	G4	NextSeq 2000
	Mean Target Coverage	20,072x	20,066x
	% Off Bait	16.4%	14.0%
	% Targets with 0x Coverage	0.26%	0.26%
Picard –	% Excluded: Low Base Quality	1.34%	1.28%
HS	% Excluded: Overlap	37.5%	38.4%
Metrics	% Excluded: Off-Target	21.0%	19.3%
	Fold 80 Base Penalty	1.48	1.66
	AT Dropout	7.24	11.48
	GC Dropout	0.06	0.01
	Precision	79.25%	77.07%
Variant Metrics	Recall	90.03%	92.77%
	F1-Score	84.30%	84.19%

Above. Picard hybrid-selection (HS) metrics and variant calling metrics for libraries sequenced via the G4 Platform and NextSeq 2000 with 2x151bp reads. Single read families with a minimum of 3 supporting reads were retained for variant calling, via varDict using a minimum allele frequency of 0.01 and minimum read support of 2.

Results **Observed vs Expected Variant Allele Frequency**



Above Left and Middle: Observed versus expected variant allele frequencies (VAF) for G4 Platform and NextSeq 2000 experiments. Observed VAFs were highly concordant with expected allele frequencies for both instruments. **Right:** Cross-platform correlation of observed VAFs.

Conclusion

We have successfully implemented a GPU-accelerated DeepVariant whole genome model for the G4 Sequencing Platform. We further demonstrated accelerated single family UMI error correction and somatic variant detection via the Parabricks umi_fgbio workflow. We anticipate that the combination of Rapid SBS chemistry and GPU-based acceleration will significantly reduce turnaround time for most time-sensitive variant detection applications.

Acknowledgements

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