Supplemental Information

Constitutive Ras Signaling and *Ink4a/Arf* **Inactivation Cooperate During the**

Development of B-ALL in Mice

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Supplemental Methods

Antibodies used for IHC, IF, Flow cytometry and WB

IHC: Cyclin D2 (M-20, Santa Cruz Biotechnology), B220/CD45R (RA3-6B2, BD Bioscience), CD138 (281-2, BD Bioscience), BCL6 (N-3, Santa Cruz Biotechnology), CD3 (A0452, DAKO), CD5 (53-7.3, BD Bioscience), CD10 (EPR5904, LifeSpan Biosciences, Inc.), IgM (BA-2020, Vector Lab), TdT (005, Supertechs, Inc.), Ki-67 (VPK451, Vectror Lab), Arf (5-C3-1, Santa Cruz Biotechnology), Myc (Y69, Abcam); IF: CD19 (6D5, Biolegend); Flow cytometry: B220/CD45R (553087, BD Biosciences); WB: p16^{Ink4a} (M-156, Santa Cruz Biotechnology), Cyclin D2 (M-20, Santa Cruz Biotechnology), and Actin (C-11, Santa Cruz Biotechnology).

Analysis of microarray gene expression data

Arrays were corrected for background, normalized, and log2-transformed using the rma function of the affy Bioconductor package¹⁻⁴. Present/absent calls were made using the mas5calls function of the affy package. Probe sets present in >20% of samples and for which the interquartile range was >log2(1.2) were retained for further analysis. The Bioconductor limma package⁵ was used to identify probe sets significantly up- or downregulated. For significance comparisons between premalignant groups (n=2 each), a combination of P value <0.01 and fold-change >2 as the cut-off were used. A more stringent adjusted P value of <0.01 (Benjamini-Hochberg method⁶) and a fold-change >2 were used for comparisons between premalignant samples (n=2) and *CD19Cre/+;KrasG12D/+;Ink4a/ArfL/+* tumors (n=10).

To compare gene expression profiles of *CD19Cre/+;KrasG12D/+;Ink4a/ArfL/+* tumors and subpopulations of normal mouse B-cells⁷, we used probe sets common to both array platforms and corrected for batch effects with ComBat⁸. Probe sets were filtered after batch correction to retain only those that passed filtering criteria described above for both data sets. In total, 9490 probe sets were retained for unsupervised hierarchical clustering based on Pearson's correlation coefficient. Determination of

differential gene expression between tumors and normal pre-B and pro-B-cells was carried out using the 8962 probe sets that satisfied filtering criteria for these arrays (i.e. excluding mature and immature B-cells as well as premalignant and Cre-only samples). In the latter case, we used an adjusted *P* value of <0.05 and fold-change of >2 as our combined cut-off. Genes satisfying these criteria were used as novel gene signatures. Signature gene symbols were mapped from mouse to human via HomoloGene (April 2012 download) for use with human ALL gene expression profiles and were included with the "Hallmark" gene sets (v.5.1) downloaded from the Broad Institute's MSigDB website (http://software.broadinstitute.org/gsea/msigdb/collections.jsp). These combined gene sets were used for gene set enrichment analysis (GSEA) 9 . Affymetrix Human Genome U133A arrays for series GSE12995 $^{\rm 10}$ were downloaded from the Gene Expression Omnibus (GEO) and processed as above. We used 10 samples from each of the 5 following ALL subtypes: TCF3-PBX1, hyperdiploid, MLL, BCR-ABL, and TEL-AML1. Data series GSE11877^{11,12} was downloaded from GEO as a MAS5-summarized scaled-expression matrix and further log2-transformed and quantile normalized before use. GSEA for human arrays was carried out using Signal-to-Noise as the metric with base 10 values (not log2 transformed). GSEA for murine premalignant and tumor samples were run on values pre-ranked by log2 fold-change. GSEA signature heat maps showed log10(FDR) x sign(NES) values such that correlation with phenotype A in an A vs B comparison was positive (NES >0) and correlation with phenotype B is negative (NES <0). Heat maps were generated using the R package pheatmap¹³.

Analysis of whole-exome sequencing data

Lane-specific sequence files in fastQ format were aligned with BWA-mem¹⁴ to the Ensembl GRCm38.75 reference genome and further processed using the Genome Analysis Toolkit (GATK)¹⁵ best practices for targeted exome sequencing. This included duplicate-read marking, base-quality score recalibration, and local *de novo* indel realignment. Read groups were assigned and lane-specific files were merged according to GATK best-practice recommendations.

BAM-format files were loaded into the GATK's Mutect 2^{16} tool for variant-calling analysis. Control samples were processed using Mutect2's recommendations for creating a panel of normals (PON); variant calling was performed individually for each sample and the resulting VCF files were combined using GATK's CombineVariants tool for each phenotype (LG = low-grade B-cell lymphoproliferative disorder and B-ALL = precursor B-cell acute lymphoblastic leukemia). The variants found in the PON were used when analyzing variants in the tumor-derived samples (i.e., normal_panel argument). All calls to Mutect2 methods were made with default arguments.

Resultant sample-specific VCF files were merged and annotated using Ensembl Variant Effect Predictor $(VEP)^{17}$ and stored in a Gemini¹⁸ database. Prior to annotation, variants were limited to those that passed Mutect2's standard quality filters (PASS status). Further, we examined potentially high-impact variants annotated as stop-gain, frameshift, splice-acceptor, splice-donor, stop-lost, start-lost, and missense.

Remaining variants were subsequently analyzed to determine recurrent, well-covered sites featuring a high percentage of reads supporting the alternate allele. Analysis was limited variant loci found in a majority of samples (>5 for ALL, >3 for LG), having a minimum read depth of 40 reads, and with alternate allele percentage of >75%. These high-quality sites were cross-referenced to those determined in the PON analysis to remove false-positive somatic variants. Indeed, many such variants were also found in the PON (often in a single sample), apparently not removed by Mutect2 in the initial enumeration of somatic variants. Following this filtration, 36 (B-ALL) and 35 (LG) sites remained with seemingly high-quality recurrent somatic variants. Visual inspection (via IGV) was performed on these sites, revealing that the "normal" samples harbored similar mutations; such sites were not included in the PON set due to lower sequencing coverage.

Supplemental Figure Legends

Supplemental Figure 1. Conditional loss of *Ink4a/Arf* **and activation of KrasG12D expression in CD19⁺ Bcells.**

(A) Southern blot of *PstI*-digested genomic DNA from the indicated tissues and mice at 8 weeks of age. The wild-type (WT) or lox-*Ink4a/Arf* (*Cdkn2a*) allele migrates at 9.0 kb, and the recombined *Ink4a/Arf* null allele (KO) corresponds to the shorter 4.6 kb band. Note that excision of the locus only occurs in the presence of Cre recombinase.

(B) Immunoblots of p16^{Ink4a} expression in CD19⁺ splenocytes from mice at 8 weeks of age. Actin was used as loading control.

(C) RT-PCR/RFLP analysis of *Lox-STOP-Lox KrasG12D* allele recombination in testicles or CD19⁺ splenocytes from mice at 8 weeks of age. PCR-amplified cDNA was untreated (-) or digested with *HindIII* (+). Recombination of the *Lox-STOP-Lox Kras^{G12D}* allele introduces a *HindIII* site into the Kras^{G12D} transcript that results in a released fragment (TG) upon digestion with restriction enzyme.

Supplemental Figure 2. Histologic and immunohistochemical evaluation of premalignant mice.

(A) and **(B)** Histological and immunohistochemical analyses of CD3, B220, and TdT expression in lymph nodes (A) and bone marrows (B) from one additional mouse of each genotype. Positive control (insert): TdT staining of thymus from the same mouse. Scale bars: $A = 1$ mm, $B = 50 \mu m$.

Supplemental Figure 3. Gene set enrichment analysis of CD19⁺ B-cells from premalignant mice.

(A), **(B)**, and **(C)** Mountain plots documenting pathways upregulated in CD19⁺ B-cells from mice of specified genotypes compared to *CD19Cre/+* controls. Top three hallmark signatures from each genotype (A). Selected 'curated' (MSigDB C2) signatures (B) and (C).

Supplemental Figure 4. Histologic and immunohistochemical evaluation of low-grade B-cell lymphoproliferative disorder from *CD19Cre/+;KrasG12D/+* **mice.**

Representative histological and immunohistochemical analyses of B220, BCL6, TdT, CD3, CD5, and CD138 expression in a nodular lymphoid infiltrate in lungs from additional *CD19Cre/+* and *CD19Cre/+;KrasG12D/+* mice. Positive control: TdT staining of thymus from the same animal. Scale bars: 50 μ m.

Supplemental Figure 5. Histologic and immunohistochemical evaluation of precursor B-ALL in *CD19Cre/+;Ink4a/ArfL/+* **and** *CD19Cre/+;KrasG12D/+;Ink4a/ArfL/+* **mice.**

Representative histologic and immunohistochemical analyses of TdT expression (in-frame) in lymph nodes, BMs, and livers of additional *CD19Cre/+* , *CD19Cre/+;Ink4a/ArfL/+*, and *CD19Cre/+;KrasG12D/+;Ink4a/ArfL/+* mice. Scale bars: black -1 mm, white - 10 μ m.

Supplemental Figure 6. Analysis of mutations and copy number variations in mouse tumors.

(A) Summary of genomic profiles of ten B-ALLs from *CD19Cre/+;KrasG12D/+;Ink4a/ArfL/+* mice. Integer-value recurrence of CNAs across the samples in segmented data (y axis) is plotted for each probe evenly aligned along the x axis in chromosomal order. Because, *Ink4a/Arf* deletion was relatively small and focal, it was omitted by the algorithm used for data segmentation.

(B) Analysis of the probe specific for the *Ink4a/Arf* confirms the deletion.

(C) Representative immunohistochemical analyses of Myc expression in lymph nodes of mice with pre-B-ALL with (n=3) and without (n=3) chromosome 15 amplification from *CD19Cre/+;KrasG12D/+;Ink4a/ArfL/+* mice. Scale bars: 20 µm.

(D) Mean coverages of WES for low-grade B-cell lymphomas (n=6; LG) from *CD19Cre/+;KrasG12D/+* mice and B-ALLs (n=10; ALL) from *CD19Cre/+;KrasG12D/+;Ink4a/ArfL/+* mice as well as two controls per group (c1 and c2). Dotted lines represent mean for tumor group. Percentiles for the coverage depth are shown below the graph.

Supplemental Figure 7. Gene expression analysis of murine B-ALL.

Mountain plots documenting nine pathways most highly upregulated in *CD19Cre/+;KrasG12D/+;Ink4a/ArfL/+* B-ALL cells compared to *CD19Cre/+;KrasG12D/+;Ink4a/ArfL/+* premalignant CD19⁺ B-cells.

Supplemental Figure 8. Gene expression analysis of human B-ALL.

Heatmap of leading edge genes from GSEA comparing the relative enrichment of genes up-regulated in *CD19Cre/+;KrasG12D/+;Ink4a/ArfL/+* tumors relative to normal pre-B-cells ("Tumor vs Pre-B up" signature) in human BCR-ABL B-ALL samples versus all other B-ALLs.

Supplemental Table 1. Differential gene expression analysis of CD19⁺ B-cells from 8-week-old engineered mice and of B-ALL tumors from *CD19Cre/+;KrasG12D/+;Ink4a/ArfL/+* **mice.**

Supplemental References

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Supplemental Figures and Tables

Supplemental Figure 3

Supplemental Figure 5

Supplemental Figure 7

