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Supplementary Methods

Patient cohort

The IP-LBCL patient cohort is enriched for patients first diagnosed many years ago (first diagnosis 1985-2014, median 2002; relapse 1990-2022, median 2008), due to selection of relapsed LBCL-IP patients and the nature of the disease that is characterized by late relapses. As a result, clinical data were only limitedly accessible and/or available from the various hospitals. In the Netherlands, primary CNS lymphoma and testicular lymphoma were largely treated per protocol from the early 1990s on. Before 2004, primary treatment for testicular lymphoma consisted of CHOP chemotherapy, intrathecal MTX and generally radiotherapy to the contralateral testis irrespective of stage (n=18). After 2004, Rituximab was added to this regimen (n=10). Primary CNS lymphomas were treated per protocol with MBVP with or without intrathecal MTX and generally with radiotherapy to both testis as prophylaxis. Treatment for relapsed disease was not protocolized for either testicular (n=24) or for CNS relapses (n=9). As a result of the patient selection (patients with relapse), treatment of the first disease episode resulted in complete remission followed by relapse with an interval of 0.5 to 18 years (median 4 years). Treatment outcome of the relapse presentation likely largely depended on the site of relapse (CNS versus testis).

Immunohistochemistry

Immunohistochemistry (IHC) for IgM and IgG was performed on 33 primary and relapse LBCL-IP cases using standard diagnostic protocols. In brief, 3 µm sections were cut from FFPE tissue. IHC was performed using the OptiView DAB IHC Detection Kit (Roche Diagnostics) on the Ventana Benchmark Ultra (Roche, Rotkreuz, Switzerland). Primary polyclonal antibodies against IgM and IgG (IgM, 1:20000, Dako A0425, IgG, 1:80000, Dako A0423) were applied for 32 minutes. IGM and IGG expression was visually assessed as positive or negative by an expert pathologist (D.d.J.).

Laboratory procedure sequencing

Genomic DNA was extracted from formalin fixed paraffin embedded (FFPE) tumor samples with the QIAamp DNA FFPE tissue kit (Qiagen, Hilden, Germany) and fragmented with a covaris ME220 (Covaris Inc, Woburn, MA, USA). Fragmented DNA (100 ng) was used for KAPA library prep with IDT adapters (KAPA Biosystems, Wilmington, MA; IDT, Coralville, IA). An aliquot was subjected for shallow whole genome sequencing single read 50 base pairs on

the Hiseq 4000 (Illumina, San Diego, CA) in equimolar pools of 24 samples of 10 nmol/l per library.

For mutation and translocation analysis a 2.5 Mb capture panel was designed containing exonic regions of 311 genes and 23 genomic regions (Roche NimbleGen, Madison, WI; IRN 1000002633) (Supplemental Table 5 and 6). For the capture, 125 ng library per sample was used and 8 samples were pooled, 3 pools were sequenced together paired-end 150 base pairs on a Hiseq 4000 sequencer (Illumina, San Diego, CA).

Copy number alteration analysis

CNA analysis was performed as previously described.¹ In short, DNA was aligned against the reference genome (hg19) using BWA aln (v.0.7.12) and duplicates were marked with picard tools markduplicates (v.2.15). For CNA analysis the following R packages were used in R (v 3.4.1), reads were divided in 100 kb bins with QDNAseq (v1.12)² and content of guanine–cytosine basepairs (gc-basepairs), mappability and germline correction was performed. Waves were removed with NoWaves (v0.6)³ and segmentation was performed with DNACopy (v1.50.1)⁴, ACE⁵(v0) was used to estimate the tumor cell percentage and absolute copy numbers. Calling was performed with CGHcall⁶(v2.38) with the tumor cell percentage as correction factor to a minimum of a tumor cell percentage of 0.2 to minimize overcalling. To compare the primary and relapse cohort the number of regions was reduced with CGHregions (v1.34)⁷ with a maximum information loss of 20% resulting in 190 regions. These regions were compared with CGHtest (v1.1)⁸ using a Wilcoxon rank test with 10000 permutations and a false discovery rate test for multiple testing correction.

Regions for the oncoprint were determined with GISTIC (v2.0)⁹. Multiple regions on the same chromosomal arm were taken together. To determine if a CNA was different between the pairs the segments in the region were compared. If the number of segments was different between the pairs or when the beginning or end point of the segment was more than 5Mb (50bins) apart, a difference in CNA was called.

Focal regions smaller than 3 Mb in the 100kb analysis were determined by GISTIC⁹ and analyzed on 15kb bins by CGHcall.⁶ When the frequency of the focal gain or loss was >10% different in the entire cohort for the 15kb bins compared to the 100kb bins, it was added as a separate feature to the oncoprint and frequency table. One location (chr 7p12.2) met these criteria, but by visual inspection it was clear that this was an artefact that was only present in lower quality samples and therefore we excluded this.

Mutation Analysis

Mutation analysis was performed as previously described.¹⁰ Paired-end data 150 bp data was demultiplexed with Bclfastq (illumina) and adapters were trimmed with seqpurge (v0.1-104)¹¹. Thereafter the reads were aligned with BWA-mem (v.0.7.12)¹² and realignment was performed with ABRA (v2.19)¹³ to improve the alignment. The aligned reads were querysorted with sambamba (v0.5.6)¹⁴ so that also secondary alignments were included in the mark duplicate analysis of Picardtools (v2.4.1)(which is important for translocation analysis). Next, reads were coordinate sorted and only the captured exonic regions were kept for the mutation calling. Three samples were sequenced twice due to low coverage after one sequence run and data was merged during the mark duplicate step. Samples with a tumor cell percentage <10% or <20% in combination with a mean target coverage <100 were excluded for further analysis.

Mutation calling was performed in two different ways. First, all samples were analyzed in tumor-only mode by Lofreq (V2.1.3.1)¹⁵ and Mutect2 (V4.1.7)¹⁶. For Lofreq indel quality scores were obtained by lofreq indelqual, after which calling was performed using the settings [--min-cov 10 --min-mq 20 --min-bq 20 --min-alt-bq 20 --max-depth 1000 --sig 0.05] and subsequently filtered with the settings [--af-min 0.05 --cov-min 15 --sb-alpha 0.05], and minimal 2 supporting reads on both the forward and reverse read. Furthermore, calls should not be present in a blacklist that was created using an in house set of 25 normal casus including 13 matched normals from this cohort, 6 reactive lymph nodes and 6 hyperplasia samples using the same pipeline. Variants present in at least 2 samples were added to the blacklist. In addition, Mutect2 was ran with the setting [--max-mnp-distance 0] and a panel of normal file based on the 25 normals. Subsequently, the read orientation and contamination was calculated. Mutation calls were filtered by gatk filtermutectcalls with the settings [--max-events-in-region 50 --min-allele-fraction 0.05 --min-reads-per-strand 3 --unique-alt-read-count 4]. The Mutect2 germline filter was not used, since multiple *MYD88* mutations were predicted as germline when they were in conjunction with a chromosomal gain on that locus. In addition, mutations were filtered when they had a ROQ score <20 or TLOD score <20 to reduce noise. Mutations detected by both tools were annotated by snpSift (v4.3)¹⁷ and funcoator (v4.1.7)¹⁸ using the databases of COSMIC (v84)¹⁹, Gnomad (v2.0.2)²⁰, gencode(v19)²¹, dbsnp (build 151)²², clinvar (20180401)²³, and HMF panel of normals (v2.0). Mutations labeled as common in dbsnp or present >3 times in HMF panel of normals were excluded. Also intronic, silent, UTR or flanking mutations were removed for further analysis.

For 12 samples, a matched normal sample was available. These samples were analyzed with Mutect2 in tumor-normal mode.¹⁶ Called mutations of the tumor only and tumor-normal pipeline were taken together, discrepancies of the pipelines were manually curated using IGV, germline

events, noisy events and events with only 2 reads evidence were excluded (105/1871 mutations). Non-shared mutations were manually scrutinized by visual inspection in IGV to confirm presence/absence of mutations. Comparisons of the number of mutations was performed with a Wilcoxon test of paired data. Frequencies of alterations were compared between primary and relapse samples using a Fisher-exact test on all alterations present in >10% of the samples, false discovery rate was calculated according to Benjamini & Hochberg method. A correlation between time of diagnosis and number of shared mutations was calculated by fitting linear regression models using the R stats package v(3.6.1).

An oncoprint was created with Complexheatmap (2.7.1.1016)²⁴ and circosplots were made with circlize (0.4.13)²⁵. For somatic hyper mutation (SHM) comparisons, intronic and silent mutations were included as well. SHM comparisons were performed for the known target genes IGLL5, PIM1, BTG1, BTG2 and OSBPL1.²⁶⁻²⁸ Analyses were performed with R (v3.6.1).

Translocation analysis

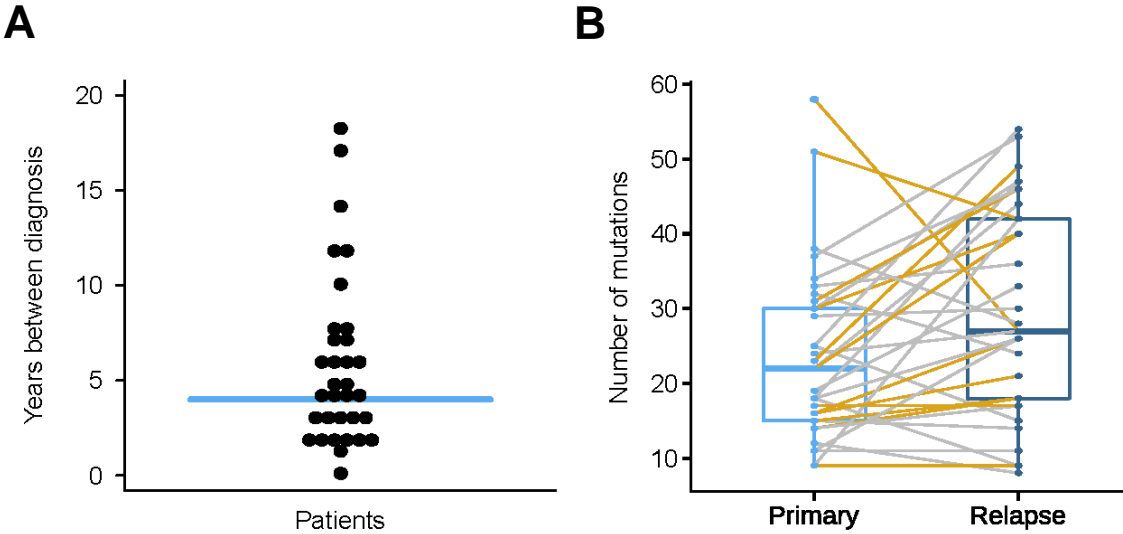
Translocation analysis was performed as previously described.²⁹ In short, 4 structural variant (SV) tools were combined using Snakemake: Breakmer (v0.0.4)³⁰, NovoBreak (v1.1.3)³¹, GRIDSS (v1.4.2)³² and Wham (v1.7.0)³³. All tools were executed using default settings except Wham which was done with a mapping quality of 10 and base quality of 5. For Breakmer, a target file containing all translocation regions of the capture panel was used, with the regions divided in non-overlapping 5kb regions. Subsequently, SVs were filtered based on the following criteria: at least one of the SV breakpoints should be within 300 bp of the capture area and SVs should meet the thresholds set for each tool (Breakmer: at least 4 split reads and 3 discordant reads, Wham: at least 8 supporting reads, GRIDSS quality score above 450 and for NovoBreak an coverage of 4 high mapping quality translocation reads). Duplicate SVs detected with the same tool were removed. Then, SVs should be detected by at least 2 of the 4 tools and breakpoints within 10bp of each other were considered to be the same SV. In addition, a blacklist was created using the 25 normal samples described above. SVs detected less than 10bp apart were added to the blacklist and areas less than 50 bp apart were merged to one larger area. The latter was done because these were often highly repetitive regions or partly homologous regions, giving rise to many false positives. SVs with a breakpoint within the blacklisted region were removed from the output. The remaining translocations were manually inspected in IGV after which only true translocations were kept.

Immunoglobulin clonality analysis

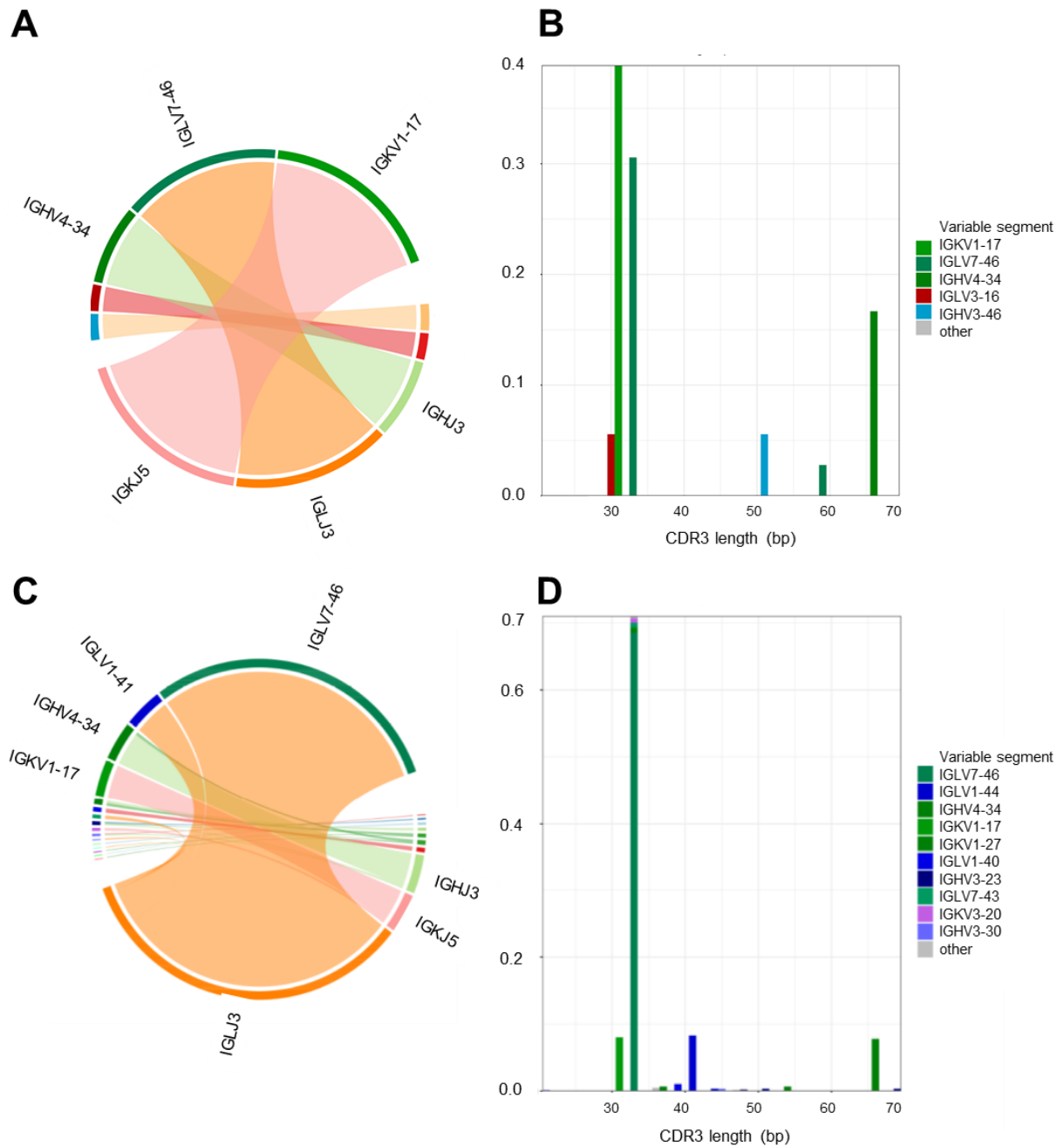
For the clonality assessment the paired-end 150bp aligned bam files were used. To improve the accuracy and speed of the analysis only reads aligned to the coding parts of IG genes were used that were partly soft-clipped as determined with samtools and samclip (v2.0). The V(D)J recombination was determined using Abstar (v0.3.5)³⁴. The JSON output of Abstar was converted using R (v3.5.1) to csv format mimicking the IMGT output. All reads with exactly the same CDR3 were assigned to the V-gene with the longest fragment that map to the V-genes, because this fragment has the highest mapping reliability. The data was visualized with VDJtools (v1.2.1)³⁵. Samples with the same CDR3-length and similar CDR3 sequence were considered clonal. Sample pairs for which no clonal V(D)J recombination was detected were manually inspected using IGV. If the same breakpoint in the IG region was found in both samples it was considered clonal as well.

To confirm validity of this method and to study 3 cases that were inconclusive with our method, 12 samples were analyzed using the novel EuroClonality assay for clonality detection of immunoglobulins. The assay was performed using two-step PCR clonality protocol for Illumina sequencing, which is an attuned protocol for FFPE- derived DNA based on the previously described two-step PCR protocol and the IG-clonality protocol for Ion Torrent sequencing.³⁶⁻³⁸ The data was evaluated by using ARResT/Interrogate and the evaluation criteria described in van den Brand et al. 2021.³⁹

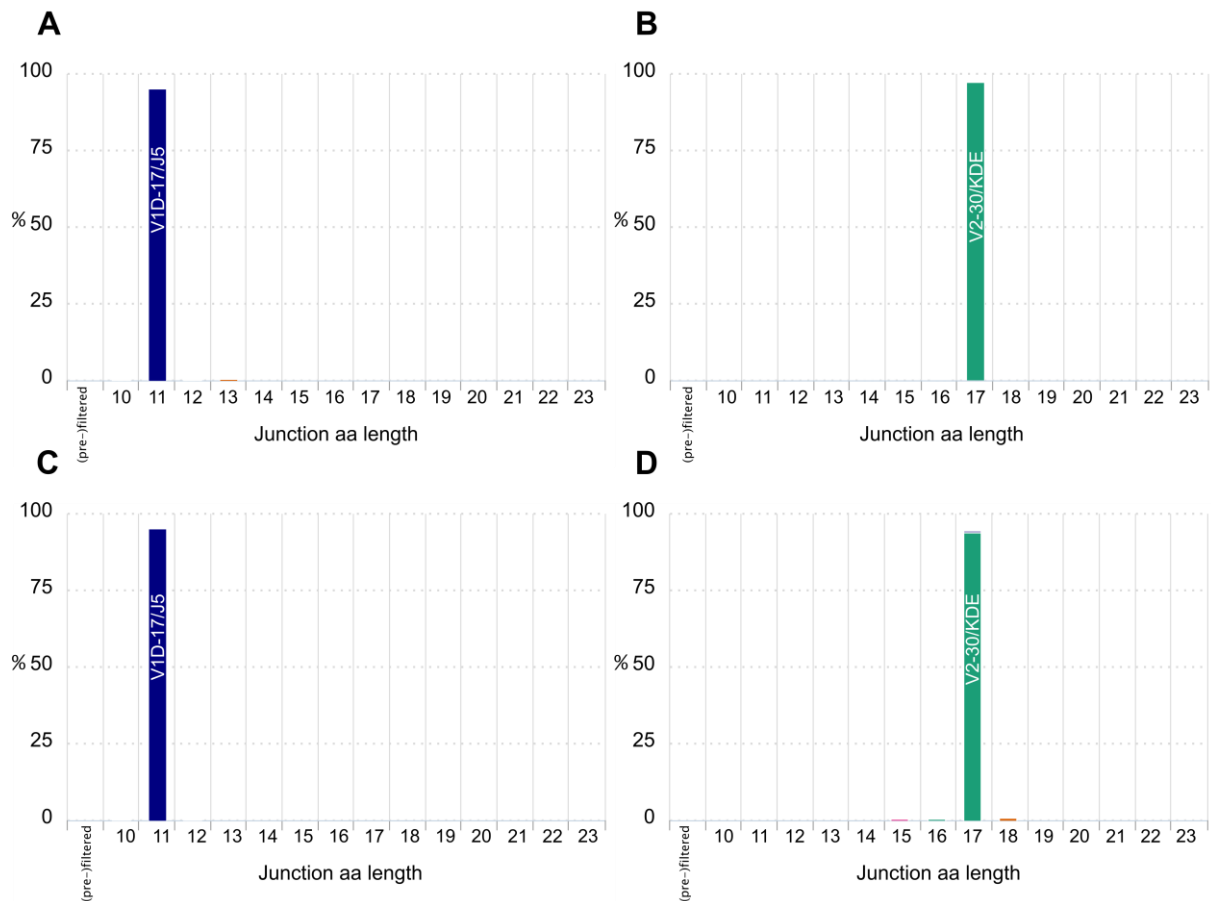
Supplementary Figures



Supplementary Figure S1. Time between diagnosis and number of mutations. (A) The time between diagnosis depicted as beeswarm plot. Each dot is 1 patient, the median time between diagnosis is 4 years, (range 0-18 years). (B) The number of non-synonymous mutations in the primary samples and the relapse samples are depicted in boxplots. Sample pairs are connected by a line. Yellow samples indicate the presence of a matched normal and samples in grey are without matched normal. The fast majority of the relapse samples have a significantly higher number of mutations as determined with a Wilcoxon rank sum test in paired sample mode (23.5 vs 30 mutations on average, P value = 0.009).

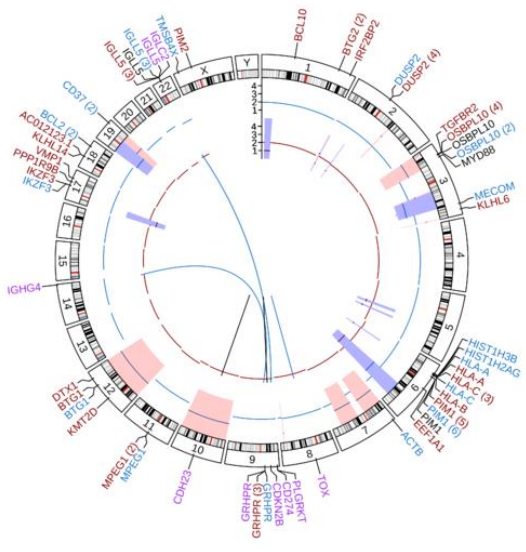


Supplementary Figure S2. Example of a clonal sample pair (patient #31). (A) Circular representation of immunoglobulin VJ recombinations determined with Abstar³⁴ and visualized with VDJtools³⁵ of the primary sample of patient #31, with in the lower part the J-segment and in the upper part the V-segment. (B) Barplots of the CDR3 lengths of the V(D)J recombination of the primary sample of patient #31. (C) Circular representation of immunoglobulin VJ recombinations of the relapse sample of patient #31. Details are as in A. Same segment combinations were seen in primary (A) and relapse (C) samples. (D) Barplots of the CDR3 lengths of the V(D)J recombination of the primary sample of patient #31. The major combinations in the primary and relapse sample are of the same length with the same V-J combination, indicating a clonal relationship.

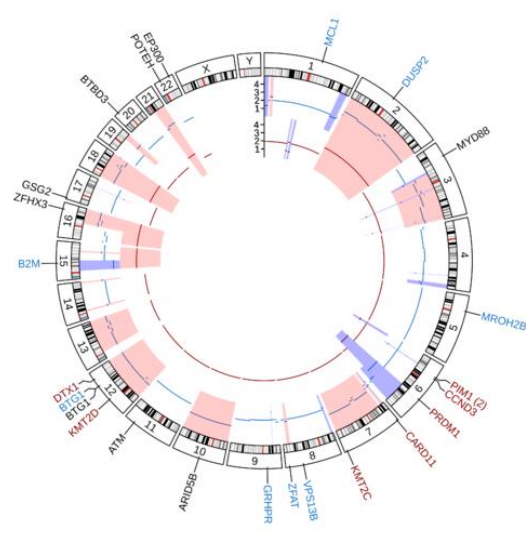


Supplementary Figure S3. Example of a clonal sample pair (patient #31) as determined with Euroclonality ARResT/Interrogate. Barplots of the amino acid junction length determined with Euroclonality ARResT/Interrogate of IGK-VJ in the primary (A) and relapse (C) sample of patient #31. Barplots of the amino acid junction length determined with Euroclonality ARResT/Interrogate of IGK-DE in the primary (B) and relapse (D) sample of patient #31.

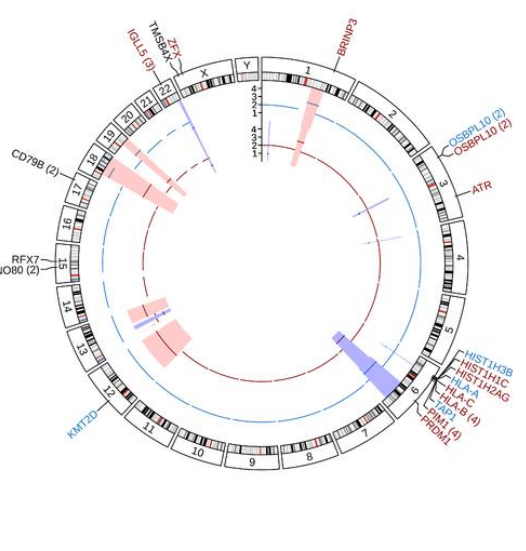
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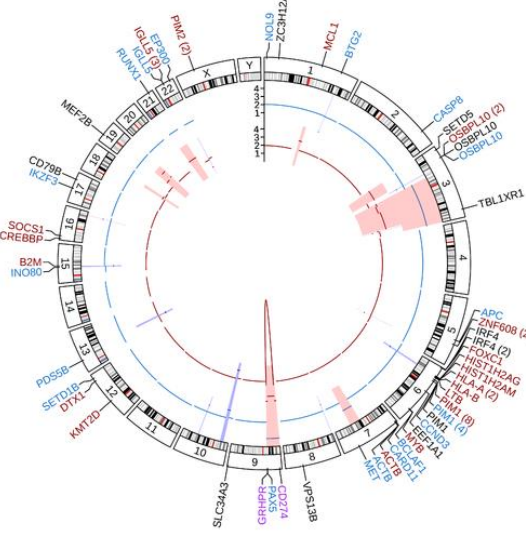
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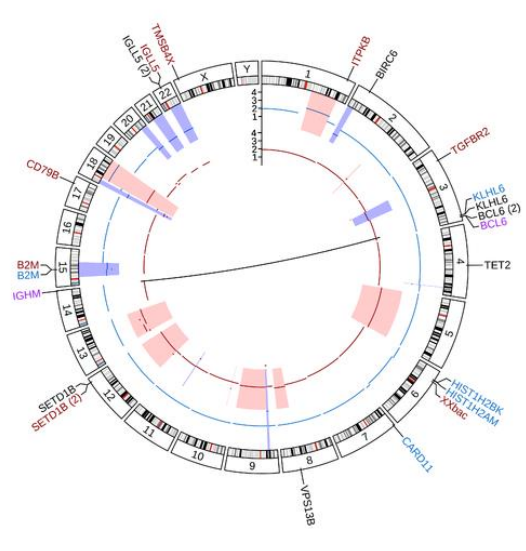
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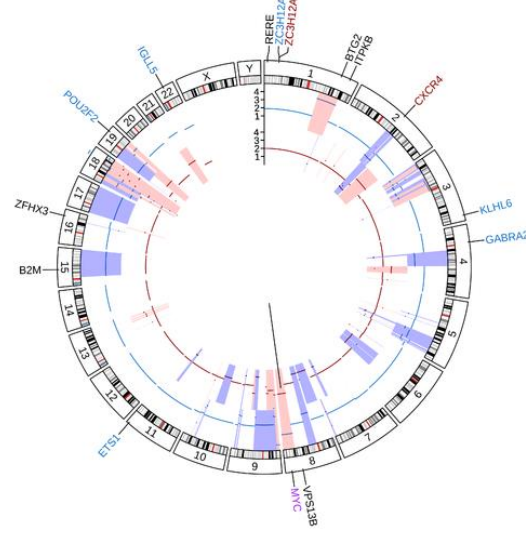
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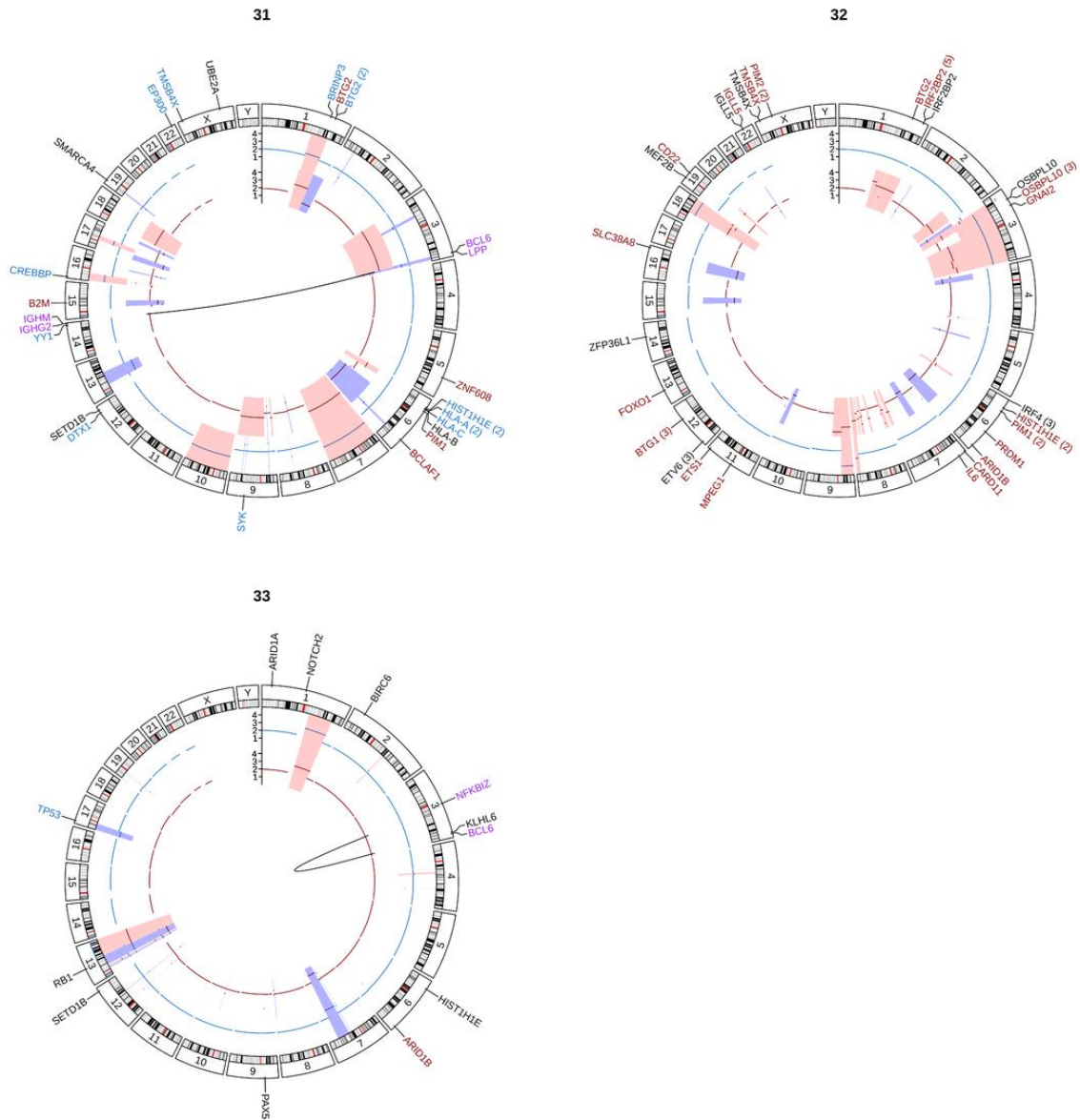


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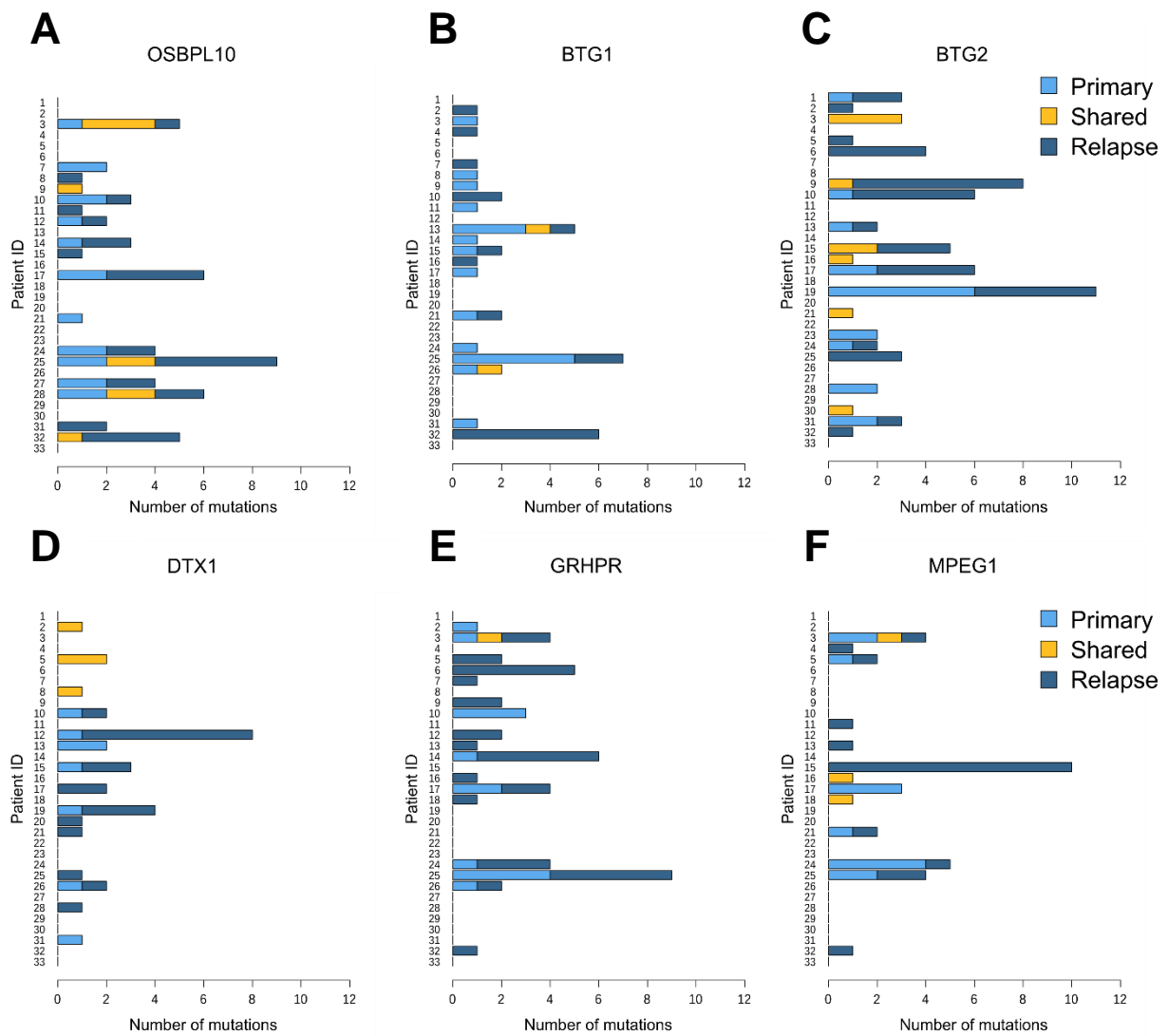


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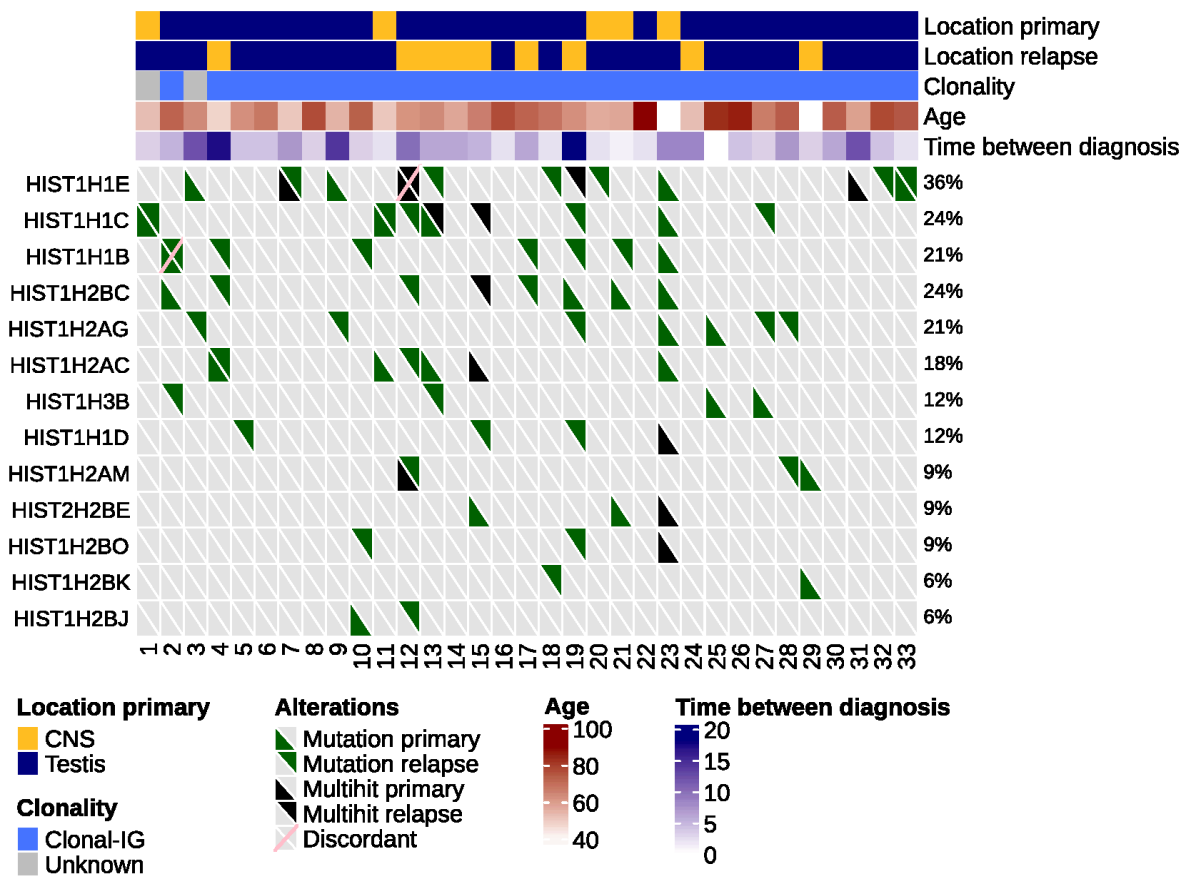




Supplementary Figure S4. Circosplots of alterations for each patient. Circosplot of all alterations in all primary and relapse samples (#1 - #33). Names of mutated genes are plotted around the ideogram of the chromosomes, in blue unique mutations of the primary sample, in red unique mutations of the relapse sample and in black the shared mutations between primary and relapse samples. If more than one mutation in the gene is present, the number of mutations is indicated in parentheses. Genes with translocations are indicated in purple, the connecting lines inside the ideogram indicate if the translocation is in the primary (blue), the relapse (red) sample or shared (black). Inside the circosplot the CNAs are depicted, in the outer circle the primary and in the inner circle the relapse CNAs. Copy number gains are depicted in red and losses in blue.



Supplementary Figure S5. aSHM frequently differed between primary and relapse LBCL-IP samples. Barplots of unique and shared aSHM. The mutations in the primary samples are indicated in light blue, the relapse samples in grey blue and the shared mutations in orange. (A) In OSBPL10 on average 14% of the SHM was shared, (B) in BTG1 on average 4% were shared mutations, (C) in BTG2 on average 23% were shared mutations, (D) in DTX1 on average 20% were shared mutations, (E) in GRHPR on average 1% were shared mutations and (F) in MPEG 1 on average 17% were shared mutations.



Supplementary Figure S6. Unique and shared linker histone mutations in paired samples. Oncoprint of linker histone alterations (rows) in sample pairs (columns). If present in the primary sample the bottom left corner is colored (according to the colors in the legend) and if present in the relapse sample the upper right corner is colored. If the alteration is discordant in the primary and the relapse sample a line is drawn across the box. In the upper panel, the location of the IP-LBCL, the clonality based on immunoglobulin analysis, age of the patient and time between diagnosis is indicated.

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