Evolving Understanding of the CLL Genome

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Over the past few years, massively parallel sequencing technologies have revealed with high resolution the tremendous genetic and epigenetic heterogeneity in chronic lymphocytic leukemia (CLL). We have learned how the molecular architecture differs not only between affected individuals but also within samples and over time. These insights have catalyzed our understanding of the pathobiology of CLL and point to critical signaling pathways in the development and progression of the disease. Several key driver alterations have been identified, which serve to refine prognostic schemata but also to inspire the development of new therapeutic strategies. Ongoing advances in technology promise to further elucidate the molecular basis of CLL, and this knowledge is anticipated to aid us in understanding and addressing the clinical challenge presented by the vast variability in the clinical course of patients with CLL.

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A hallmark of chronic lymphocytic leukemia (CLL) is its tremendously variable clinical course. As many as 80% of CLL patients are asymptomatic at diagnosis, but many progress to extensive lymphadenopathy, hepatosplenomegaly, and life-threatening cytopenias within only a few years. Others, however, remain asymptomatic over decades, with 20% to 30% having a life expectancy not significantly different from the general population.\textsuperscript{1,2}

An enduring goal of CLL studies has been to better understand the basis of this clinical variability. Of note, because of its high prevalence, relatively slow progression, and the ready availability of leukemia samples from patient peripheral blood, CLL has been continuously at the forefront of genomic research. Thus, while the first prognostic schema, established in the 1970s,\textsuperscript{3,4} was based on clinical features, newer studies have focused on the role of somatic genomic alterations in the pathogenesis of CLL and, in turn, have examined their impact on clinical outcome. For example, mutational status of the immunoglobulin heavy chain variable (\textit{IGHV}) gene divides CLL into two genetically distinct groups, which likely reflects different cells of origin and has emerged as an important disease-associated prognostic factor.\textsuperscript{5,6} In a separate landmark study, presence of four common CLL-associated cytogenetic aberrations, deletions of chromosomes 11q, 13q, and 17q as well as trisomy of chromosome 12, could stratify CLL patients into prognostically distinct groups and, for the first time, was linked with clinical course and survival.\textsuperscript{7}

Remarkably, the introduction of next-generation sequencing has led to a breathtakingly exponential increase in the knowledge of the molecular underpinnings of CLL over the last 3 years (Figure 1). These recent studies have revealed several notable, even surprising and paradigm-shifting insights such that our perception about this disease has greatly evolved in a relatively short span of time. First, whole-exome sequencing (WES) of large sample cohorts has clearly shown the high degree of genetic variability among CLL patients, with the discovery of novel common gene mutations that likely play a role in the pathobiology of CLL. Furthermore, a startlingly high degree of intra-sample clonal heterogeneity was recently uncovered, based on the detection and quantification of leukemia-specific alterations that mark various cell subpopulations within a CLL sample. At the same time, genome-wide approaches to examine the DNA methylome have revealed epigenetic heterogeneity among patients and within individual samples. Collectively, these observations demonstrate the complex interrelationship between genetic and epigenetic features of each sample. Herein we review these novel discoveries, which are now undergoing evaluation as features to incorporate into new prognostic and therapeutic schema that promise to improve the clinical care of patients with CLL.
PROFILING GENETIC HETEROGENEITY IN CLL

Massively parallel sequencing techniques have provided the ability to rapidly sequence millions of DNA fragments with relatively low sample input. As an alternative to whole genome sequencing, selective restriction of reads to the coding regions of the genome by WES has drastically reduced the costs of sequencing per sample. This latter approach has facilitated the rapid sequencing of large sample cohorts, which in turn has enabled the drawing of associations between genetic alterations and clinical features. Over the last few years, the results of up to a dozen whole CLL genomes\(^8\)–\(^{10}\) and \(\approx 300\) whole CLL exomes\(^11\)–\(^{14}\) across different centers worldwide have been reported.

UNBIASED DISCOVERY OF KEY CLL DRIVERS

A major goal of the large-scale cancer sequencing studies has been the identification of key alterations that drive malignancy. The development of massively parallel sequencing has led to the parallel development of advanced computational algorithms for analyzing these big datasets. In general, these algorithms detect cancer-specific alterations with a high probability of being cancer drivers on the basis of whether they are present at a significantly higher-than-expected rate given the known background mutation rate of the cancer. In CLL, these efforts have corroborated known CLL-associated alterations (ie, mutations in \(TP53\) and \(ATM\)) but importantly have identified numerous previously unknown somatic changes, the majority of which have been confirmed across independent sample cohorts. The first studies of DNA sequencing in CLL found mutations in \(MYD88\), \(NOTCH1\), and \(XPO1\), as well as in \(BIRC3\).\(^8\),\(^{11}\) Subsequently, results of WES of two well-powered cohorts of \(\approx 100\) patients each further detected several novel somatic alterations in CLL (in \(FBXW7\), \(POT1\), and \(CHD2\)). Strikingly, both studies identified the novel finding of recurrent mutations in the splicing machinery co-factor \(SF3B1\) in 10% to 15% of patients.\(^9\),\(^{15}\) Most recently, the largest single CLL sequencing cohort to date was reported, comprising 160 patients, in which numerous lower frequency mutations (in \(NRAS\), \(KRAS\), \(HIST1H1E\), \(SAMHD1\), and \(MED12\)) were identified.\(^{13}\)

These somatic alterations are present in critical components of a number of cellular pathways and include DNA damage and cell cycle control (\(TP53\), \(ATM\), \(POT1\), and \(BIRC3\)), mRNA processing (\(XPO1\) and \(SF3B1\)), NOTCH signaling (\(NOTCH1\)), inflammatory pathways (\(MYD88\)), and chromatin modification (\(CHD2\)) (Table 1).\(^9\),\(^{13}\),\(^{15}\) Several of the significantly mutated genes display a clustering of mutations in hot spots within highly evolutionarily conserved gene regions and strongly support the idea that they are positively selected gain-of-function

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**Figure 1.** Evolution and growth in our understanding of chronic lymphocytic leukemia (CLL) heterogeneity over time.
alterations. For example, mutation in MYD88, a critical adaptor molecule of the interleukin-1 receptor–Toll-like receptor (TLR)-signaling pathway, has been found almost exclusively at position L265P, which is localized within the interleukin-1 receptor–TLR domain. It is noteworthy that this particular mutation seems to occur selectively in B-cell malignancies, most prominently in Waldenström’s macroglobulinemia and diffuse large B-cell lymphoma.16,17 In another example, NOTCH1, a transmembrane protein central to the Notch-signaling pathway, is recurrently affected by a 2 base pair frameshift deletion in the C-terminal PEST domain,11 leading to pathway activation, increased cell survival, and resistance against pro-apoptotic stimuli.18 SF3B1 encodes the core catalytic subunit of the spliceosome complex, and its mutations localize to 900 base pairs within the C-terminal region9,15,19 and have been noted to affect splicing at 3’ splice sites.20,21 Another recurrently mutated gene affecting RNA processing is the nuclear transport gene XPO1, with mutations clustering at a highly conserved site at residue E571K.9,9,22,23 Finally, the shelterin POT1, encodes a protein essential for telomere function, of which recurrent mutations in CLL affect key residues required to bind telomeric DNA and lead to substantial telomeric dysfunction associated with increased genomic instability and numerous chromosomal abnormalities.14

The significantly mutated CLL genes also include examples of tumor suppressor genes (TP53, BIRC3, and ATM). TP53 is furthermore involved in the region of chromosome 17p, and BIRC3 and ATM at 11q, which are often found deleted in CLL and which correspond to poor prognosis.9,13,24

Further clues on the functional role of alterations can be inferred based on the patterns of co-segregation and mutual exclusivity (Table 1). Interestingly, the significantly mutated genes in CLL seem to be differentially represented between the IGHV mutated and unmutated CLLs. Although the former seems to be associated with del(13q) and mutations in MYD88, the latter commonly display mutations affecting NOTCH1, SF3B1, ATM, and TP53 and associations with trisomy 12, del(11q) and del(17p), respectively.16,17,25–27 Lesions of BIRC3 and TP53 have been noted to occur in a mutually exclusive fashion. Likewise, mutations in SF3B1, NOTCH1, and MYD88 also seem to be exclusive of each other. These patterns suggest possible distinct evolutionary paths, whereby certain subclonal alterations may confer advantage when occurring in the genomic context of particular ancestor lesions. Alternatively, mutual exclusivity could indicate that alterations have highly similar downstream effects; thus, functionally redundant secondary mutations do not provide any further advantage to the tumor cell. On the other hand, consistent co-occurrence suggests synergistic effects between alterations that enhance fitness of the malignant clone and promote selection of driver combinations.

As the numbers of studies examining the incidence of these key mutations in CLL have grown, it has also become clear that their frequency in patient groups largely depends on the composition of the sequenced cohort. Thus, although mutation frequency in SF3B1 ranges between 4% and 12% in early CLL, it rises to 17% to

Table 1. Evidence for Co-segregation and Mutual Exclusivity of Genetic Alterations in CLL

<table>
<thead>
<tr>
<th>Pathway</th>
<th>Alteration</th>
<th>Co-segregation</th>
<th>Mutual Exclusivity</th>
<th>Location</th>
</tr>
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<tr>
<td>Chromatin modification</td>
<td>CHD2</td>
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<tr>
<td>Inflammatory pathways</td>
<td>MYD88</td>
<td>Del(11q)9,22,25</td>
<td>SF3B1,25 NOTCH125</td>
<td>15q26</td>
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<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>3p22</td>
</tr>
<tr>
<td><strong>Alterations associated with U-IGHV</strong>8,9,11,14,15,22,25,27,28,82,83</td>
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<tr>
<td>DNA damage response, cell cycle control</td>
<td>ATM</td>
<td>Del(11q)9,22,25</td>
<td>TP53,24,29 NOTCH124</td>
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<td>SF3B1,28 Tris(12),49 BIRC324</td>
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<td>Tris(12)9,22</td>
<td>SF3B1,22 NOTCH19</td>
<td>4q31.3</td>
</tr>
</tbody>
</table>

*Involved in corresponding common chromosomal alterations.

Abbreviations: CLL, chronic lymphocytic leukemia; M-IGHV, mutated immunoglobulin heavy chain variable (IGHV) gene; UM-IGHV, unmutated IGHV gene.
24% of patients by the time of disease progression. Similarly, mutations in TP53, ATM, and NOTCH1 have a higher incidence in cohorts with advanced disease across independent studies. The mutation rate of NOTCH1 is further markedly increased in patients with lymphomatous transformation. By contrast, mutation rates of MYD88 appear stable throughout the course of CLL (Figure 2). Altogether, these patterns of association between distinct genetic alterations suggest the presence of different potential trajectories in the development of the CLL genome. Moreover, the patterns of genetic alterations are distinct in each individual CLL, and they demonstrate the tremendous interindividual heterogeneity in CLL (Figure 3A).

**INTRA-LEUKEMIC GENETIC HETEROGENEITY AND CLONAL EVOLUTION**

Even as we have gained greater understanding of variation in the spectrum of genetic alterations among patients with CLL, analysis of WES data has revealed the extensive degree of genetic heterogeneity within individual samples. The existence of genetically distinct subpopulations was already suggested by studies using fluorescence in situ hybridization, in which many chromosomal alterations were observed only in subfractions of cells. Moreover, the number of genomic alterations was shown to increase throughout the course, from newly diagnosed to progressive and further to relapsed CLL. Application of WES and whole-genome sequencing methods, however, have provided more precise quantitation and more global assessment of this phenomenon.

Using these technologies, a series of recent studies have indicated that a single time point genetic profile of a CLL sample represents a snapshot of multiple different tumor cell populations that are related to each other and changing over time. By integrating information on the allelic frequencies of mutations together with local copy number and purity information, Landau et al. demonstrated the possibility of inferring proportions of cell subpopulations harboring a genetic alteration (Figure 3B). Interestingly, certain somatic mutations were detected preferentially in clonal or subclonal fashion, suggesting an order of alterations corresponding to earlier and later drivers. Presence of a subclonal driver by itself also indicated poor prognosis and more rapid disease progression. Furthermore, of the 149 patients in the study, a subset of samples assessed longitudinally showed clear patterns of clonal evolution commonly after therapy, which was also associated with worse overall outcome. Importantly, aggressive subclones representing a majority of the tumor cell population at relapse were often already detectable in pretreatment samples. Complex and highly heterogeneous evolutionary dynamics with linear and branching subclones, as well as marked clonal shifts over time and multiple cycles of therapies, were also seen in smaller series of 22 patients studied by using DNA arrays as well as in three patients repeatedly assessed by next-generation sequencing in a detailed multi time-point analysis over up to 7 years.

**PROFILING EPIGENETIC HETEROGENEITY IN CLL**

The recent explosion in the number of mutations in known epigenetic regulatory genes identified across human
Cancers has underscored the importance of epigenetic control in tumor suppression. It is highly likely that these epigenetic changes cooperate with genetic mutations to mold the evolutionary tumor landscape. The best-studied epigenetic modification to date is CpG methylation, which occurs by conversion of cytosine in DNA to 5-methylcytosine by addition of methyl groups to CpG sites regulating gene expression. In cancer, global genome-wide hypomethylation is accompanied by localized hypermethylation and an increase in expression of DNA methyltransferase.

Epigenetic Differences among Patients

CpG methylation profiles clearly differ among prognostic subcategories of CLL. These aberrantly methylated loci have been shown to include genes involved in CLL pathobiology, such as BCL2, TCL1, death-associated protein kinase 1 (DAPK1), LPL, ZAP70, and NOTCH1, as well as gene regulators and pathways involved in B-cell signaling. Also aberrantly methylated in CLL are several microRNA (miRNA) promoters, and this change can affect the expression of those miRNAs with known roles in CLL. Finally, two lincRNA isoforms were found deregulated in CLL through changes in their promoter methylation and degree of histone modification. They map to the frequently deleted region of chromosome 13q14 and seem to play a role in controlling transcription of multiple other genes at this locus.

Highly comprehensive genome-wide methylation profiling has been performed by using the Illumina 450 K arrays. These DNA hybridization chips evaluate 485,000 methylation sites per sample, covering >95% of CpG islands but also miRNA promoters and CpG sites outside of CpG islands. In a study analyzing samples from 139 CLL patients gathered by using this technology, significant methylation differences were found in association with CLL genotypes (ie, mutations in SF3B1 and NOTCH1; trisomy 12 or del[11q]). Furthermore, methylation

Figure 3. (A, B) Inter- and intra-leukemic genetic and (C, D) epigenetic heterogeneity in chronic lymphocytic leukemia (CLL) revealed by next-generation sequencing. Panels A, B, and C were adapted with permission from Wang et al, Landau et al, and Kulis et al, respectively. Panel D was provided by Landau et al.
profiling was able to categorize patients into three distinct clinicopathologic groups (Figure 3C). Specifically, CLLs with unmutated *IGHV* were strongly related to naive B cells, whereas CLLs with predominantly mutated *IGHV* aligned with mature B cells. These results suggest a methylation imprint corresponding to the putative cell of origin. Unexpectedly, a third group was uncovered with mainly mutated *IGHV* but with a methylation signature more similar to naive B cells. Compared with the other two groups, patients in the third group had an intermediate clinical outcome. It was noted that methylation of CpGs in the gene bodies outside CpG islands showed strongest correlation with gene expression.45

**Intra-leukemic Epigenetic Heterogeneity**

Given the existence of genetically distinct subclones within individual patient cancer cells, it might be expected that epigenetic alterations could also show complex intra-tumoral differences with changes over time. Indeed, recent analysis of 450 K array data revealed high heterogeneity within individual CLL samples compared with healthy B cells (Figure 3D).46 Also using this platform, Cahill et al42 examined methylation in paired diagnostic and follow-up samples. Although they found > 2000 sites differentially methylated between *IGHV* mutated and unmutated CLL, they observed no significant differences in methylation patterns over time or between peripheral blood and lymph nodes. These data support the idea that altered methylation is an earlier, rather than later, leukemogenic event. A recent 450 K array analysis on 28 longitudinally followed up CLL cases, however, showed that the majority of genetically evolved cases also showed epigenetic changes but that epigenetic evolution was not observed in the absence of genetic changes.47 These results suggest a temporal hierarchy in which genetic alterations precede marked epigenetic changes and yet co-operate together. In the same study, Oakes et al additionally applied next-generation targeted bisulfite sequencing on 28 selected regions. In contrast to array-based detection, massively parallel sequencing technologies provide the opportunity to study methylation at base pair resolution and with sequence context. Thus, allele-specific methylation in CLL could be identified, exhibiting a stochastic pattern with random distribution between neighboring CpGs, which contrasted starkly with physiologically imprinted regions that showed methylation consistently occurring on the same allele. Reduced-representation bisulfite sequencing is a recently introduced NGS technology that enables genome-scale methylation examination restricted to CG-rich sites, and hence is cost-effective. A recent reduced-representation bisulfite sequencing study of 104 primary CLL samples uncovered the high degree of intra-tumoral methylation heterogeneity in CLL compared with normal B cells. This heterogeneity was evaluated to stem primarily from stochastic variation in DNA methylation, termed “locally disordered methylation.” Thus, the high heterogeneity in methylation status within patient samples seems to arise not from an admixture of cell subpopulations with distinct ordered methylation patterns but rather from an increased proportion of cells with disordered methylation of their genome. This high degree of “noise” in the CLL methylome was also linked with altered transcriptional regulation, clonal evolution, and adverse clinical outcome.46

**CLINICAL IMPLICATIONS OF GENOMIC DISCOVERIES IN CLL**

Many of the recent insights on CLL gained by the various large-scale sequencing studies have substantial clinical implications. Overall, they provide new markers that serve to refine prognostic information but also point to promising novel therapeutic targets and new strategies for rationally approaching CLL treatment.

**Refining CLL Prognostic Schemata**

Multiple studies on unselected cohorts comprising mainly untreated patients have demonstrated the prognostic value of four driver alterations: mutations in *SF3B1*, *NOTCH1*, *BIRC3*, and *TP53*.27,31,11,19,22,24,28 Other studies assessed the impact of these lesions in more selected cohorts, such as patients enrolled in therapeutic clinical trials. For example, analysis of the German CLL4 trial cohort, which compared fludarabine with or without cyclophosphamide, found that the poor prognostic impact of *TP53* mutations was equivalent to chromosomal del(17p), and that these lesions together were associated with extraordinary poor response to chemotherapy.49 Similarly, in the UK LRF CLL4 trial, comparing different fludarabine-containing regimens, integration of mutations in *TP53* together with those in *SF3B1* and *NOTCH1* confirmed that patients with *TP53* alterations had the shortest survival and poorest therapeutic response. However, *SF3B1* and *NOTCH1* mutations, albeit not associated with fludarabine response, both had an independent negative impact on overall survival.27 Finally, in the German CLL2H cohort, who received alemtuzumab in the setting of fludarabine refractoriness, patients with *NOTCH1* mutations also had longer progression-free survival, suggesting that this subgroup might particularly benefit from the anti-CD52 antibody treatment. In contrast, *SF3B1* mutation had no impact on response rates or overall and progression-free survival in this population.50 These contrasting results indicate that different drivers may have a variable role, depending on the context.

Given the explosion of studies that have genotyped patients across unselected cohorts and a limited set of clinical trial cohorts, Rossi et al12 sought to integrate this mutational information (on *TP53*, *NOTCH1*, *SF3B1*, and *BIRC3*) together with conventional FISH cytogenetic data in an effort to improve the predictive ability of clinical prognostic schema. Indeed, when this was done, predictive accuracy was significantly enhanced in a large cohort of
637 newly diagnosed CLL cases and was validated in an independent group of 370 patients. Patients with TP53 or BIRC3 lesions consistently had the worst prognosis, followed by patients with mutations in SF3B1 and NOTCH1 and del(11q). Notably, these high-risk lesions developed in ~20% of initially low-risk subjects during the course of the disease, among whom del(11q) and abnormalities of TP53, NOTCH1, and SF3B1 as well as BIRC3 again represented a major fraction. Another large study assessing mutations of TP53, NOTCH1, SF3B1, FBXW7, MYD88, and XPO1 together with cytogenetics and IGHV mutational status in 1160 patients confirmed the strong prognostic value of NOTCH1, TP53, and SF3B1 alterations with independent impact on overall survival of the latter two. 52

The aforementioned studies have addressed the question of whether the presence or absence of a driver mutation in a CLL sample is prognostic. However, it has become evident that addressing the fraction of CLL cells bearing a driver mutation within a clonally heterogeneous population and which may participate in clonal evolution can present daunting challenges to physicians treating patients with this disease. Clearly, therapy imposes a strong selective pressure that affects the relative proportions of subclones harboring driver mutations within the malignant population. Thus, the dominant driver mutations in relapse samples are frequently already present in minor subclones before treatment. 13 These observations support the well-established “watch and wait” strategy of withholding treatment for asymptomatic CLL. However, a more thorough understanding of these interclonal dynamics might anticipate which subpopulations could become problematic in the future and thereby suggest approaches to personalize antecedent treatment. 71 Early detection of driver lesions in small subclones may potentially require surveillance using ultra-deep sequencing approaches, although validation of this strategy is lacking. In support of this idea, however, a recent study demonstrated that presence of minor subclones harboring TP53 mutations composed of as little as 2% of the cancer cells within a population in early CLL predated its emergence as a dominant subclone at relapse and the development of a chemo-refractory phenotype. 52

**Novel Therapeutic Targets**

To date, the only genetic features that have driven therapeutic decisions are deletions of 17p and TP53 alterations. 48 However, the relative high frequency of somatic mutations in specific genes indicates that certain pathways are likely essential to CLL and that targeting these gene pathways could provide effective therapy. Indeed, a number of agents targeting these genes and pathways are under development, and they include inhibitors of Notch signaling (γ-secretase inhibitors), 55 the spliceosome, 55 nuclear RNA export, 23 and telomerase. 55 Another potential new target in CLL includes activation-induced cytidine deaminase (AID), which drives somatic hypermutation in normal B-cell development by the generation of point mutations and initiation of double-strand breaks. 56,57 More than 40% of CLLs have marked expression of AID; in these cases, inhibition of homologous recombination in vitro preferentially induced apoptosis in leukemia cells. 58

Besides pointing to potentially vulnerable pathways in CLL, recurrently mutated genes in CLL have also recently been used to define the subclonal architecture of CLL. These studies have strongly suggested a specific temporal hierarchy in the acquisition of genetic and epigenetic changes, with clonal events (ie, “earlier”) enriched for alterations, which more selectively affect B cells. 13 These observations suggest that elimination of CLL early in disease course may be particularly effective, before its genetic diversification. Consistent with the idea that targeting the common “trunk” 59 might be an effective therapeutic strategy, clinical studies targeting B cell–specific pathways, such as by using anti-CD20 antibodies or novel kinase inhibitors of B-cell receptor signaling pathways, have demonstrated striking efficacy in CLL. 60

**FUTURE DIRECTIONS**

Within the same time frame that studies using massively parallel sequencing in CLL have been reported, analysis of other cancers using these technologies have been also completed. These datasets together with CLL data have yielded information regarding the fundamental mechanisms underlying somatic alteration in cancer and have also identified processes that are common or unique to CLL compared with other tumors. 36,57 Recent NGS efforts have examined the mutation spectrum of other mature B–cell lymphomas as well; not surprisingly, similar alterations appear across the malignancies, but CLL also has its own unique constellations of abnormalities (Figure 4). 8,11,13,15,17,61–69

Overall, these studies across cancers have yielded a number of further important insights. First, overall mutation rates of different malignancies have been established, and it is clear that CLL is a low mutation-rate tumor, similar to other leukemias, and is mutated up to 10-fold less than the carcinogen–induced cancers. 36,57,70 Second, examination of the spectrum of mutations and the organization of affected gene regions have revealed evidence that cancers occasionally undergo catastrophic shattering events (ie, so-called “chromothripsis”) as well as coordinated genomic rearrangements across different chromosomes. 71,72 Finally, comparison of CLL in relation to other cancers has revealed the spectrum of mutations in CLL to be consistent with a footprint of somatic hypermutation that is conventionally catalyzed by enzymes of the AID/APOBEC family of cytidine deaminases. 36,57

At this juncture, the continued advances in technology and analytic tools promise the ability to gain answers to a number of critical questions in the near term. First, by
Figure 4. Recurrent putative driver alterations in mature B-cell non-Hodgkin lymphomas. Abbreviations: BL, Burkitt’s lymphoma; CLL, chronic lymphocytic leukemia; DLBCL, diffuse large B-cell lymphoma; FL, follicular lymphoma; MCL, mantle cell lymphoma; MM, multiple myeloma assessed by NGS (TS, targeted sequencing [NGS or Sanger Sequencing], RNAseq, or SNP array; WES, whole-exome sequencing; WGS, whole-genome sequencing); SMZL, splenic marginal zone lymphoma; WM, Waldenström’s macroglobulinemia. Shown are all alterations that were significant in CLL in at least one study. For alterations in other entities, selected alterations have been validated and occurred in a significant and high proportion (>10% of cases) in at least one study or were identified across independent studies and were significant in at least one study.

<table>
<thead>
<tr>
<th>Mutated in % of samples:</th>
<th>MCL</th>
<th>FL</th>
<th>BL</th>
<th>DLBCL</th>
<th>SMZL</th>
<th>CLL</th>
<th>WM</th>
<th>MM</th>
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<tbody>
<tr>
<td>&lt; 3%</td>
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<tr>
<td>≥ 3%</td>
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### Mutated in % of samples:

- **TP53**: ++++
- **NOTCH1**: ++
- **SF3B1**: ++
- **ATM**: ++
- **MYD88**: +++
- **XPO1**: +
- **CHD2**: +
- **POT1**: +
- **SAMHD1**: +
- **FBXW7**: +
- **BIRC3**: ++
- **HIST1H1E**: ++
- **LRP1B**: ++
- **KLHL6**: ++
- **DDX3X**: ++++
- **ITK**: +
- **ERG2**: +
- **KRAS**: +
- **NRAS**: +
- **MED12**: +
- **ZMYM3**: +
- **BCOR**: +
- **RIPK1**: +
- **CCND1**: +
- **ML2**: ++
- **WHSC1**: +++
- **DAP10**: +++
- **TRMP6**: ++
- **MEF2B**: ++
- **NOTCH2**: ++
- **SP140**: ++
- **CREBBP**: ++++
- **KDM2B**: ++
- **EZH2**: +++
- **TNFRSF14**: +++
- **CARD11**: ++
- **TNAP**: ++
- **STAT6**: ++
- **EFP1**: ++
- **CD79B**: +
- **MYC**: +++
- **ID3**: ++
- **ML3**: +++
- **SMARCA4**: +++
- **CCND3**: ++
- **GNA13**: ++
- **EP300**: ++
- **PIM1**: ++
- **ARID1A**: ++
- **PRDM1**: ++
- **BCL2**: +++
- **TBL1X1R1**: ++
- **FAM46C**: ++
- **DIS3**: ++

**References**: 61, 62, 63, 64, 65, 67, 8, 11, 13, 15, 17, 68, 69
integration of genetic and transcriptional data, we can understand genotype and phenotype relationships. Distinct RNA expression modules have already been linked to CLL genotypes, and >60% of these associations were corroborated in an independent validation cohort yielding defined hypotheses for experimental studies.\(^7\) Second, the low mutation rate of CLL also suggests that we can likely gain comprehensive analysis of the mutation landscape in the near future with a saturating number of exomes.\(^6\) Third, single-cell sequencing technologies promise to yield opportunities to dissect genetic heterogeneity in CLL, further trace the evolutionary tree of individual cases, and strengthen understanding of genotype and phenotype relationships in this disease.\(^2\) Lastly, investigating CLL development by comparison between monoclonal B-cell lymphocytosis and CLL, as well as by learning how germ line variants predispose to CLL, might help us to further understand evolutionary trajectories in the disease. Recent meta-analyses of genome-wide population studies to examine CLL susceptibility have assessed on the order of magnitude of thousands of patients and control subjects. These efforts have corroborated 13 previously known risk loci for CLL and have newly found another 13 loci associated with inherited disease susceptibility. Interestingly, some of the identified genes suggest an overlap with regions of somatic driver alterations, involving \(BCL2\) or the 3' UTR of \(POT1\).\(^11,12\) Striking reports on CLL cases with germ line mutations in putative drivers exist for \(DAPK1\), a nuclease involved in DNA damage response;\(^7\) and for genes of the microRNA precursor \(miR-16-1\).\(^2\)

**CONCLUSIONS**

Recent technologic developments have yielded a breathtaking increase in our knowledge about the CLL genome. As we have reviewed herein, the insights gained from dissecting the nature and basis of clonal evolution in CLL fuel a myriad of further lines of investigation that can have concrete impact on the clinical practice and treatment of CLL. From a more basic biologic standpoint, future studies will likely address whether different genomic alterations provide distinct roles at defined stages in development of the disease, from predisposition to initial transformation of a clonal B-cell population and to progression with relapsed disease after treatment and even transformation.

**REFERENCES**