



Therapeutic targets and microenvironment in sequential biopsies of classical Hodgkin lymphoma at diagnosis and relapse

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Abstract

Classical Hodgkin lymphoma is dominated by the non-neoplastic microenvironment, while the neoplastic Hodgkin-Reed-Sternberg cells compose only a minority of cells in the lymphoma tissue. Both the Hodgkin-Reed-Sternberg cells due to their expression of CD30 and PD-L1 and the microenvironment with abundant T cells and expression of PD1 are specifically targeted by new treatment concepts. We aimed to understand the dynamics of therapeutic targets in patients treated with conventional chemotherapy. We analyzed sequential biopsy specimens obtained at diagnosis and at relapse from the same patient for morphology, immunophenotype, and microenvironmental components. The morphological subtype changed between primary and relapse biopsy in 20% of cases. The immunophenotype was stable with respect to CD30, CD3, and LMP1 but variable with respect to CD15 and CD20 expression. Gene expression revealed 8 upregulated and 20 downregulated genes at relapse ($p \leq 0.05$) with a consistent logarithmic fold change direction in at least 75% of all cases. For PD1, we found discrepant results between gene expression analysis (decrease at relapse) and number of PD1-positive cells assessed by immunohistochemistry (unchanged at relapse). PD-L1 in the neoplastic cells appeared unchanged between primary diagnosis and relapse. The expression of the therapeutic targets CD30, PD1, and PD-L1 can reliably be assessed in tumor specimen at first diagnosis and is unchanged under conventional chemotherapy.

Keywords Hodgkin lymphoma · Microenvironment · PDL1 · PD1 · CD30

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Introduction

Classical Hodgkin lymphoma (cHL) is characterized by an abundant microenvironment and a low number of neoplastic Hodgkin-Reed-Sternberg cells (HRSC). The composition of the microenvironment rather than the morphology or immunophenotype of the neoplastic HRSC is used to identify morphological variants of the disease, of which mixed cellularity (MC), nodular sclerosis (NS), and lymphocyte-rich are the most frequent [1]. The relative abundance of these subtypes strongly correlates with patient age at diagnosis and the socioeconomic status of the geographic area in which the patient resides [2]. The composition of the microenvironment in cHL is most likely a result of the interplay between the HRSC, which secrete multiple chemokines, and the microenvironmental cells, which provide survival signals to HRSC [3].

Although the histologically defined subtypes of cHL are not prognostically relevant under current treatment regimens [1], the microenvironment of cHL has gained considerable attention in molecular studies over the last years [4–7]. On

the one hand, gene expression profiling of the whole lymphoma tissue indicated a correlation between the cellular composition of the microenvironment and patient outcome and suggested that a high load of macrophages within the tissue is associated with a poor prognosis [4, 8]. On the other hand, blocking T cell signaling between programmed cell death 1 (PD1) expressed on T cells and PD1-ligand (PD-L1) led to exceptional therapeutic success [9].

Despite the fact that the phenotype and the histological subtypes of cHL have been known for decades, to the best of our knowledge, studies systematically analyzing the dynamics of these features are lacking. The content of macrophages in primary biopsies has been associated with outcome and is discussed as a relevant prognostic biomarker [10] but only few studies have been assessing the macrophage content in paired biopsies at diagnosis and at relapse [11]. Moreover, it is currently not known whether therapeutic targets like PD-L1/PD1 and CD30 change under conventional chemotherapy and which biopsies need to be assessed if the status should be assessed for targeted treatment at relapse. In fact, we recently showed that the number of PD1-positive cells increased in patients relapsing after/under anti-PD1 treatment [12]. We thus analyzed multiple biopsy specimens from cHL patients and compared histological subtype, immunophenotype, and microenvironment at diagnosis and at relapse to understand which tissue needs to be analyzed when assessing the presence of therapeutic targets CD30, PD1, and PD-L1 or prognostic biomarker such as the macrophage content in cHL.

Material and methods

Patients and tissue specimens

A test cohort of patients ($n = 18$) was identified in the files of the Lymph Node Registry Kiel on the basis of the availability of formalin-fixed paraffin-embedded (FFPE) tissue specimens of sufficient size and quality obtained at the time of diagnosis and relapse. Three patients were represented by more than two biopsies (total number of sequential biopsies $n = 39$). The morphology and immunophenotype including expression of CD30 of the test cohort were studied based on the available tissue sections. Additionally, gene expression and staining for CD68, CD163, PD1, and PD-L1 were performed using freshly cut FFPE sections. An independent validation cohort of sequential specimens was obtained from the files of the Departments of Pathology in Lübeck, Tübingen, and Ulm using the same procedure for patient identification as for the test cohort ($n = 14$ patients). The validation cohort was analyzed for morphology and immunophenotype (including CD30) using the available sections and for CD68 using freshly cut sections. The study was conducted in accordance with

the recommendations of the ethics board of the Medical Faculty, University of Kiel.

The median time between primary and relapse biopsy was 38 months (range 12 to 98) for the test cohort and 30 months (range 12 to 105) for the validation cohort. Due to the retrospective nature of the study, clinical data on the patients were not available.

Histopathology and immunophenotype

The histological subtype and the immunophenotype were assessed by at least one experienced hematopathologist (FF, CT, PM, WK). Stainings for CD20, CD30, CD15, CD3, PAX5, and LMP1 were performed using standard techniques in the participating labs and scored as positive if any fraction of HRSC stained positive. The following antibodies and staining protocols were used: CD68 (Dako M0876, clone PG-M1, pH 6, Histofine, manual staining), CD163 (Leica NCL-CD163, clone 10D6, pH 6, Histofine, manual staining), PD1 (Cell Marque 315M-96, clone MRQ22, ER1-pH 6, Leica-Bond stainer), PDL1 (Cell Signaling 13,684, clone E1L3N, ER2-pH 8 Leica-Bond stainer). The number of PD1, PD-L1, CD68, and CD163-positive cells was counted in three high-power fields (HPF, 1000-fold magnification using oil immersion). The average of the three HPF was calculated for further analysis. The HPFs were selected from areas containing HRSC. At least one unambiguously identified HRSC was placed in the center of the HPF in order to determine the cell count in very close proximity to the neoplastic cells. PD-L1 expression on malignant cells was analyzed by visual inspection and scored in steps of 0%, 1–25%, 26–50%, 51–75%, and > 76% positive HRSC.

Gene expression profiling

RNA was extracted from FFPE tissue sections using ExpressArt FFPE Clear RNA Ready Kit (AmpTec, Hamburg, Germany), as previously described [13]. Gene expression analysis was performed on a cohort of $n = 39$ specimens using the Nanostring (nCounter) technology and the Immunology Code Set Kit that target 594 genes associated with the immune system (nCounter, nanoString, Seattle, USA), including 15 housekeeper genes.

Sample quality control was assessed by using the quality metrics as implemented in the Bioconductor package *NanoStringQCPro* in version 1.8.0 [14]. Of the 39 specimens (18 primary tumors and 21 relapse samples), 8 specimens were excluded due to low/insufficient RNA abundance. As a consequence, only 13 lymphomas remained with good quality data for both the primary and the relapse sample. For one of the 13 lymphomas, we had two sequential relapse samples for the same primary sample. We included both such that we analyzed 14 pairs in total.

The gene expression data were normalized by first taking the log₂ on the raw data and then subtracting the individual median gene count from each sample. Next, the median of all sample medians was added to each specimen to ensure positive normalized count values for each gene.

Differential gene expression analysis was performed using generalized linear models of the negative binomial family as implemented in the Bioconductor package *NanoStringDiff* (version 1.6.0) [15–17]. Additionally, the paired nature of the cohort was included in the model and housekeeper genes ($n = 15$) were excluded from analysis. Resulting p values were adjusted for multiple testing according to Benjamini and Hochberg and a significance level for differential expression was set to a False Discovery Rate (FDR) of 0.05.

Results

Histological subtype, immunophenotype, and expression of CD30

In a combined analysis of the test and validation cohorts, the histological subtype remained stable between primary and relapse biopsy in the vast majority of cases (80%, Table 1). However, a change from mixed cellularity (MC) to nodular sclerosis (NS) subtype and vice versa as well as complex pattern subtype alterations over time was also observed (Table 1). A considerable number of cases varied in HRSC phenotype in sequential biopsy specimens with respect to the expression of CD15 (53%), CD20 (36%), and CD3 (6%) (Table 2, Fig. 1). Expression of CD30 and the EBV protein LMP1 were unchanged between primary and relapse biopsy in all cases analyzed (Table 2). CD30 expression was detectable on virtually all HRSC at first diagnosis as well as at relapse. We did not identify a case in which the CD30 expression was absent on a considerable number of HRSC (Fig. 1).

Macrophage content (CD68 and CD163)

We analyzed macrophage marker expression in the specimens using stainings for CD163. The frequency of CD163-positive

Table 1 Change of morphological subtype in sequential biopsies in all sequential specimens analyzed (MC = mixed cellularity, NS = nodular sclerosis, complex = multiple alterations in three available biopsy specimens: NS to lymphocyte depleted to NS, NS to MC to NS, and MC to lymphocyte rich to NS)

MC unchanged	10/30 (33%)
NS unchanged	14/30 (47%)
NS to MC	1/30 (3%)
MC to NS	2/30 (7%)
Complex	3/30 (10%)

macrophages did not show a significant difference between primary and relapse biopsies ($p = 0.3068$, paired t test). We additionally used CD68 as macrophage marker and extended the analysis to the validation cohort but failed to find a difference between primary and relapse biopsy (Fig. 2). Gene expression analysis using Nanostring technology in the test cohort revealed 8 upregulated and 20 downregulated genes at relapse ($p \leq 0.05$, Supplementary figure 1, Supplementary table 1) with a consistent logFC direction in at least 75% of all cases. MRNA expression CD163 not significantly changed between primary biopsy and relapse confirming the findings of immunohistochemistry analysis (Supplementary tables 2 and 3). However, we were able to detect the previously reported inverse correlation of relative changes in macrophage and B cell signatures between primary and relapse specimens (Supplementary figure 2).

PD1 and PD-L1

PD1-positive cells in the microenvironment were detected only in two specimens at a low level (16% in one primary biopsy and 2% in one relapse biopsy). None of the other specimens showed PD1-positive T cells in the 16 paired biopsies available for staining. HRSC were negative for PD1 in all cases. For 13 paired biopsies, PD-L1 staining yielded results in both the primary and the relapse biopsy (Fig. 1). PD-L1 expression in the microenvironment did not show any significant difference between primary biopsy and relapse ($p = 0.9946$, paired t test). More than 75% of HRSC were positive for PD-L1 in primary and relapse biopsy in 10 patients and between 50 and 75% positive HRSC in two patients. A slight change in PD-L1 expression was observed only in one patient in whom the primary biopsy showed 50–75% and the relapse > 75% PD-L1-positive HRSC.

By Nanostring gene analysis PD-L1 (CD274), mRNA expression was not different between primary and relapse confirming the results of immunohistochemistry. Of note, expression of PD1 (PDCD1) mRNA was significantly downregulated at relapse (Supplementary table 1).

Discussion

In cHL, the lymphomatous enlarged tissue is predominantly composed of a non-neoplastic microenvironment. The abundance of non-neoplastic cells allowed the development of prognostic signatures based on the composition of the microenvironment and the identification of therapeutic targets to [4–7, 18]. Especially the latter has attracted considerable attention during the last years [19]. The introduction of immune-checkpoint inhibitors that potentially interfere with the interaction of HRSC and microenvironmental cells has proven to be highly effective in cHL [9]. Since conventional

Table 2 Immunophenotypic alterations of Hodgkin-Reed-Sternberg cells in primary and relapse biopsy

	CD20	CD3	CD30	CD15	LMP1
Positive unchanged	0/17 (0%)	0/17 (0%)	17/17 (100%)	7/17 (41%)	9/17 (47%)
Positive to negative	4/17 (24%)	1/17 (6%)	0/17 (0%)	3/17 (18%)	0/17 (0%)
Negative to positive	2/17 (12%)	0/17 (0%)	0/17 (0%)	6/17 (35%)	0/17 (0%)
Negative unchanged	11/17 (65%)	17/18 (94%)	0/17 (0%)	1/17 (6%)	8/17 (53%)

chemotherapy cures the vast majority of cHL patients at initial disease presentation, the application of new drugs such as anti-

CD30 antibody-drug conjugates or immune-checkpoint inhibitors targeting the PD1-PD-L1 axis has been tested

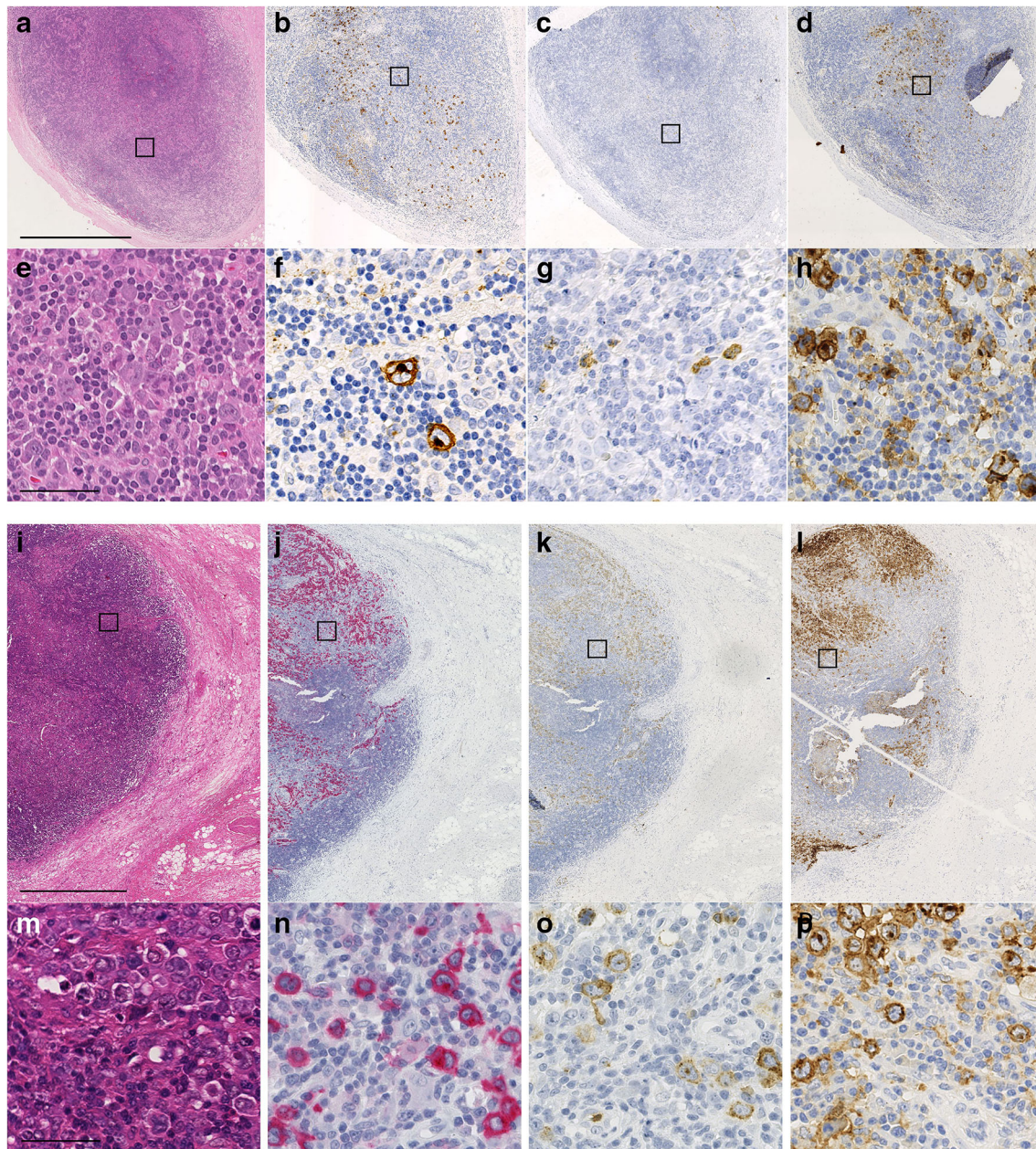
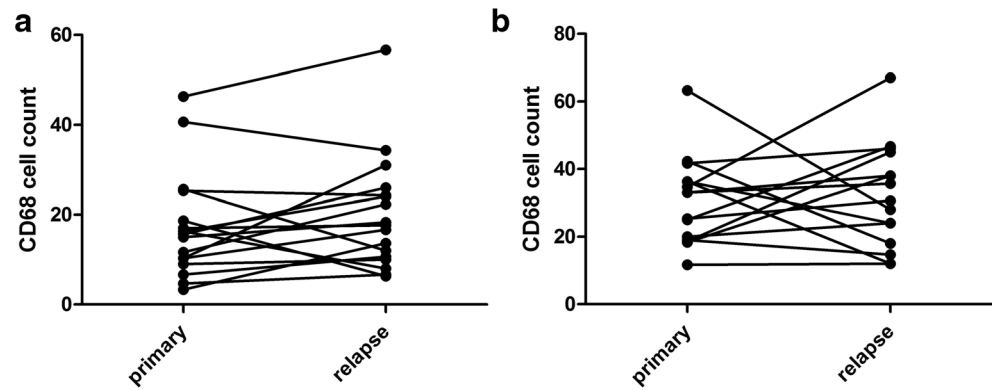


Fig. 1 Histology and immunophenotype. The upper panel represent primary (a–d low, e–h high magnification) and the lower panel relapse biopsies (i–l low and m–p high magnification). a, e, i, m = hematoxylin and eosin staining; b, f, j, n = CD30 (expression unchanged); c, g, k, o = CD15 (negative in primary and positive in relapse biopsy); d, h, l, p =

PDL1 (expression unchanged). Black boxes in low magnification indicate areas selected for high magnification of respective staining. The scale bar for low magnification represents 1000 μ m, scale bar for high magnifications 50 μ m

Fig. 2 Content of CD68-positive cells assessed immunohistochemically in sequential biopsy specimens. CD68 cell content in the test cohort (**a**, $p = 0.2403$) and validation cohort (**b**, $p = 0.7839$). Primary and relapse biopsy are connected by a line. Y-axis indicates average number of cells per three high-power fields



predominantly in relapsed patients [20]. Although there have been major advances in the therapy of relapsed cHL, biological studies of lymphoma tissue at relapse are very rare. Moreover, analyses of sequential biopsies of lymphomas at diagnosis and at relapse are virtually absent from the published literature until recently [11]. Therefore, we do not know so far whether predictive and prognostic biomarkers of cHL are stable between pretreatment tissue and relapse and whether the assessment of the biomarkers at initial diagnosis is valid for later relapses.

Similar to a previous report [11], we illustrate phenotypic shifts in cHL after chemotherapy exemplified in changes of the morphological subtype and the immunophenotype of HRSC. With respect to biomarkers, we found potentially relevant markers such as CD68/macrophage count to be unchanged between primary biopsy and relapse. Previously published studies suggested that CD30 expression might change in sequential biopsies [21–23]. In contrast to these previous publications, we demonstrate constant expression of CD30 between primary biopsy and relapse. Phenotype shifts between primary and relapse biopsies have been described in sequential mediastinal gray zone lymphoma [24]. However, our cohort only contains pure cHL. Of note, all biopsy specimen analyzed in our study were from the pre-brentuximab era. We cannot rule out that the variability of CD15 expression is to some extent due to technical reasons since expression can be weak or affect a subpopulation of cells only and might thus be missed by visual inspection.

PD1 and PD-L1 are both therapeutic targets and biomarker of outcome [25]. A recent publication found increasing levels of PD1-positive T cells in cHL at relapse after conventional chemotherapy [26]. We were not able to confirm these results. Both methods applied—immunohistochemistry and gene expression—did not detect an upregulation of PD1 at relapse. In fact, PD1 was downregulated on the RNA level at relapse and we did not find changes of PD1-positive cell counts at relapse. These data are in line with those obtained by another group that published PD1 levels in pairs of biopsies from primary diagnosis and relapse after conventional [27]. Unfortunately, the data of the so far only previously published

gene expression data on biopsy pairs of cHL are not publicly available to the best of our knowledge and thus cannot be used to confirm our results [11].

The discrepancy between gene expression and cell counts in our study may be explained in several ways. First, protein expression might be influenced by post-transcriptional mechanisms. Second, within a cell population, the expression of the identifying gene/protein might be variably high expressed but a broad spectrum of expression allows cells to be registered as positive by immunohistochemistry. PD1 has been shown to be downregulated in exhausted T cell which are primed to be reactivated [28]. We cannot exclude that this population is missed by our staining but detected by others [29]. Third, the methods of assessing cell counts used in our study are quantitative (counting) but restricted to a small preselected area in the immediate proximity of the HRSC (microenvironment) and do not necessarily reflect the cell content in the whole tissue (macroenvironment) as gene expression profiling does.

To gain further insight into the functional relevance of microenvironment changes at relapse, methods need to be applied that assess quantitatively cell counts and simultaneously allow to study the macro- and microenvironment within the tissue sections, e.g., by image analysis of whole scanned sections. Finally, spatial organization of cell types in combination with their abundance has been shown to provide relevant insight into the interaction between HRSC and the microenvironment in vivo [29]. Of note, our method of immunohistochemistry in fact is able to detect variability in PD1 expression as we have shown recently in patients suffering from a relapse after anti-PD1 treatment [12].

In summary, the number macrophages, a potential biomarker of outcome, remained stable between primary biopsy and relapse. The expression of therapeutic targets and potentially predictive biomarkers CD30, PD1, and PDL1 remain stable as assessed by immunohistochemistry between primary diagnosis and relapse. Thus, these biomarkers can be reliably assessed using the primary biopsy, e.g., if the biopsy confirming a relapse is not available or insufficient for ancillary studies. Of note, these findings

are restricted to patients treated with conventional chemotherapy and biomarker might well be more variable when patients are treated with immune-checkpoint therapy [12]. Furthermore, the findings should be regarded as preliminary since the patient cohort analyzed in our study is limited.

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Author contribution AS, CWK, KR, and MS performed the research; AS, MS, and WK designed the research study; CT, FF, BF, PM, and RS contributed essential reagents or tools; CWK, RS, SR, and WK analyzed the data; and WK wrote the paper.

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Compliance with ethical standards

The study was conducted in accordance with the recommendations of the ethics board of the Medical Faculty, University of Kiel.

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