High PTX3 expression is associated with a poor prognosis in diffuse large B-cell lymphoma

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Abbreviations: ABC/non-GCB, activated B cell-like; CSF1R, colony stimulating factor 1 receptor; DLBCL, diffuse large B-cell lymphoma; EBER, Epstein–Barr virus-encoded small RNA; EBV, Epstein–Barr virus; FFPE, formalin-fixed paraffin-embedded; FL, follicular lymphoma; GCB, germinal center B cell-like; HGBC, high-grade B-cell lymphoma; IHC, immunohistochemistry; IL-10, interleukin-10; IPI, International Prognostic Index; LDH, lactate dehydrogenase; MITF, microphthalmia transcription factor; NK, natural killer; NF-κB, nuclear factor-κB; NOS, not otherwise specified; OS, overall survival; PD-L1, programmed death-ligand 1; PFS, progression-free survival; PTX3, pentraxin 3; R-CHOP, rituximab, cyclophosphamide, hydroxydaunorubicin (doxorubicin), oncovin (vincristine), prednisone; RGS1, regulator of G-protein signaling 1; sIL-2R, soluble interleukin-2 receptor; TAM, tumor-associated macrophage; TNF, tumor necrosis factor; Treg, regulatory T lymphocyte.

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Abstract
Tumor-associated macrophages (TAMs) are associated with a poor prognosis of diffuse large B-cell lymphoma (DLBCL). As macrophages are heterogeneous, the immune polarization and their pathological role warrant further study. We characterized the microenvironment of DLBCL by immunohistochemistry in a training set of 132 cases, which included 10 Epstein–Barr virus-encoded small RNA (EBER)-positive and five high-grade B-cell lymphomas, with gene expression profiling in a representative subset of 37 cases. Diffuse large B-cell lymphoma had a differential infiltration of TAMs. The high infiltration of CD68 (pan-macrophages), CD16 (M1-like), CD163, pentraxin 3 (PTX3), and interleukin (IL)-10-positive macrophages (M2c-like) and low infiltration of FOXP3-positive regulatory T lymphocytes (Tregs) correlated with poor survival. Activated B cell-like DLBCL was associated with high CD16, CD163, PTX3, and IL-10, and EBER-positive DLBCL with high CD163 and PTX3. Programmed cell death-ligand 1 positively correlated with CD16, CD163, IL-10, and RGS1. In a multivariate analysis of overall survival, PTX3 and International Prognostic Index were identified as the most relevant variables. The gene expression analysis showed upregulation...
of genes involved in innate and adaptive immune responses and macrophage and Toll-like receptor pathways in high PTX3 cases. The prognostic relevance of PTX3 was confirmed in a validation set of 159 cases. Finally, in a series from Europe and North America (GSE10846, R-CHOP-like treatment, n = 233) high gene expression of PTX3 correlated with poor survival, and moderately with CSF1R, CD16, MITF, CD163, MYC, and RGS1. Therefore, the high infiltration of M2c-like immune regulatory macrophages and low infiltration of FOXP3-positive Tregs is associated with a poor prognosis in DLBCL, for which PTX3 is a new prognostic biomarker.

KEYWORDS
diffuse large B-cell lymphoma, IL-10, PD-L1, CD163, PTX3

1 | INTRODUCTION

Diffuse large B-cell lymphoma accounts for 30%-40% of newly diagnosed non-Hodgkin lymphomas. Gene expression profiling has classified DLBCL into the GCB subtype, which is associated with a better prognosis, and ABC/non-GCB subtype, which has a more aggressive clinical evolution. The poorer prognosis of patients with the ABC/non-GCB type is partly associated with the high infiltration of macrophages and the constitutive activation of the NF-κB pathway.

The macrophage lineage is heterogeneous. Classical or M1 macrophages are potent effector cells with pro-inflammatory and antitumoral functions. However, M2 macrophages are immunosuppressive and protumoral and are stratified into the M2a, M2b, and M2c subtypes. M2c macrophages have immune regulatory functions, express CD163, PTX3, and IL-10 markers, and induce Tregs.

Tumor-associated macrophages are also named M2d and are abundant in solid cancers, such as gastric, ovarian, breast, and lung adenocarcinomas. Although M1-like TAMs enhance antitumoral host immune responses, M2-like TAMs have been implicated in tumor progression, metastasis, resistance to therapy, angiogenesis, and immune suppression.

Pentraxin 3, also known as TNF-inducible gene 14 protein (TSG-14), is a protein that contributes to the regulation of innate resistance to pathogens, inflammatory reactions, and the clearance of self-components. Pentraxin 3 regulates the inflammatory activity of macrophages and it is expressed by macrophages with M2-like polarization, namely, the M2c-like subtype. In the context of IL-10 stimulation, B lymphocytes acquire regulatory properties. Pentraxin 3 makes a crucial contribution to tumor inflammation and is highly expressed in liposarcomas and lung and pancreatic carcinoma; however, its role in hematolymphoid neoplasia remains unclear.

Due to the importance of TAMs in the pathogenesis of solid and hematolymphoid neoplasia, the role of the IL-10 molecule in the regulation of host immune responses and immune checkpoints, the targeting of TAMs and the IL-10 regulatory pathway is an important strategy for DLBCL therapy.

In this study, we undertook IHC and a gene expression analysis of DLBCL samples collected from patients receiving R-CHOP therapy to investigate the role of macrophages and Tregs in the pathogenesis of DLBCL and elucidate their impact on clinical outcomes. The results obtained revealed that PTX3 is a powerful marker that identifies patients with different survival outcomes.

2 | MATERIALS AND METHODS

The training set comprised 132 cases of DLBCL diagnosed according to the 2016 WHO criteria. Cases were collected from Tokai University Hospital between 2004 and 2011. This study was carried out according to the Declaration of Helsinki of Ethical Principles for Medical Research Involving Human Subjects (IRB 14R-080).

This series of patients had conventional DLBCL characteristics. Detailed information is shown in Tables 1, 2, and S1. Staging maneuvers and the assessment of treatment responses were standard. The median age of patients was 69 years (range, 14-97 years) and the male / female ratio was 1.54. The IPI was retrospectively assessed in 106 patients (80%): low risk (37, 34.9%), low-intermediate risk (36, 34%), high-intermediate risk (17, 16%), and high risk (16, 12.1%). Treatments were R-CHOP in 73.2% of cases and R-CHOP-like (mainly RCOP) in 22.0%; only 4.9% had other treatments (such as RT, CHOP, and THP-COP). Among the 118 patients with assessable responses, 89 (75.4%) achieved a complete response, 12 (10.2%) progressive disease, 16 (13.6%) a partial response, and 1 (0.8%) stable disease. One-, 3-, 5-, and 10-year OS rates were 85%, 68%, 59%, and 46%, respectively.

The validation set included 159 cases of DLBCL diagnosed between 2009 and 2012. The clinicopathologic characteristics of these cases are shown in Table S2. In summary, this series was compatible with a conventional series of DLBCL NOS, with 73% of cases being older than 60 years, 37.7% nodal, 44.9% with IPI high-intermediate and high, and 96.6% treated with rituximab. One-, 3-, 5-, and 10-year OS rates were 75%, 59%, 53%, and 41%, respectively.
TABLE 1 Main pathological characteristics in patients with diffuse large B-cell lymphoma (training set)

| Characteristic             | Frequency (%) |
|----------------------------|---------------|
| Histological               |               |
| CD3+                       | 0s (1.6)      |
| CD5+                       | 12/129 (9.3)  |
| CD20+                      | 129/131 (98.5)|
| CD10+                      | 39/130 (30.0)|
| BCL6+                      | 88/129 (68.2)|
| MUM1+                      | 89/129 (69.0)|
| Non-GCB (Hans classifier)  | 82/129 (63.6)|
| BCL2+                      | 89/129 (69.0)|
| EBV EBER+                  | 10/125 (7.6)  |
| RGS1-high                  | 64/108 (59.3)|
| MYC-high                   | 62/119 (52.1)|

Molecular

| Marker                        | Frequency (%) |
|-------------------------------|---------------|
| BCL2 split (FISH)             | 13/121 (10.7)|
| MYC split (FISH)              | 13/124 (10.5)|
| BCL2 split (FISH)             | 18/106 (17.0)|
| BCL2 and MYC split (FISH)     | 2/119 (1.7)   |
| BCL2 IHC+ and MYC-IHC-high    | 39/123 (31.7)|
| MYC+t and/or MYC-IHC-high     | 67/132 (50.8)|
| High grade B-cell lymphoma    | 5/125 (4.0)   |
| MYD88 L265P mutation          | 12/121 (9.9)  |

Note: The cut-off for positivity for the immunohistochemical (IHC) markers for B lymphocytes of DLBCL using the Hans algorithm (CD10, BCL6, IRF4/MUM1) was set at the conventional 30%. BCL2 was considered positive if more than 50% of the tumor cells were positive. High regulator of G-protein signaling 1 (RGS1) expression associates with poor prognosis in DLBCL. RGS1-high corresponds to expression 2-3+ as we have previously described. MYC translocation positive cases positively correlated with higher IHC expression for MYC protein (Fisher’s exact test, P = .003); MYC-high cut-off was 40%, and after digital image quantification was set at 22%.

Validation of the PTX3 gene marker in an independent series of DLBCL from Europe and North America was undertaken in GSE10846, which is publicly available in the NCBI database.

2.2 Phenotypic characterization of DLBCL samples

Immunohistochemical analysis was carried out on FFPE whole tissue sections using automatic equipment (Leica Bond-Max system and reagents; Leica Biosystems). Immunophenotypes included the markers of CD3, CD5, CD20, CD10, MUM1 (IRF4), BCL2, BCL6, and Ki-67 (Novocastra/Leica Biosystems), RGS1 (rabbit polyclonal; Thermo Fisher Scientific), and MTIF (CS/D5/MAB10775; Abnova). We used the standard 30% cut-off for the positivity of the Hans classifiers CD10, BCL6, and IRF4/MUM1. BCL2 was considered positive if 50% or more of the tumor cells were positive, and MYC if 40% or more of the tumor nuclei were positive. The macrophagic signature included the pan-macrophage markers of CD68 (514H12), M1-like CD16 (2H7), M2-like CD163 (10D6), PTX3 (PPZ1228; Perseus Proteomics), and IL-10 (LS-B7432; Lifespan Bioscience). We used FOXP3 to identify Tregs (236A; Cell Signaling Technology) and CSF1R (#FER216; CNIO) IHC were recovered from our recent previous publications (https://doi.org/10.3390/ai201 0008 and https://doi.org/10.3390/hemat02020011, respectively).


2.3 | Gene expression analysis

Gene expression profiling of a representative set of 37 cases was undertaken using RNA extracted from FFPE samples. The nCounter Immuno-oncology and Lymph2Cx assay panels were used (NanoString Technology). Housekeeping gene normalization was calculated using the log2((normalData[i]/hkGeomMeans[i])) formula.

2.4 | Statistical analysis

All statistical analyses were undertaken using SPSS software (version 26; IBM). The χ² and/or Fisher's exact tests and the Mann–Whitney U test were used for group comparisons, and the Kaplan–Meier and log-rank tests and Cox’s regression analysis for survival analyses. Overall survival was defined from the date of diagnosis to the last contact date. Progression-free survival was defined from the date of diagnosis to disease progression. Bivariate correlation was carried out using Pearson and Spearman's tests. The significance level was set at .05. 24–29

TABLE 3  Distribution of markers in the series of patients with diffuse large B-cell lymphoma (training set)

| Marker | Frequency | Low | High |
|--------|-----------|-----|------|
|        | Mean ±SD  | Median | Best OS cut-off | No % | No % |
| CD68   | 15.2 8.4  | 13.5 28.0 | 120.0 92.3 | 10.0 7.7 |
| CD16   | 6.2 9.6  | 0.9 0.7 | 56.0 45.9 | 66.0 54.1 |
| MITF   | 2.9 2.8  | 2.1 3.1 | 81.0 66.4 | 41.0 33.6 |
| CD163  | 22.7 19.4 | 17.9 20.5 | 75.0 57.3 | 56.0 42.7 |
| PTX3   | 11.0 17.0 | 1.5 49.0 | 125.0 95.4 | 6.0 4.6 |
| IL-10  | 8.3 9.9  | 4.9 9.6 | 39.0 66.1 | 20.0 33.9 |
| FOXP3  | 2.2 2.8  | 1.0 4.5 | 111.0 84.1 | 21.0 15.9 |
| RGS1   | 9.3 11.7 | 4.8 3.0 | 44.0 40.7 | 64.0 59.3 |

Note: The immune microenvironment markers were initially evaluated as an ordinal variable a 0, 1+, 2+, and 3+ as <1%, 1%-5%, 5%-20%, and >20%, respectively, under the optical microscope. After digitalization the percentage of positive cells was quantified using Fiji software. The best cut-off for the overall survival (OS) was found from the quantitative data (ie, the most significant P value). Figure 1 shows immunohistochemical images of the different immune markers with the evaluation reference under the microscope and a characteristic image of high values for each marker that were associated with poor OS.

Abbreviations: IL-10, interleukin-10; MITF, microphthalmia transcription factor; PTX3, pentraxin 3; RGS1, regulator of G-protein signaling 1.

3 | RESULTS

3.1 | Clinical and histological features of patients in the training set

Detailed information is shown in Tables 1, 2, and S1. The most relevant histological features of this series were as follows: DLBCL was positive for CD5 in 9.3% of cases, BCL2 in 69.0%, and EBER in 7.6%, and RGS1 was highly expressed in 59.3% of cases. The cell-of-origin analysis based on the Hans classifier showed non-GCB in 63.6% of cases. Molecular analyses revealed the MYC FISH split signal in 10.5% of cases and the MYD88 L265P mutation in 9.9%. Clinical variables that were associated with unfavorable OS were a patient age older than 60 years, high LDH, high sIL-2R, more than one extranodal site, stage III–IV, and IPI high-intermediate and high. Similar results were obtained for PFS.

According to the revised 4th Edition of the WHO Classification of Tumours of Haematopoietic and Lymphoid Tissues, the HGBL with MYC and BCL2 and/or BCL6 rearrangements category therefore includes: (a) double-hit cases previously classified as B-cell lymphoma, unclassifiable, with features intermediate between DLBCL and Burkitt lymphoma; (b) blastoid cases with a double hit; and (c) cases with a DLBCL NOS morphology that, upon evaluation, have rearrangements of MYC and BCL2 and/or BCL6. In the training series, we identified five cases of DLBCL NOS morphology that, following FISH evaluation, were reclassified into HGBCL (Table 1). Also, in the training set, we identified 10 cases that were positive for EBV by in situ hybridization. Those cases fell within the WHO category of EBV-positive DLBCL NOS.

3.2 | Distribution of markers in DLBCL and relationships with histopathological features in the training set

The frequencies of all markers are shown in Table 3, including means ±SD and medians. Characteristic images for each marker are shown in Figure 1. The expression of the markers was low, with a mean percentage of less than 20% of all cells, except CD163, which had a frequency of 22.7% ± 19.4%. The following correlations were observed with pathological features (Table 4): (a) the non-GCB subtype expressed high CD16, CD163, PTX3, and IL-10; (b) EBER-positive cases expressed high CD163 and low PTX3; (c) HGBCL had higher CD163 but lower PTX3; and (d) double-expressor (ie, BCL2-positive, MYC above 22% by digital quantification) by high CD16 (Table S3).
The markers were also correlated with the double-expressor, MYD88 L265P mutation, and cell-of-origin in DLBCL NOS (Tables S3-S5).

The immunohistochemical analysis of PD-L1 showed a positive correlation with CD16, CD163, IL-10, and RGS1 (Table S6).

### 3.3 Relationships with OS and PFS in the training set

The best (ie, optimal) cut-off for each marker was defined based on the P value in the log-rank test for OS. On average, between all markers, 31.1% of cases were in the poor prognosis group (Table 3).

Detailed information is shown in Table 3. Overall survival figures are shown in Figure 2. In the univariate analysis, the markers associated with poor OS were macrophages with high expression of CD68, CD16, MITF, CD163, PTX3, and IL-10, and low infiltration of FOXP3+ Tregs. In the multivariate analysis of all of these markers and IPI, only PTX3 and IPI retained their significance (Table 5). Therefore, PTX3 was the most relevant marker in this model in addition to IPI.

Similar results were obtained for PFS (Table 5 and Figure S1). In the univariate analysis, the high expression of CD68, CD16, CD163, PTX3, and IL-10 correlated with unfavorable PFS.

The analysis was repeated for all markers stratifying for the DLBCL NOS, EBER+ DLBCL, and HGBCL subtypes (Figure S1 and Table 6). The results showed that the previous findings in DLBCL were also confirmed in the DLBCL NOS group.

### 3.4 Gene expression analysis in the training set

The results of the gene expression analysis are shown in Figure 3 and Table S7. The 38 genes in the Lymph2Cx assay panel were merged with the 770 genes in the immuno-oncology panel. In 37
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representative cases, gene expression was compared between two groups based on the IHC expression of PTX3: high vs low, using a cut-off of 10%. In this analysis, we identified which genes and pathways differed between the two groups to elucidate the role of PTX3 in the pathogenesis of DLBCL. Of the 37 cases, 4 were EBER+ DLBCL and 33 DLBCL NOS.

FIGURE 2  Relationship between markers and overall survival (OS) in the training set of patients with diffuse large B-cell lymphoma (DLBCL). High expression of macrophage markers was associated with poor prognosis, including OS and progression-free survival (PFS). High infiltration of regulatory T lymphocytes was associated with a favorable prognosis (OS). Of note, the analysis was repeated stratifying for DLBCL not otherwise specified (NOS), Epstein–Barr virus-encoded small RNA+ DLBCL, and high-grade B-cell lymphoma, both for OS and PFS. In the DLBCL NOS group, the same results were found. Cum., cumulative; GCB, germinal center B cell-like
Overall survival

- **FOXP3**
  - High
  - Low
  - \( P = .002 \)

- **CD68**
  - High
  - Low
  - \( P = .002 \)

- **MITF**
  - High
  - Low
  - \( P = .020 \)

- **CD163**
  - High
  - Low
  - \( P = .002 \)

- **PTX3**
  - High
  - Low
  - \( P = .020 \)

- **IL-10**
  - High
  - Low
  - \( P = .017 \)

- **Hans+PTX3**
  - Non-GCB PTX3 low
  - Non-GCB PTX3 high
  - \( P = .017 \)

- **Hans+IL10**
  - Non-GCB IL10 low
  - Non-GCB IL10 high
  - \( P = .031 \)
Nonvariant genes were excluded from the analysis, and significant genes were ranked according to \( P \) values and group associations. Additionally, genes were annotated by the following 20 immune response categories: adhesion, antigen processing, B-cell functions, cell cycle, cell functions, chemokines, complement, cytokines, cytotoxicity, interleukins, leukocyte functions, macrophage functions, microglial functions, NK cell functions, pathogen defense regulation, senescence, T-cell functions, Toll-like receptor, TNF superfamily, and transporter functions (Table S7). Genes were also annotated using the STRING database of known and predicted protein–protein interactions. Interactions included direct (physical) and indirect (functional) associations.

In Figure 3, the gene expression is shown using a relative color scheme. The relative color scheme used the minimum and maximum values in each row to convert values to colors and allowed for the easier visualization of differences between the two groups.

Eighteen genes were identified in the low PTX3 group. The most significantly expressed genes were \( \text{TGFB2} \) (interleukins, \( P = .009 \)), \( \text{KIR2DS1} \) (regulation and NK cell function, \( P = .009 \)), \( \text{PPIA} \) (signaling, apoptosis, HK, \( P = .012 \)), \( \text{IL22RA1} \) (chemokine, \( P = .016 \)), and \( \text{ITGB4} \) (adhesion, \( P = .017 \)). This group had a higher number of genes involved in regulation, interleukin, adhesion, and NK cell functions.

Twenty-one genes were identified in the high PTX3 group. The most significantly expressed genes were \( \text{CYFIP2} \) (innate immune response, \( P = .002 \)), \( \text{NUBP1} \) (signaling, apoptosis, HK, \( P = .002 \)), \( \text{CD53} \) (adaptive immune response, \( P = .008 \)), \( \text{IL2RG} \) (chemokines, \( P = .008 \)), and \( \text{IKBKE} \) (innate immune response, \( P = .011 \)). Other genes of interest were \( \text{BTK} \) (adaptive immune response, \( P = .021 \)), \( \text{SYK} \) (macrophage function, \( P = .023 \)), and \( \text{MAP2K1} \) (innate immune response, \( P = .028 \)). \( \text{MYD88} \) (innate immune response, Toll-like receptor) was slightly overexpressed in this group (\( P = .128 \)). This group had higher numbers of genes involved in antigen processing.

### Table 5: Correlation between markers and outcome in patients with diffuse large B-cell lymphoma

| Marker  | OS P value | HR  | 95% CI     | PFS P value | HR  | 95% CI     |
|---------|------------|-----|------------|-------------|-----|------------|
| **Univariate analysis**          |           |     |            |             |     |            |
| CD68   | .002       | 3.2 | 1.5        | .72         |      |            |
| CD16   | .049\textsuperscript{a} | 1.5 | 0.9        | 2.6         |      |            |
| MITF   | .020       | 1.9 | 1.1        | 3.2         |      |            |
| CD163  | .002       | 2.1 | 1.3        | 3.5         |      |            |
| PTX3   | .020       | 2.8 | 1.1        | 7.1         |      |            |
| IL-10  | .017       | 2.4 | 1.1        | 5.0         |      |            |
| FOXP3  | .002       | 0.1 | 0.0        | 0.6         |      |            |
| **Multivariate analysis\textsuperscript{b}** |           |     |            |             |     |            |
| CD68   | .457       | 2.7 | 0.2        | 37.4        |      |            |
| CD16   | .604       | 0.7 | 0.2        | 2.8         |      |            |
| MITF   | .386       | 1.8 | 0.5        | 6.5         |      |            |
| CD163  | .829       | 1.2 | 0.3        | 5.3         |      |            |
| PTX3   | .033       | 6.1 | 1.2        | 31.9        |      |            |
| IL-10  | .287       | 2.1 | 0.5        | 8.0         |      |            |
| FOXP3  | .146       | 0.3 | 0.1        | 1.5         |      |            |
| IPI    | .035       | 2.0 | 1.1        | 3.9         |      |            |
| **Multivariate analysis\textsuperscript{c}** |           |     |            |             |     |            |
| PTX3   | .020       | 4.5 | 1.3        | 18.2        |      |            |
| IL-10  | .009       | 7.3 | 1.7        | 15.6        |      |            |
| FOXP3  | .111       | 0.3 | 0.1        | 1.3         |      |            |
| EBER   | .023       | 4.8 | 4.8        | 1.2         |      |            |
| CD68   | –          | –   | –          | –           |      |            |

**Abbreviations:** –, no data; CI, confidence interval; HR, hazard ratio; OS, overall survival; PFS, progression-free survival.

\textsuperscript{a}Log-rank (Mantel-Cox) \( P = .136 \); Breslow (generalized Wilcoxon) \( P = .049 \).

\textsuperscript{b}Multivariate analysis, method = enter.

\textsuperscript{c}Multivariate analysis, Cox regression, method = backward conditional, predictors = CD68, CD16, microphthalmia transcription factor (MITF), CD163, pentraxin 3 (PTX3), interleukin-10 (IL-10), FOXP3, International Prognostic Index (IPI), cell-of-origin (Hans classifier), Epstein-Barr virus-encoded small RNA (EBER), and high-grade B-cell lymphoma genotype.
and adaptive and innate immune responses. IKKBKE, BTK, SYK, and MYD88 formed a cluster with the biological processes Gene Ontology (GO) of innate immune responses and the Toll-like receptor signaling pathway. Using the STRING database, the 21 genes associated with the high PTX3 group corresponded to the pathway for the positive regulation of the immune system, NF-κB signaling, MHC II, and B-cell receptors. In contrast, the 27 genes associated with the low PTX3 group were involved in the defense response pathway, NK cytotoxicity, MHC I, and cytokine signaling (Figure 3).

3.5 | Evaluation of the PTX3 marker in the validation set

The correlation between PTX3 and the survival outcomes of patients was validated in an independent series from Tokai University Hospital (Figure 4). The clinicopathologic characteristics of this validation series are shown in Table S2. The immunohistochemical analysis of PTX3 showed that high expression correlated with poor OS (high [n = 28] vs low expression [n = 120], hazard risk = 2.4, $P = .003$; 95% CI, 1.4-4.2). Additionally, PTX3 stratified the non-GCB subtype, as shown in Figure S2 (hazard risk = 1.8, $P = .004; 95\% \text{ CI, 1.2-2.8}$). In the Figure S2A-B, the analysis also included the stratification for EBER- and EBER+ DLBCL cases, and OS (A) and PFS (B).

3.6 | Validation of PTX3 marker in an independent series

The publicly available GSE10846 DLBCL gene expression dataset was used to validate the association of PTX3 and poor OS of patients. This series comprises 414 cases; 181 (44%) were treated with CHOP-like therapy, and 233 (56%) with R-CHOP-like therapy. The R-CHOP-like cases were comparable to the Tokai series. In concurrence, we found high expression of PTX3 in the R-CHOP-like therapy group was associated with poor OS of patients: 5-year OS, 76% (95% CI, 85%-67%) vs 55% (95% CI, 68%-42%), Kaplan–Meier with log-rank test, $P = .001$; hazard risk = 2.3 (95% CI, 1.4-3.8, $P = .001$) (Figure 5).

High PTX3 expression also correlated with other clinical variables including ECOG performance status of 2 or higher, LDH ratio greater than 1, and National Comprehensive Cancer Network-IPI high-intermediate/high (all $P$ values ≤0.01) (Table 7). Interestingly, at RNA levels, PTX3 positively correlated with CD16, MITF, CD163, CSF1R, and RGS1 (Table 8).

4 | DISCUSSION

In this study, we showed that: (a) the tumor infiltration of macrophages in DLBCL differed and the expression of macrophage markers varied; (b) the high expression of macrophage markers and low infiltration of FOXP3+ Tregs were associated with a poor prognosis in DLBCL; (c) PTX3 and IL-10 both played an important role in the pathogenesis of DLBCL and were associated with a poor prognosis; and (iv) gene expression differed between the high vs low PTX3 groups.

We examined various factors in a large series of DLBCL from Japan. IPI including age, LDH, clinical stage, and extranodal sites as well as sIL-2R and the lack of a clinical response correlated with shorter OS. The biological factors associated with a poor prognosis were MUM1 positivity, the non-GCB phenotype (Hans classifier), and EBER positivity (Table S1). The present results are consistent with the previously reported findings from Europe and North America; therefore, our series corresponds to a conventional series of DLBCL.
Tumor-associated macrophages are abundant in solid cancers; M1-like TAMs enhance antitumoral host immune responses, whereas M2-like TAMs have been implicated in tumor progression. Survival outcomes correlate with TAMs in hematological malignancies. A high infiltration of CD68-positive TAMs has been associated with a poor prognosis in patients with FL. CD68-positive TAMs, specifically CD163-positive M2-like TAMs, have been associated with a poor prognosis in classical Hodgkin lymphomas and the progression of adult T-cell leukemia/lymphoma, cutaneous T-cell lymphoma, and T-cell/reticulum cell sarcoma. Gene expression profiling showed that the ABC subtype of DLBCL was characterized by the activation of the NF-κB pathway and by markers of macrophages. This study confirmed the prognostic relevance of TAMs in de novo DLBCL, including pan-macrophages (CD68), M1-like (CD16), and M2-like (CD163, ITGAM, PTX3, and IL-10) TAMs, and FOXP3+ Tregs. The high expression of CD68, CD16, CD163, PTX3, and IL-10, but low expression of FOXP3 correlated with poor OS. The high expression of CD16 also correlated with poor OS, but only during the first years of observation (Kaplan-Meier and Breslow statistical tests, $P = .049$). The cell-of-origin and presence of EBV also influenced the prognosis of DLBCL patients. The non-GCB type had higher expression of CD16, CD163, PTX3,

**FIGURE 3** Gene expression analysis of the training set of patients with diffuse large B-cell lymphoma (DLBCL). A set of 37 cases of DLBCL from Tokai University Hospital was selected and a gene expression analysis was carried out using Lymph2Cx and immuno-oncology NanoString panels. Nonvariant genes between the two groups were excluded and significant genes were ranked according to $P$ values and group associations. The heatmap visualization shows how gene expression between high and low pentraxin 3 (PTX3) groups differed, indicating a different pathogenic mechanism. A relative color scheme uses the minimum and maximum values in each row to convert values to colors and allows for the easier visualization of differences between the two groups. Genes involved in innate and adaptive immune responses and macrophage and Toll-like receptor pathways, including CYFIP2, NUBP1, CD53, IL2RG, IKBKE, BTK, SYK, MAP2K1, and MYD88, were upregulated in the high PTX3 group. Genes involved in regulation, interleukins, adhesion, and natural killer (NK) cell function pathways, including TFF3, KIR2DS1, PPIA, IL22RA1, and ITGB4, were upregulated in the low PTX3 group. Using the STRING database, the 21 genes associated with the high PTX3 group (top right panel) corresponded to the pathways positively regulating the immune system, nuclear factor-κB signaling, MHC II, and B-cell receptors. In contrast, the 27 genes associated with the low PTX3 group (bottom right panel) corresponded to the defense response pathway, NK cytotoxicity, MHC I, and cytokine signaling.
and IL-10. The EBER-positive cases also more highly expressed CD163 and PTX3. Regulator of G-protein signaling 1 is a marker for the architecture of immune tissues and lymphocyte migration and is associated with a poor prognosis in DLBCL.21,41,42 The present results point out a relationship between RGS1 and the composition of the immune microenvironment. Regulator of G-protein signaling 1 correlated with the expression of CD163 and PTX3 (P < .05), and with FOXP3 (P = .062).

A relationship has already been reported between the expression of CD163 and poor prognosis in DLBCL43,44; however, we here revealed correlations with other macrophage markers representing different differentiation states. We examined several markers of pan-macrophages, M1-like, and M2-like polarization, and Tregs. High infiltration of macrophages but low infiltration of Tregs were associated with poor prognosis. In the multivariate Cox regression analysis of OS between all macrophage markers and FOXP3, PTX3 and IL-10 retained their prognostic value. Therefore, these markers are the most relevant in this prognostic model.

We determined the prognostic relevance of PTX3, which will be an important biological marker in lymphoma. Pentraxin 3 is a marker of immune regulatory M2c-like polarization and NF-κB activation.5-8 The non-GCB type can be stratified according to PTX3 expression levels: (a) GCB; (b) non-GCB PTX3 low; and (c) non-GCB PTX3 high. In comparisons with GCB, non-GCB PTX3 high was associated with a poorer prognosis, with a hazard risk of 3.9 (P = .007). Although a previous study reported that PTX3 was associated with a poor prognosis in pancreatic carcinoma,13 this is the first study to show this relationship in hematological malignancies.

We identified the IL-10 marker. Interleukin-10 is a major immune regulatory cytokine that acts on many cells in the immune system, in which it exerts marked antiinflammatory effects, thereby limiting excessive tissue disruption by inflammation. It downregulates the expression of Th1 cytokines, MHC class II antigens, and costimulatory molecules on macrophages. It also promotes B-cell survival, proliferation, and Ab production. Interleukin-10 contributes to peripherally derived Tregs.45 It is also a marker of macrophage M2c polarization and many of the M2c-specific genes identified to date have been implicated in angiogenesis, matrix remodeling, and phagocytosis, including CD163, MMP8, TIMP1, VCAN, SERPINA1, MARCO, PLOD2, PCOCLE2, and F5.46 Interleukin-10 also contributes to the pathogenesis and/or development of autoimmune diseases and cancer.47 We recently reported the expression of IL-10 in the DLBCL subtype of methotrexate-associated lymphoproliferative disorder in patients with rheumatoid arthritis22; however, its role in DLBCL NOS had not yet been investigated by our group. In this study, IHC revealed that IL-10 marked cells with a macrophage morphology. In
TABLE 7 Correlation between clinicopathological characteristics and PTX3 gene expression (GSE10846 dataset) in diffuse large B-cell lymphoma (DLBCL)

| Variable                      | PTX3 |     |     |  P  value |
|-------------------------------|------|-----|-----|----------|
| Diagnosis                     |      |     |     |          |
| DLBCL                         | High | Low |     |          |
| Tissue                        |      |     |     |          |
| Lymph node                    |      |     |     |          |
| Treatment                     |      |     |     |          |
| R-CHOP-like                   |      |     |     |          |
| No.                           |      |     |     |          |
| 79/233 (34)                   | 154/233 (66) |     |          |
| Sex                           |      |     |     |          |
| Female                        |     |     |     |          |
| 69/99 (70)                    | 30/99 (30) |     | .318    |
| Male                          |     |     |     |          |
| 85/134 (63)                   | 49/134 (37) |     |          |
| Age, y                        |      |     |     |          |
| ≤60                           |     |     |     |          |
| 72/113 (64)                   | 41/113 (36) |     | .457    |
| >60                           |     |     |     |          |
| 82/120 (68)                   | 38/120 (32) |     |          |
| Ann Arbor stage               |      |     |     |          |
| I- II                         |     |     |     |          |
| 78/105 (74)                   | 27/105 (26) |     | .014    |
| III- IV                       |     |     |     |          |
| 71/121 (59)                   | 50/121 (41) |     |          |
| ECOG performance status       |      |     |     |          |
| <2                            |     |     |     |          |
| 116/158 (73)                  | 42/158 (27) |     | <.001   |
| ≥2                            |     |     |     |          |
| 21/52 (40)                    | 31/52 (60) |     |          |
| LDH ratio                     |      |     |     |          |
| <1                            |     |     |     |          |
| 77/99 (78)                    | 22/99 (22) |     | <.001   |
| >1                            |     |     |     |          |
| 48/93 (52)                    | 45/93 (48) |     |          |
| >1 extranodal site            |      |     |     |          |
| 0-1                           |     |     |     |          |
| 117/173 (67)                  | 56/173 (32) |     | .062    |
| >1                           |     |     |     |          |
| 15/30 (50)                    | 15/30 (50) |     |          |
| NCCN-IPI                      |      |     |     |          |
| Low-slow-intermediate risk    |     |     |     |          |
| 73/101 (72)                   | 28/101 (28) |     | .010    |
| High+high-intermediate risk   |     |     |     |          |
| 33/63 (52)                    | 30/63 (48) |     |          |
| Cell of origin                |      |     |     |          |
| GCB                           |     |     |     |          |
| 71/107 (66)                   | 36/107 (34) |     | .938    |
| ABC+ unclassified             |     |     |     |          |
| 83/126 (66)                   | 43/126 (34) |     |          |

Note: Data are shown n/N (%).
Abbreviations: –, no data; ABC, activated B-cell-like; GCB, germinal center B-cell-like; LDH, lactate dehydrogenase; NCCN-IPI, National Comprehensive Cancer Network-International Prognostic Index; R-CHOP, rituximab, cyclophosphamide, hydroxydaunorubicin (doxorubicin), oncovin (vincristine), prednisone.

FL, the high expression of FOXP3 was associated with a good prognosis,19 which is consistent with the present results for DLBCL.

We examined the differential gene expression between cases with high and low expression of PTX3, because PTX3 is a novel marker of prognosis and pathogenesis of DLBCL. We found that expression profiles differed between the two groups. Genes involved in innate and adaptive immune responses and macrophage and Toll-like receptor pathways, including CYFIP2, NUBP1, CD53, IL2RG, IKBKE, BTX, SYK, MAP2K1, and MYD88, were upregulated in the high PTX3 group. Genes involved in regulation, interleukins, adhesion, and NK cell function pathways, including TFFB2, KIR2DS1, PPIA, IL22RA1, and ITGB4, were upregulated in the low PTX3 group.

The analysis of PTX3 in a validation series of cases confirmed the correlation with OS in DLBCL patients, including the stratification of the Hans classifier. Therefore, PTX3 is a relevant marker for the pathogenesis of DLBCL.

Epstein–Barr virus produces a specific cytotoxic T cell immunity response,48,49 that plays an important role in controlling the proliferation of virus-infected cells. Cytotoxic T cells detect EBV peptides bound to MHC molecules in the surface of affected cells, and eliminate these cells by direct lysis. Conversely, the presence of EBV within the lymphoma cell acts as a potential target, using mechanisms of immune evasion. Indeed, EBV-positive DLBCL cases are associated with changes in the tumoral immune microenvironment. Histological analysis shows high infiltration of macrophages that express high levels of PD-L1 and indoleamine 2,3-dioxygenase50,51, associated with changes in the tumoral immune microenvironment.

We have recently described the importance of PD-L1, CSF1R, TNFAIP8, and CASP8 in the pathogenesis and prognosis of DLBCL, and shown how systems biology can help improve the understanding of disease mechanisms.59-65 In this research, we showed how PD-L1 correlated with CD16, RGS1, and IL-10. The relationship with PTX3 will be analyzed in future publications.
Finally, we used the gene expression data of the GSE10846 dataset, which included 233 cases of DLBCL patients treated with a R-CHOP-like therapy. Using this independent series, we confirmed that high gene expression of PTX3 correlated with poor OS of patients.

In conclusion, we showed that the M2c immune regulatory pathway is associated with an unfavorable prognosis in de novo DLBCL. These results provide novel insights into the pathogenesis of DLBCL with applicability to current and future clinical trials.

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CONFLICT OF INTEREST
None to declare.

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