New MYC IHC Classifier Integrating Quantitative Architecture Parameters to Predict MYC Gene Translocation in Diffuse Large B-Cell Lymphoma

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Abstract: A new automated MYC IHC classifier based on bivariate logistic regression is presented. The predictor relies on image analysis developed with the open-source ImageJ platform. From a histologic section immunostained for MYC protein, 2 dimensionless quantitative variables are extracted: (a) relative distance between nuclei positive for MYC IHC based on euclidean minimum spanning tree graph and (b) coefficient of variation of the MYC IHC stain intensity among MYC IHC-positive nuclei. Distance between positive nuclei is suggested to inversely correlate MYC gene rearrangement status, whereas coefficient of variation is suggested to inversely correlate biological regulation of MYC protein expression. The bivariate classifier was compared with 2 other MYC IHC classifiers (based on percentage of MYC IHC positive nuclei), all tested on 113 lymphomas including mostly diffuse large B-cell lymphomas with known MYC fluorescent in situ hybridization (FISH) status. The bivariate classifier strongly outperformed the “permutation of MYC IHC-positive nuclei” methods to predict MYC+ FISH status with 100% sensitivity (95% confidence interval, 94–100) associated with 80% specificity. The test is rapidly performed and might at a minimum provide primary IHC screening for MYC gene rearrangement status in diffuse large B-cell lymphomas. Furthermore, as this bivariate classifier actually predicts “permanent overexpressed MYC protein status,” it might identify non-translocation-related chromosomal anomalies missed by FISH.

Key Words: MYC, IHC, Burkitt lymphoma, DLBCL, image analysis, minimum spanning tree graph

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MYC is the master regulator of nearly all important cellular functions, especially growth regulation and cellular metabolism. As of 2003, an integrated database of genes responsive to MYC transcription factors had 647 entries clustered in different functional groups, including 120 genes related to metabolism, 82 related to protein synthesis, 25 associated with cell cycle control, and 5 involved in DNA replication. Cell division requires considerable metabolic support and the importance of MYC is illustrated by its ability to coordinate both. MYC was initially identified as an oncogene with the demonstration of the reciprocal translocation between MYC and immunoglobulin heavy loci t(8;14)(q24;q32) (MYC+) in Burkitt lymphoma. In recent years, different reports have demonstrated MYC+ in a small subset (3% to 16%) of diffuse large B-cell lymphomas (DLBCL). When MYC+ is associated with another translocation such as t(14;18)(q32;q21) (BCL2+), the involved DLBCL is labeled as double hit (DH). DH (MYC+/BCL2+) prevalence is also imprecise and varies between 0% and 12% in recent studies. Synergistic action of MYC+ and BCL2+ confers a dismal outcome and requires more aggressive therapy. DH also refers to other associations, such as MYC+ and BCL6 t(3;16)(q27;p11) (BCL6+), which also predicts a dismal outcome. Rare triple-hit B-cell lymphomas (MYC+/BCL2+/BCL6+) are also reported.

Currently, MYC+ is assessed with fluorescent in situ hybridization (FISH). There are 2 different FISH approaches to detect MYC rearrangement, which are the dual-fusion and the break-apart techniques. The former is less sensitive to detect MYC+ especially when the translocation involves a non-IGH partner; the latter detects MYC independently of the translocation partner. The break-apart technique might, however, miss cryptic rearrangements that should be detected by the dual-fusion technique. Consequently, utilization of both techniques may increase MYC+ detection sensitivity. There are other mechanisms that convey MYC protein overexpression in DLBCL: MYC gene amplification and
micro-RNA alteration\textsuperscript{14,15} are relatively frequent alterations. Whereas amplification can be detected by FISH (ideally pending specific probes\textsuperscript{16}), micro-RNA alteration is not. Nonetheless, these dysregulated MYC DLBCL, not associated with chromosomal translocation, appear to be associated with similar negative clinical implications.\textsuperscript{17} Therefore, the utilization of MYC IHC might be an alternative to FISH: actually, many investigations, assessing the percentage of nuclei IHC positive for MYC protein, have concluded that MYC IHC is capable of predicting MYC\textsuperscript{+}.\textsuperscript{7,18–21} These studies, however, report a variety of thresholds (30\%, 40\%, and 50\%) to predict MYC\textsuperscript{+}. Of interest, in at least 2 studies,\textsuperscript{18,22} authors warn readership about potential interobserver and intraobserver variations in MYC IHC scoring, especially when the IHC stain shows heterogeneity. These observations are the basis of the current work wherein we suggest a new approach to MYC IHC assessment.

The aforementioned crucial roles of MYC in cellular growth and metabolism require tight regulation of MYC gene expression to maintain cellular homeostasis. Among safeguard mechanisms, both MYC mRNA and MYC protein have short life (\textsim 30 \text{min} for the protein) which are effective means of gene regulation.\textsuperscript{1,23} It is hypothesized that persistence and loss of MYC gene regulation are, respectively, associated with modulated and constant protein expression with observable IHC architectural modifications. First, the amount of MYC protein per nucleus depends on MYC regulation: in permanently upregulated MYC neoplasms, the MYC IHC stain is expected to be strong and, more important, similar among all malignant nuclei; the opposite situation is regulated MYC expression, wherein both spatial and temporal variation in MYC IHC stain reflect physiological cycling of MYC. Hence, MYC stain variability is hypothesized to be proportional to MYC gene regulation. Second, the spatial distribution of MYC IHC-positive nuclei is regulation dependent: in permanently upregulated MYC neoplasms, positive MYC IHC nuclei appear gregarious or “clonal”; in regulated MYC neoplasms, positive nuclei are randomly scattered among negative ones reflecting persistence of normal regulation patterns within the tissue. Relative distance between positive IHC MYC nuclei is thus hypothesized to be proportional to gene regulation (Fig. 1). Combined stain variability and relative distance between MYC IHC-positive nuclei will be compared with the currently used “percentage of positive IHC MYC nuclei” to predict MYC\textsuperscript{+}.

**METHODOLOGY**

**Dimensionless Variables as Predictors of MYC Gene Status**

Coefficient of variation (CV = \sigma/\nu) of MYC IHC stain intensity is defined as the division of the SD (\sigma) of stain intensity by the mean (\nu) of IHC stain intensity measured in MYC IHC-positive nuclei. CV and \sigma are both measurements of dispersion, but the former has the advantage to be dimensionless. Relative distance between positive MYC IHC nuclei is obtained by utilizing a second dimensionless variable defined as the division of the average distance between positive IHC MYC nuclei by the average distance between all nuclei. Distance between nuclei is the distance between closest neighbor. This latter is obtained with the utilization of euclidean minimum spanning tree\textsuperscript{24} (EMST), which provides for each nucleus

**FIGURE 1.** A, DLBCL MYC\textsuperscript{–} with scattered positive MYC IHC nuclei and variable stain intensity. B, DLBCL MYC\textsuperscript{+} with gregarious positive MYC IHC nuclei and less variable stain intensity. Both distance between positive nuclei and stain variability are hypothesized to be directly proportional to MYC gene regulation (Fig. 2).
the distance of the closest neighbor. EMST is assessed twice: first for positive MYC IHC nuclei only and a secondly for all nuclei. A ratio EMST is then computed by dividing the former EMST by the latter. The larger ratio EMST, the larger the number of MYC IHC myc negative nuclei intercalated between positive ones. CV and ratio EMST are the 2 suggested dimensionless and independent variables which will be used to predict the categorical-dependent variable “MYC gene rearrangement status.”

MYC IHC

IHC was performed on 4μm formalin-fixed paraffin-embedded tissues sections that were baked for 1 hour at 60°C. Slides were stained using the Ventana XT platform and Optiview detection kit (760-700; Ventana Medical Systems Inc., Tucson, AZ) after retrieval with CC1 (pH 8.5) for 72 minutes. A rabbit monoclonal anti-MYC (1/50, EP121, 395R-15; Cell Marque, Rocklin, CA) was incubated for 40 minutes at 37°C. Slides were counterstained using

FIGURE 2. The 4 pictures represent the same MYC IHC preparation (DLBCL). (A) Detection of nuclei using “find maxima” algorithm, (B) superimposed threshold image of DAB component (red), (C) euclidean minimum spanning tree (all nuclei), (D) euclidean minimum spanning tree (positive MYC IHC nuclei). CV of stain intensity is measured in positive MYC IHC nuclei and ratio EMST computed from division of “average edge length” from D over “average edge length” from C.
Gill’s Hematoxylin (380150; Leica Biosystems, Richmond, IL) for 15 seconds. An onslide control consisting of BL (MYC+), DLBCL (MYC+/C0), and melanoma was used to ensure consistent staining in and between staining runs.

### Image Acquisition, Analysis, and Algorithm Description

Image analysis was performed using the ImageJ platform, a public domain, Java-based image processing program developed at the National Institutes of Health (Bethesda, MD). To get CV and ratio EMST, 1 pathologist (G.B.) digitized 1 field on each of the 113 MYC IHC slides using a ×20 objective targeting the visually most intensely stained area. To get percentages of MYC IHC-positive nuclei, another pathologist (W.-F.D.) digitized 2 fields from each of the same 113 slides using ×20 and ×40 objectives and again targeting the visually most intensely stained area. Two different microscopes were utilized: (G.B.) Nikon Eclipse E600 microscope, 0.25 aperture (×20) and 0.65 (×40) and (W.-F.D.) Nikon Eclipse 80i microscope, 0.5 aperture (×20) and 0.75 (×40) (Nikon Instruments Inc., Melville, NY). Both microscopes are equipped with a QImaging Micropublisher 5.0 RTV camera (QImaging Corp., Surrey, BC), which uses a Sony ICX282 progressive scan interline CCD producing 24-bit color pictures with a resolution of 2560 × 1920 pixels. For all images, a priori background correction was applied. Percentage of positive MYC IHC nuclei were computed with Immunoratio, an open source software based on the ImageJ platform. CV and ratio EMST were computed using an in-house developed algorithm (G.B.) using combination of ImageJ macros and plugins. In summary, for each MYC IHC image, a “find maxima” algorithm provides (x,y) coordinates (centroids) of all nuclei (Fig. 2A); color deconvolution plugin isolates DAB staining which permits identification of positive nuclei (Fig. 2B); a first EMST is built with centroids from all nuclei (Fig. 2C) and a second EMST is built with centroids from positive MYC IHC nuclei (Fig. 2D). DAB intensity is measured in positive nuclei averaging 9 samples per nucleus from a 3 × 3 matrix whose center is the centroid (x,y). Finally, CV is computed from all measured DAB intensities in positive MYC IHC nuclei.

### Case Selection

A total of 113 lymphomas (19 BL, 77 DLBCL, 6 intermediate between BL and DLBCL, and 11 unclassified aggressive B-cell lymphomas) diagnosed between 2010 and 2015 with known MYC status were
selected. MYC IHC stains were produced in 2014 and 2015. All specimens had been fixed in formalin (37% formaldehyde in aqueous solution) and embedded in paraffin. Although a small group of cases (~20%) was studied retrospectively, all cases in this analysis were stained utilizing freshly cut thin sections.

**Interphase MYC FISH Analysis**

A nuclear preparation was made from paraffin-embedded tissue. Two 20 μm curls (contiguous to sections used for IHC) were cut from paraffin block. The curls were deparaffinized, hydrated in ethanol, digested in 10% proteinase K, washed, and the resulting nuclear suspension deposited onto glass slides. FISH was performed with the VYSIS LSI MYC Dual Color Break-apart rearrangement probe set (Abbott Molecular, Des Plaines, IL) that hybridizes to the band region 8q24. This probe detects the most prevalent translocation t(8;14)(q24;q32)IGH-MYC involving the heavy-chain immunoglobulin locus but also detects 2 additional variants t(2;8)(p11.2;q24.1)IGK-MYC and t(8;22)(q24.1;q11.2)IGL-MYC involving light-chain immunoglobulin loci. The established normal cutoff for a positive result is 10% of the cells showing an abnormal signal pattern. Appropriate negative and positive controls were used. An average of 100 interphase nuclei were examined independently by 2 observers using an Olympus Provis microscope system on ×1000 magnification.

**MYC IHC Stain and Variables Predictors Stability**

Immunohistochemistry technique relies on numerous preanalytical and analytical variables that influence the immunoreactivity of proteins. As some antibodies require freshly diluted antibody or freshly cut unstained slides, the stability of MYC antibody dilution over 6 weeks was verified as well as antigen preservation within unstained thin section over time. The same onslide controls [1 BL (MYC+) and 1 DLCBL (MYC−)] were simultaneously submitted to variable preanalytical conditions including incremental delay after block sectioning, various aging antibody dilution, or combination of both. The goal was to see whether the MYC IHC stain would remain stable within a timeframe of 6 weeks, and if not, to measure the influence, if any, of stain variability on ratio EMST, CV, and percentage of positive MYC IHC nuclei.

**Statistical Analysis**

Statistical analysis and statistical figures were created using R language (version 3.2.2; R Foundation, Vienna). To predict MYC gene rearrangement, simple logistic regression model was individually applied on the 2 variables: percentage ×20 and percentage ×40; multiple logistic regression was applied to the bivariate (CV, ratio EMST). To estimate how accurate the 3 models would perform, an exhaustive cross-validation (Leave-one-out cross-validation) was applied to each model. The statistical null hypothesis is
that the probability of $MYC^+$ is not associated with the aforementioned variables. Statistical significance was evaluated with Wald $Z$-statistic test. Performance of predictors was assessed with receiver operating characteristic (ROC) and accuracy performance using the ROCR package. To compare ROC curves of the 3 different predictors and to measure the confidence intervals (CI) of sensitivity and specificity, the pROC R package was utilized. CI assessment relied on 2000 stratified bootstrap replicates. A $P$-value of $\leq 0.05$ was selected as the level of significance in all analyses. Figure 2 was created with the FigureJ plugin. This study was approved by the Alberta Cancer Research Ethics Committee (file CC-15.0247).

**RESULTS**

**Predictive Capacity of Variables**

Fifty cases were $MYC^+$ (19 BL, 21 DLBCL, and 10 others) and 63 were $MYC^-$ (0 BL, 56 DLBCL, and 7 others) by FISH. For each predictor, computation of individual variables took $<2$ minutes. Descriptive statistics of the total number of harvested nuclei per case are: mean (3105), median (3022), first quartile (2349), third quartile (3768), and range (1266 to 5945). $MYC$ gene rearrangement classification of the 113 cases is illustrated in Figures 3 and 4, where visual separation of $MYC^+$ and $MYC^-$ is obvious for all predictive models. The null hypothesis is rejected: in all logistic regression analyses, all variables were statistically significant ($< 0.05$). For the 3 models, probability of $MYC^+$ are, respectively, computed as follows:

\[
\frac{1}{1+e^{-(3.6676+0.09668 \times \% 20X)}}
\]

\[
\frac{1}{1+e^{-(4.2325+0.09932 \times \% 40X)}}
\]

and

\[
\frac{1}{1+e^{-(24.267-6.2209 \times \text{ratio EMST}-0.5033 \times \text{CV})}}
\]

If all 3 models can predict $MYC$ status, visual inspection suggests that the multiple logistic regression using bivariate (ratio EMST, CV) is superior than individual simple logistic models. Superiority of bivariate model is confirmed by inspecting superimposed ROC curves (Fig. 5), where areas under the curve are 97.1% (95% CI, 0.95-0.99), 90.6 (95% CI, 0.85-0.96), and 90.5 (95% CI, 0.85-0.96) for, respectively (ratio EMST, CV), $\times 40$ and $\times 20$. Inspection of curves of test accuracy and associated density distribution of $MYC^+$ and $MYC^-$ clearly shows the superiority of the multiple logistic regression model (Fig. 6).

Figure 7 illustrates the expected sensitivities and specificities of the 3 predictive models obtained by selecting different cutoffs. These results are in line with previous figures. The multivariate logistic regression model shows the possibility to maintain a virtual 100% sensitivity associated with strong specificity: choosing the
cutoff of 11% of probability, the associated sensitivity and specificity are 100% (95% CI, 94%-100%) and 80 (95% CI, 76%-95%). The 2 other simple logistic regressions models can have 100% sensitivity but only with the cost of a much lower specificity.

MYC IHC Stain and Variables Predictors Stability

Stain intensity drop was found to be proportional to ages of both antibody dilution and unstained tissue section. Both freshly prepared antibody dilution and freshly cut thin section are required to maintain constant high MYC IHC stain intensity over time. Of interest, the extent of stain variability depends on the amount of MYC antigen within the tissue. For the MYC+ control, the average stain intensity per nucleus varies approximately in a range of 1:1.5 (Fig. 8A, y-axis); for the MYC− control, the average stain intensity per nucleus varies approximately in a range of 1:7.5 (Fig. 8B, y-axis). Computed predictors variables vary differently: for the MYC+ control, percentage of positive MYC IHC nuclei, CV, and ratio EMST are stable and are barely influenced by stain intensity (Fig. 8A, x-axis). In contrast, for the MYC− control, percentage of positive MYC IHC nuclei varies with stain intensity, whereas CV and ratio EMST remain relatively stable (Fig. 8B, x-axis).

DISCUSSION

A new MYC IHC classifier to predict MYC gene rearrangement status in DLBCL using the dimensionless variables ratio EMST and CV is presented. On the basis of automated image analysis, this classifier provides fast assessment with potentially minimal interobserver variability. Careful inspection of Figure 2A shows that centroid detection by the “find maxima” algorithm is reasonably precise, but not perfect: with thousands of nuclei harvested by the algorithm, the test has shown statistical power. The choice of dimensionless variables (ratio EMST and CV) should allow for the production of comparable results independent of the utilized image analysis/microscope environment.

In order for this classifier to partially replace or supplement FISH by triaging DLBLC, the test would need to meet class II biomarker requirements. The reproducibility of the MYC IHC assay used in this study requires freshly prepared antibody dilution and freshly cut unstained slides. Hence, it might then also be necessary to define preanalytical guidelines (similar to those described for the consistent performance of biomarker IHC in breast cancer,34,35 that ideally ensure reliable hormone receptor results between laboratories). This new MYC IHC classifier integrates 2 independent variables that are based on IHC stain variability (CV) and morphologic data. It was
suggested that CV is a measure of the maintenance of physiological cycling protein expression. The integration of IHC information with a quantitated architecture parameter is thought to explain the superior classification capacity compared with currently used assessments based on the percentage of MYC IHC-positive nuclei alone. As suggested by one of the reviewers of this manuscript, careful documentation of lymphomas with unusual IHC staining, for instance heterogenous MYC staining, should be documented with both IHC and FISH techniques. Furthermore, reproducibility of this method should be demonstrated in other laboratories: to this end, the software is available from the Web site GitHub: https://github.com/gilbertbigras/MYC-IHC.

This paper also suggests that CV and ratio EMST are more robust variables than percentage of MYC IHC-positive nuclei, especially when preanalytical conditions create IHC stain variability. CV robustness would be explained as follows: even if the average IHC stain intensity of nuclei varies (secondary to preanalytical conditions), stain intensity dispersion around a given average (measured by CV) would not, as every single nucleus stain intensity is shifted with the average. Furthermore, even if stain intensity variation directly influences the number of positive nuclei and consequently the percentage, the average edge length forming the EMST graph is not. It has been demonstrated that neoplastic 2D architecture can be quantified by different graphs such as EMST and Voronoi. Computed parameters derived from these graphs show stability after including a certain number of centroids. This is particularly relevant for tissue with a small number of nuclei positive for MYC and/or low content of MYC antigen per nucleus (Fig. 8B, MYC –). With high content of MYC antigen per nucleus (Fig. 8A, MYC +), the preanalytical deficiencies do not affect the stability of variables.

If the classifier strongly predicts MYC gene rearrangement status, it would be more precise to suggest that it predicts permanent dysregulated MYC protein overexpression. Therefore, if a MYC + status confers to a DLBLC a permanent dysregulated MYC protein overexpression, the opposite is not always true: as already discussed there are other molecular anomalies which are MYC –. This might explain the presence of some MYC – (green circles) found in “positive” territory (Figs. 3, 4). If these cases are confirmed to be associated with dismal outcomes, bivariate MYC IHC classifier might outperform FISH. New investigation should, however, also demonstrate that these cases are associated with molecular anomalies not related to chromosomal translocation.

FIGURE 7. Receiver operating characteristics curves of the 3 logistic regression models (similar to Fig. 5) with cutoff probabilities: for the 3 models, it is possible to determine a cutoff probability with 100% sensitivity. For the ×20, ×40, and bivariate (ratio EMST, CV) these cutoffs are, respectively, 8%, 5%, and 17%. The associated specificities are, however, significantly different and are approximately 40%, 25%, and 85%.
In conclusion, we propose that this new IHC analytical method that combines IHC, image analysis, and graph theory could be utilized to triage DLCBL. Providing a probability cutoff of 11%, any DLBLC with a probability <11% could be considered MYC/C0 and FISH would not be necessary. In contrast, any DLBLC with a probability ≥11% should be investigated with FISH to confirm MYC gene rearrangement, with subsequent BCL2 and BCL6 FISH assessments to identify DH and/or triple-hit DLBLC. A significant number of patients could be safely triaged as MYC− at relatively low cost and with quick turnaround time.

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