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POSTER ABSTRACTS

508.BONE MARROW FAILURE: ACQUIRED

Predicting Clonal Evolution to Secondary Myeloid Neoplasms in Aplastic Anemia through Machine Learning

Ahmet Celal Toprak, B.S.¹, Taha Bat, MD², Mutlu Mete, PhD³, Carmelo Gurnari, MD⁴, Zoe Bass, B.S.A.¹, Jaroslaw P. Maciejewski, MD, PhD, FACP⁴, Hussein Awada⁴, Alper Olcal, MD², Ibrahim F. Ibrahim, MD⁵

¹University of Texas Southwestern Medical Center, Dallas, TX

²Department of Hematology and Oncology, University of Texas Southwestern Medical Center, Dallas, TX

³Department of Computer Science and Information Systems, Texas A&M University - Commerce, Commerce, TX

⁴Department of Translational Hematology and Oncology Research, Taussig Cancer Institute, Cleveland Clinic, Cleveland, OH

⁵Department of Hematology and Oncology, UT Southwestern Medical Center, Dallas, TX

Background:

One of the major long-term complications in aplastic anemia (AA) is the clonal evolution to secondary myeloid neoplasms (sMNs) such as acute myeloid leukemia (AML) and myelodysplastic syndrome (MDS)¹. Hematopoietic stem cell transplantation (HSCT) is largely considered a curative treatment option for AA and is effective at preventing the clonal evolution of sMNs². However, patients over 40 years old or those with poor donor options are contraindicated for HSCT due to elevated post-transplant mortality rates³. When treated with non-transplant strategies, patients with AA have indeed a high risk for acquisition/expansion of somatic myeloid mutations, with some harbingers of future clonal progression to full-blown MDS or AML. Older age, presence of myeloid mutations other than *BCOR/L* and PIGA at onset, and lack of response to standard immunosuppressive treatment (IST) have been identified as factors predictive for subsequent sMN evolution ^{4,5,6}. If an individual patient's risk of developing a sMN could be assessed through a predictive model that uses information collected at initial AA diagnosis, providers may consider HSCT for traditionally contraindicated patients. However, no such predictive model that could aid in risk assessment currently exists.

Methods:

In our study, we utilized a large, multi-institutional cohort of patients diagnosed with AA to generate a comprehensive dataset. This dataset contained 76 variables, encompassing relevant clinical and molecular annotations. On this dataset, we trained a machine learning (ML) model to predict patients who would evolve sMNs. Variables with the highest feature impact score (FIS) in predicting clonal evolution to sMN were then chosen to be included in the final ML model, and the algorithm's fitness was calculated using the Leave-One-Out cross-validation method. Variables missing in more than 30% of subjects were imputed using K-Nearest Neighbors (N=20).

Of our initial cohort of 455 patients, 213 patients were excluded from the study. 35 patients were excluded because they were treated with HSCT. 39 pediatric cases (defined as <18 years old at the time of diagnosis) were excluded from the study. 91 patients were removed for not receiving specific treatment (IST or HSCT). The remaining 77 patients were excluded due to having a sMN detected at the time of or prior to AA diagnosis, insufficient data, or detection of a germline mutation indicative of an inherited pathophysiology. Our final cohort included a total of 242 acquired AA patients from UT Southwestern Medical Center and Cleveland Clinic.

Results:

Among our final cohort of 242 patients, 119 (49.2%) were male. The cohort's mean and median ages at AA diagnosis were 53.1 years (SD = 18.9 years) and 58.1 years, respectively. The median follow-up time was 4.04 years. 80 (33.1%) of our patients had a clinically significant PNH clone (defined as >0.001%). 29 patients developed sMNs, with 24 developing MDS and 5 developing AML. Of patients who underwent clonal evolution to sMNs, the median age at the time of AA diagnosis and sMN diagnosis was 63.1 years and 67.2 years, respectively. The median time elapsed from initial AA diagnosis to sMN evolution was 3.78 years. The median time elapsed between initial AA diagnosis and earliest IST administration was 51 days. Chi-squared analysis showed that patients who received ATG treatment were more likely to develop sMNs, with 78.0% of all patients receiving ATG, while 96.6% of patients who developed sMNs had received ATG (p=0.00927).

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Our ML model achieved a sensitivity of 0.759, a specificity of 0.728, and an AUC of 0.78 using the ten variables with the highest feature impact score (FIS) to predict clonal evolution [Table 1]. All of the variables except for the presence of a PNH clone are positively correlated with sMN evolution. Nine of these variables are supported by existing literature ^{7,8}. The feature with the third highest feature impact score, the *time elapsed* (days) between diagnosis and IST treatment, has not been established by previous studies and may be an important consideration in establishing optimal AA treatment guidelines.

Conclusion:

The strategic combination of this ML algorithm with clinical expertise has tremendous potential in improving health outcomes. Identification of patients at high risk for sMN and offering HSCT as a curative option upfront may lead to improved overall survival in AA patients.

Disclosures Bat: Alexion pharma: Membership on an entity's Board of Directors or advisory committees. **Maciejewski:** Novartis: Honoraria, Speakers Bureau; Alexion: Membership on an entity's Board of Directors or advisory committees; Regeneron: Consultancy, Honoraria; Omeros: Consultancy.

Variable	Feature Impact Score	Mean ± Standard Deviation		
		All Patients (N=242)	Negative (N=213)	Positive (N=29)
Age at diagnosis (years)	1.18	53.13 ±18.89	52.29 ± 19.42	59.34 ± 12.81
Meets SAA or VSAA according to Camitta criteria (1/0)	0.95	0.68 ±0.23	0.67 ± 0.23	0.76 ± 0.21
Time elapsed from AA diagnosis to IST treatment (days)	0.79	178.62 ± 215.83	171.87± 168.10	228.16 ± 422.36
DNMT3A mutation detected at diagnosis (1/0)	0.75	0.03 ± 0.14	0.01 ± 0.10	0.12 ± 0.30
TET2 mutation detected at diagnosis (1/0)	0.61	0.03 ± 0.14	0.02 ± 0.12	0.09 ± 0.25
PNH clone present (1/0)	0.60	0.39 ± 0.45	0.41 ± 0.46	0.23 ± 0.38
RUNX1 mutation detected at diagnosis (1/0)	0.55	0.03 ± 0.14	0.02 ± 0.12	0.08 ± 0.25
TP53 mutation detected at diagnosis (1/0)	0.51	0.02 ± 0.13	0.01 ± 0.10	0.08 ± 0.25
U2AF1 mutation detected at diagnosis (1/0)	0.46	0.02 ± 0.13	0.02 ± 0.10	0.08 ± 0.25
Patient received ATG treatment (1/0)	0.45	0.78 ± 0.41	0.76 ± 0.43	0.97 ± 0.18

Table 1: Ten Most Significant Variables Ranked by Impact Scores. The variables are ranked by their feature impact scores in relation to the development of sMN in patients. The patient groups are categorized as "Negative" for those who did not evolve sMN and "Positive" for those who did. Mean values and standard deviations for each variable are provided for the respective patient groups. Boolean values were numerically represented as 1 or 0, where a value of 1 indicates the presence of a variable, and a value of 0 indicates its absence. *Abbreviations: AA = Aplastic anemia. ATG = Antithymocyte globulin. IST = Immunosuppressive therapy. NSAA = Non-severe aplastic anemia. PNH = Paroxysmal Nocturnal Hemoglobinuria.*

Figure 1

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