Correlation Analysis and Prognostic Impacts of Biological Characteristics in Elderly Patients with Acute Myeloid Leukemia

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Background: The significant heterogeneity of elderly AML patients’ biological features has caused stratification difficulties and adverse prognosis. This paper did a correlation study between their genetic mutations, clinical features, and prognosis to further stratify them.

Methods: 90 newly diagnosed elderly acute myeloid leukemia (AML) patients (aged ≥60 years) who detected genetic mutations by next-generation sequencing (NGS) were enrolled between April 2015 and March 2021 in our medical center.

Results: A total of 29 genetic mutations were identified in 82 patients among 90 cases with a frequency of 91.1%. DNMT3A, BCOR, U2AF1, and BCORL1 mutations were unevenly distributed among different FAB classifications (p < 0.05). DNMT3A, IDH2, NPM1, FLT3-ITD, ASXL1, IDH1, SRSF2, BCOR, NRAS, RUNX1, U2AF1, MPO, and WT1 mutations were distributed differently when an immunophenotype was expressed or not expressed (p < 0.05). NPM1 and FLT3-ITD had higher mutation frequencies in patients with normal chromosome karyotypes than abnormal chromosome karyotypes (p<0.001, p=0.005). DNMT3A and NRAS mutations predicted lower CR rates. DNMT3A, TP53, and U2AF1 mutations were related to unfavorable OS. TET2 mutation with CD123+, CD11b+ or CD34- predicted lower CR rate. IDH2+/CD34- predicted lower CR rate. ASXL1+/CD38+ and SRSF2+/CD123- predicted shorter OS. ASXL1+/CD38+ and SRSF2+/CD123- predicted shorter OS.

Conclusion: The study showed specific correlations between elderly AML patients’ genetic mutations and clinical features, some of which may impact prognosis.

Keywords: elderly AML, genetic mutation, FAB subtype, karyotype, immunophenotype, prognosis

Introduction

Acute myeloid leukemia is a malignant tumor of myeloid cells derived from bone marrow. Recent surveys show that the initiation and progression of AML are often related to genetic mutations that control multiple functions. For example, FLT3 and KIT promote the proliferation and survival of malignant cells. CEBPA, RUNX1, and NPM1 impair the differentiation and apoptosis of hematopoietic cells. DNMT3A, TET2, IDH1, and IDH2 regulate the epigenetic mechanism. NPM1, CEBPA, and GATA2 mutations predict favorable prognosis, FLT3-ITD, RUNX1, ASXL1, MLL, TP53, PHF6, and U2AF1 mutations predict adverse prognosis, especially FLT3-ITD, ASXL1, RUNX1, and TP53 mutations predict adverse prognosis under certain conditions.1–14

The Morphologic-Immunologic-Cytogenetic classification (MIC classification) is used to make a definitive diagnosis of AML. Many studies have focused on the correlation between MIC classification and prognosis. For example, karyotype analysis based on cytogenetics as an important reference marker for prognosis has been written into National Comprehensive Cancer Network (NCCN) and European Leukemia Net (ELN) guidelines.15,16 Among the common immunophenotypes of AML, CD19, CD38, and MPO are the immunophenotypes reported in the literature to be of favorable prognosis.17–20 The immunophenotypes that predict adverse prognosis include CD7, CD11b, CD34, CD56, and CD123.21–28 CD13, CD33, and CD117, which are particularly...
important for the diagnostic classification of AML patients, and their relations with prognosis are currently inconclusive.\textsuperscript{21} CD14 and CD64 are mainly expressed in M4 and M5 subtypes,\textsuperscript{29,30} and their associations with prognosis have been less reported. However, the MIC classification alone cannot effectively and accurately predict the prognosis. It also needs to be analyzed in conjunction with the genetic mutations of the patients.

AML is a senile disease (median age at diagnosis is 65–70 years).\textsuperscript{31} Unlike the precise current stratification of young and middle-aged AML patients, the clinical treatment of elderly patients is still ineffective, and their prognosis is adverse.\textsuperscript{32} In this paper, we did correlation studies with genetic mutations, clinical features, and prognosis comprehensively evaluated the clinical and biological characteristics of elderly AML patients, and provided some reference to this particular group’s prognosis and clinical treatment.

\textbf{Methods}

\textbf{Patients}

In this study, we enrolled 90 elderly (aged $\geq$ 60 years) AML (non-APL) patients who were newly diagnosed in the First Affiliated Hospital of University of Science and Technology of China (Anhui Provincial Hospital Affiliated to Anhui Medical University) and targeted NGS was performed on bone marrow (BM) from all enrolled patients between April 2015 and March 2021. According to the 2016 WHO classification of myeloid neoplasms and acute leukemia,\textsuperscript{33} these patients were diagnosed and classified. The clinical features of the patients, including age, gender, BM blast percentage, and so on. Details are shown in Table 1.

\textbf{NGS Testing}

NGS testing was submitted to Shanghai Siteide Biotechnology Co., Ltd. 34 common AML genetic mutations were tested, including ASXL1, BCOR, BCORL1, CALR, CBL, CEBPA, CSF3R, DNMT3A, ETV6, EZH2, FLT3, GATA2, IDH1, IDH2, JAK2, KIT, KRAS, KMT2A, MPL, NPM1, NRAS, PDGFRA, PHF6, PIGA, RUNX1, SETBP1, SF3B1, SH2B3, SRSF2, TET2, TP53, U2AF1, WT1, and ZRSR2.

\textbf{Definitions and Statistical Analysis}

Overall survival (OS) was defined as the length of time between the date of diagnosis and the date of last follow-up or death. Relapse-free survival (RFS) was defined as the period from first complete remission to relapse, death, or the last follow-up time. According to the NCCN guidelines,\textsuperscript{15} the efficacy after the initial induction chemotherapy was defined classified as complete remission (CR), and non-remission (NR).

The chi-squared test or Fisher’s exact test was used to infer the correlation between categorical variables, and Mann Whitney U-test was used for continuous variables. Kaplan Meier analysis (Log rank test) and the COX proportional hazard regression analysis were used for survival analysis. SPSS 26.0 software was used for all statistical analysis, and differences with $p<0.05$ were considered significant.

\textbf{Results}

\textbf{The Spectrum of Genetic Mutations in Elderly Patients with AML}

As shown in Figure 1, a total of 29 genetic mutations were identified in 82 patients among 90 cases with a frequency of 91.1%. There were 17 genetic mutations in more than 5% of patients, the most commonly genetic mutation was TET2 (26.7%, 24/90), followed by DNMT3A (25.6%, 23/90), IDH2 (24.4%, 22/90), NPM1 (24.4%, 22/90), FLT3-ITD (18.9%, 17/90), CEBPA (15.6%, 14/90), ASXL1 (14.4%, 13/90), IDH1 (14.4%, 13/90), SRSF2 (14.4%, 13/90), BCOR (12.2%, 11/90), TP53 (11.1%, 10/90), NRAS (8.9%, 8/90), RUNX1 (8.9%, 8/90), U2AF1 (7.8%, 7/90), BCORL1 (7.8%, 7/90), WT1 (7.8%, 7/90), and FLT3-TKD (7.8%, 7/90).

\textbf{Association of Genetic Mutations with Clinical Features of Elderly AML Patients}

There were 17 genetic mutations in more than 5% of patients. We studied the relationships of these genetic mutations with clinical features of elderly AML patients, including MIC classification, cytogenetic risk, the origin of disease, and response to
chemotherapy (Supplementary Tables S1 and S2). Overview of gene mutations, some clinical features (FAB classification, chromosome karyotypes, cytogenetic risk, response to chemotherapy) and overall survival in elderly AML are shown in Figure 2.

**Table 1 Clinical Characteristics**

| Variant                          | Value                         |
|----------------------------------|-------------------------------|
| Median age (years old)           | 68 (60–81)                    |
| Gender ratio (male/female)       | 51:39                         |
| Median BM blasts (%)             | 48.25 (20–97)                 |
| Median WBC count (×10^9/L)       | 5.78 (0.4–214)                |
| Median platelet (×10^9/L)        | 50.5 (7–324)                  |
| Median hemoglobin (g/L)          | 75.5 (34–136)                 |
| Origin of disease                |                               |
| de novo AML                      | 82 (91.1%)                    |
| s-AML                            | 7 (7.8%)                      |
| t-AML                            | 1 (1.1%)                      |
| ELN risk assessment              |                               |
| Favorable-risk                   | 40 (44.4%)                    |
| Intermediate-risk                | 23 (25.6%)                    |
| High-risk                        | 27 (30%)                      |
| FAB-classification                |                               |
| M1                               | 2 (2.2%)                      |
| M2                               | 35 (38.9%)                    |
| M4                               | 7 (7.8%)                      |
| M5                               | 26 (28.9%)                    |
| Unclassified                     | 20 (22.2%)                    |
| Response to chemotherapy         |                               |
| CR                               | 47 (52.2%)                    |
| NR                               | 43 (47.8%)                    |
| Induction chemotherapy           |                               |
| Standard-intensive chemotherapy  | 41 (45.6%)                    |
| Low-intensive chemotherapy       | 39 (6.7%)                     |
| Chemotherapy plus transplantation| 10 (11.1%)                    |
| Median follow-up term (months)   | 8.32 (0.13–78.23)             |

**Abbreviations:** BM, bone marrow; WBC, white blood cell; s-AML, secondary AML; t-AML, therapy-related AML; ELN, European Leukemia Net; FAB, French-American-British; CR, complete remission; NR, no response.

Association of Genetic Mutations with MIC Classification of Elderly AML Patients

In this study, in terms of analyzing the association between gene mutation and FAB classification, we excluded 20 cases that failed to get a specific FAB classification and conducted the study on the remaining 70 elderly AML patients. We
Figure 1: Overview of gene mutations, clinical features, and overall survival in elderly AML. Each column represents an individual patient, and the presence of the mutation is indicated in red. The upper panel shows the gene mutation, and the lower panel shows the clinical features and overall survival.

Abbreviations: FAB, French-American-British; CR, complete remission; NR, no response; OS, overall survival.

Figure 2: Mutation frequency of each gene in 90 elderly AML patients.

https://doi.org/10.2147/CIA.S375000

Dove Press

Clinical Interventions in Aging 2022:17

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studied the relationships of gene mutations with FAB classification of elderly AML patients and found that DNMT3A, BCOR, U2AF1, and BCORL1 mutations were unevenly distributed among different subtypes \((p<0.05)\). DNMT3A had the highest mutation frequency in M5 subtype (50%, 13/26), followed by M4 subtype (40%, 2/5), and was rare in M1 and M2 subtypes \((p=0.007)\). BCOR mutation mainly appeared in M5 (23.1%, 6/26), M2 (2.9%, 1/35), and was not found in M1 and M4 subtypes \((p=0.048)\). U2AF1 and BCORL1 had the highest mutation frequency in M5 subtype (15.4%, 4/26; 19.2%, 5/26), and were rare in other subtypes \((p=0.039, 0.014)\).

We then studied the relationship between genetic mutations and immunophenotypes and found that many genetic mutations were distributed differently when an immunophenotype was expressed or not expressed \((p<0.05)\) \((\text{Supplementary Table S2})\). DNMT3A mutation was distributed differently in patients when CD123 was expressed (CD123+) or not expressed (CD123-), this difference was considered significant \((p=0.036)\). The mutation-positive rate of IDH2 was different between CD7+ and CD7- patients \((p=0.05)\), CD34+ and CD34- patients \((p=0.033)\). The distributions of NPM1 mutation were different in patients with CD7+/CD7-, CD11b+/CD11b-, CD34+/CD34-, CD38+/CD38-, CD64+/CD64-, CD117+/CD117- \((p=0.024, 0.002, <0.001, 0.012, 0.045, 0.03)\). The patients with CD34+/CD34+, CD64+/CD64+, CD123+/CD123- had different distributions of FLT3-ITD mutation \((p=0.016, 0.035, 0.009)\). ASXL1 mutation had different distributions in patients with CD33+/CD33-, CD56+/CD56- \((p=0.035, 0.028)\). The distributions of IDH1 mutation varied according to different patients who were CD33+/CD33-, CD123+/CD123- \((p=0.035, 0.006)\). The distributions of SRIF2 mutation in of patients with CD33+ differed from the patients with CD33- \((p=0.035)\). Among the patients with CD34+ or CD34-, BCOR mutation distributed differently \((p=0.031)\). The distributions of NRAS mutation were different in patients with CD11b+/CD11b-, CD14+/CD14-, CD34+/CD34-, CD117+/CD117- \((p=0.001, 0.014, 0.035, 0.008)\). RUNX1 mutation was distributed differently in patients with CD33+ and CD33- \((p=0.042)\). The patients with MPO+ had different distributions of U2AF1 mutation from the patients with MPO- \((p=0.004)\). The distributions of WT1 mutation varied from patients with CD56+ to CD56- \((p=0.048)\) and patients with CD123+/CD123- was the same way \((p=0.039)\).

We studied the relationships of genetic mutations with chromosome karyotypes of elderly AML patients and found that NPM1 and FLT3-ITD mutations were unevenly distributed among different chromosome karyotypes \((p<0.001, p=0.005)\). NPM1 had a higher mutation frequency in patients with normal chromosome karyotypes (36.2%, 21/58) than abnormal chromosome karyotypes (3.1%, 1/32). The distributions of FLT3-ITD mutation were different in patients with normal chromosome karyotypes (27.6%, 16/58) and abnormal chromosome karyotypes (3.1%, 1/32).

**Association of Genetic Mutations with Other Clinical Features of Elderly AML Patients**

According to the 2017 ELN risk stratification, 90 elderly AML patients were classified into three risk groups: favorable-risk, intermediate-risk, and high-risk. \(16\) We studied the relationships of genetic mutations with cytogenetic risk. We found that NPM1, CEBPA, ASXL1, TP53, NRAS, RUNX1, and U2AF1 mutations were unevenly distributed among different risk groups \((p<0.001, p=0.002, p<0.001, p<0.001, p=0.037, p<0.001, p=0.008)\). NPM1, CEBPA, and NRAS mutations mainly appeared in favorable-risk group (50%, 20/40; 30%, 12/40; 15%, 6/40). ASXL1, TP53, RUNX1, and U2AF1 mutations mainly appeared in high-risk group (40.7%, 11/27; 37%, 10/27; 29.6%, 8/27; 18.5%, 5/27).

Acute myeloid leukemia was divided into three types according to the origins of the disease. De novo AML was defined as a disease without previous chemotherapy treatment or hematological disorders. Secondary AML (s-AML), AML with previous hematological disorders. Therapy-related AML (t-AML), AML related to the history of chemotherapy or toxic chemicals. We studied the relationship between genetic mutations and the origins of disease and found no significant difference in genetic mutations among each root.

After one or more cycles of induction chemotherapy, 47 patients achieved complete remission and 43 patients did not. Regarding the response to chemotherapy, 90 elderly AML patients were classified into two groups: CR and NR. We studied the relationships of genetic mutations with response to chemotherapy of elderly AML patients and found that DNMT3A, NPM1, and NRAS mutations were unevenly distributed among different groups \((p=0.015, 0.028, 0.025)\). They all had higher mutation frequency in NR-group (37.2%, 16/43; 34.9%, 15/43; 16.3%, 7/43) than CR-group (14.9%, 7/47; 14.9%, 7/47; 2.1%, 1/47).
Genetic Mutation Predicted Rate of Complete Remission and Survival

We evaluated the impact of single genetic mutation on the rate of complete remission in elderly AML patients. In the total cohort of 90 elderly AML patients, the rate of CR was significantly lower in patients with DNMT3A (30.4% vs 59.7%, \( p = 0.015 \)) or NRAS mutation (12.5% vs 56.1%, \( p = 0.025 \)) than in those with wild type. And in the 63 patients in the low and intermediate-risk group, patients with DNMT3A (31.3% vs 63.8%, \( p = 0.023 \)) or NRAS mutation (12.5% vs 61.8%, \( p = 0.018 \)) also had a significantly lower rate of CR than those without the above genetic mutations.

Our study evaluated the impact of a single genetic mutation on the survival of elderly AML patients. We found DNMT3A mutation predicted significantly unfavorable OS in the total cohort of 90 elderly AML patients (1-year OS: 21.7% vs 47.8%, \( p = 0.029 \)) and the 63 patients in the low and intermediate-risk group (1-year OS: 25% vs 55.3%, \( p = 0.036 \)). In the total cohort of 90 elderly AML patients, TP53 mutation had a trend of inferior OS (1-year OS: 10% vs 45%, \( p = 0.043 \)), and mutant U2AF1 was also related to an unfavorable OS (1-year OS: 0% vs 44.6%, \( p = 0.039 \)).

Univariate survival analysis showed that DNMT3A, TP53, NRAS, and U2AF1 mutations predicted shorter OS (\( p = 0.013 \), Figure 3A; \( p = 0.021 \), Figure 3B; \( p = 0.002 \), Figure 3C; \( p < 0.001 \), Figure 3D); NPM1 predicted longer RFS (\( p = 0.043 \), Figure 4A); RUNX1, U2AF1, WT1, and FLT3-TKD mutations predicted shorter RFS (\( p = 0.02 \), Figure 4B; \( p < 0.001 \), Figure 4C; \( p = 0.015 \), Figure 4D; \( p = 0.012 \), Figure 4E).

Figure 3 Kaplan-Meier curves for survival of patients with or without specific genetic mutations in elderly AML. The red and blue lines represent the survival of patients without or with mutations, respectively. (A) OS for patients with or without DNMT3A mutation (\( p = 0.013 \)). (B) OS for patients with or without TP53 mutation (\( p = 0.021 \)). (C) OS for patients with or without NRAS mutation (\( p = 0.002 \)). (D) OS for patients with or without U2AF1 mutation (\( p = 0.001 \)).

Abbreviations: OS, overall survival; wt, wild type; mut, mutant.
**Figure 4** Kaplan-Meier curves for survival of patients with or without specific genetic mutations in elderly AML. The red and blue lines represent the survival of patients without or with mutations, respectively. (A) RFS for patients with or without NPM1 mutation ($p=0.043$). (B) RFS for patients with or without RUNX1 mutation ($p=0.02$). (C) RFS for patients with or without U2AF1 mutation ($p<0.001$). (D) RFS for patients with or without WT1 mutation ($p=0.015$). (E) RFS for patients with or without FLT3-TKD mutation ($p=0.012$).

**Abbreviations:** RFS, relapse-free survival; wt, wild type; mut, mutant.
Association of Genetic Mutation with MIC Classification Predicted Rate of Complete Remission and Survival

There were specific correlations between elderly AML patients’ genetic mutations and MIC classification, including FAB subtype, immunophenotype, and chromosome karyotype. We evaluated the impacts of these correlations on complete remission rate and survival in elderly AML patients. The complete remission rate was distributed differently among different subtypes (M5: 12.5%, M2: 66.7%, M4: 0; \( p = 0.043 \)) in patients with TET2 mutation. The CR rate was significantly different between patients with CD123+/CD123- (20% vs 64.3%, \( p = 0.047 \)), CD11b+/CD11b- (0 vs 57.9%, \( p = 0.041 \)), and CD34+/CD34- (66.7% vs 11.1%, \( p = 0.013 \)) when TET2 was mutant. The complete remission rate was significantly different between patients with CD34+/CD34- (75% vs 20%, \( p = 0.010 \)) when IDH2 was mutant. In patients with ASXL1 mutation, CD38+ predicted shorter OS than CD38- (1-year RFS: 0 vs 66.7%, \( p = 0.038 \)). When SRSF2 was mutant in patients, CD123+ predicted longer OS than CD123- (1-year OS: 80% vs 12.5%, \( p = 0.032 \)). But the association of genetic mutations with chromosome karyotypes did not impact the CR rate and survival of elderly AML patients in this study.

Multivariate Analysis of Survival of Elderly AML Patients

Variables including BM blast percentage, WBC count, platelet count, the origin of disease, response to chemotherapy, induction chemotherapy, and genetic mutations were included in the multivariate analysis. For OS, platelet count less than \( 63 \times 10^9/L \), failure to achieve CR after 3 or more cycles of induction chemotherapy (NR), DNMT3A mutation, and TP53 mutation were independent unfavorable factors. For RFS, platelet count less than \( 63 \times 10^9/L \), RUNX1 mutation, U2AF1 mutation, and FLT3-TKD mutation were independent unfavorable factors (Table 2).

**Table 2 Multivariate Analysis for Survival of New Diagnosed Elderly AML Patients**

| Factors                              | HR (95% CI)   | p value |
|--------------------------------------|---------------|---------|
| **OS**                               |               |         |
| PLT count (\( \leq 63 \times 10^9/L \) vs \( > 63 \times 10^9/L \)) | 2.104 (1.175–3.769) | 0.012 |
| Response to chemotherapy (NR vs CR)  | 10.358 (5.158–20.801) | <0.001 |
| DNMT3A (mutant vs wild)              | 2.219 (1.203–4.039) | 0.011 |
| TP53 (mutant vs wild)                | 2.405 (1.033–5.599) | 0.042 |
| **RFS**                              |               |         |
| PLT count (\( \leq 63 \times 10^9/L \) vs \( > 63 \times 10^9/L \)) | 3.511 (1.190–10.359) | 0.023 |
| RUNX1 (mutant vs wild)               | 8.548 (2.060–35.467) | 0.003 |
| U2AF1 (mutant vs wild)               | 12.622 (1.730–92.100) | 0.012 |
| FLT3-TKD (mutant vs wild)            | 9.891 (1.903–51.424) | 0.006 |

**Abbreviations:** OS, overall survival; CR, complete remission; NR, no response; RFS, relapse-free survival.

Discussion

Elderly AML patients have significant heterogeneity, often accompanied by adverse karyotype and various comorbidities, and have a low response rate to standard chemotherapy. Especially patients older than 75 years have shorter OS.\(^3^4\) Therefore, it is still essential to comprehensively evaluate the clinical and biological characteristics of elderly AML patients and analyze the correlations among them to make more accurate diagnostic classification and risk stratification. In this paper, we correlated genetic mutations with their MIC classification and other clinical features in elderly AML patients, and further analysis was done in combination with prognosis.
Previous studies have shown that FLT3-ITD, RUNX1, ASXL1, MLL, TP53, PHF6, and U2AF1 mutations in adult AML patients predict adverse prognosis. However, the impact of these genes on prognosis still needs further exploration in elderly patients. In addition to TP53 and U2AF1 mutations, we found that DNMT3A mutation significantly predicted inferior OS. RUNX1, U2AF1, and FLT3-TKD mutations were independent unfavorable factors for RFS. And TET2 mutation is an unfavorable prognostic factor for RFS in AML, but there are few studies on its association with CR rate. Our study found that TET2 mutation with CD123+ or CD11b+ in elderly AML patients predicted a low CR rate. We speculate that the adverse effects of CD123 and CD11b on prognosis may be related to their resulting lower CR rate.

DNMT3A mutation significantly predicted a lower CR rate and inferior OS. We also found that DNMT3A was common in M5 and M4 subtypes, and had a higher mutation frequency in patients with CD123+ than CD123-. Considering the adverse impacts of DNMT3A mutation and CD123+ on elderly AML patients, we need to pay more attention to patients with DNMT3A mutation and CD123+, and further explore follow-up treatment. And we found that IDH2 mutation had a higher mutation frequency in patients with CD34- than CD34+. The CR rate was significantly lower in patients with CD34- than CD34+, especially when IDH2 or TET2 was mutant in our study, this differed from previous studies in which CD34 predicted a low CR. CD34 may interact with IDH2 or TET2 mutations, reducing each other’s negative effects on prognosis.

NPM1 and CEBPA mutations were widely reported as favorable prognostic factors of survival in AML. Our study found they mainly appeared in the favorable-risk group, NPM1 mutation had a higher mutation frequency in patients with normal chromosome karyotypes than abnormal, and NPM1 mutation was associated with CD7-, CD11b+, CD34+, CD38+, CD64+, and CD117-. However, our study did not find a predictive impact of these associations on CR rate and overall survival. Previous studies have shown that FLT3-ITD mutation predicts an adverse prognosis. In this study, it was common in CD34-, CD64+, and CD123+ patients. The prognostic significance of CD64 has been less reported. CD64 may be associated with FLT3-ITD mutation, which predicts an adverse prognosis. FLT3-ITD mutation also had a higher mutation frequency in patients with normal chromosome karyotypes than abnormal. ASXL1 mutation mainly appeared in the high-risk group, was common in CD33- and CD56- patients in this study. CD38 was reported to be related to a favorable prognosis. While in our study, patients with ASXL1 mutation and CD38+ had inferior OS. We found IDH1 mutation was common in CD33- and CD123- patients, while SRSF2 mutation was also more associated with CD33- than CD33+, which had no association with prognosis in AML in our study. CD123 predicts an adverse prognosis. Yang et al suggested that SRSF2 mutation was not an independent prognostic factor in AML patients and not be associated with unfavorable prognosis in Chinese AML patients. In our study, when SRSF2 was mutant in patients, CD123+ might predict longer OS than CD123-. These results suggested that when SRSF2 was mutant in elderly AML patients, CD123 might not be associated with an adverse prognosis. Terada et al suggested that BCOR or BCORL1 gene mutation predicted inferior OS and RFS. In our study, BCOR mutation appeared in patients with CD34+ and not CD34-, BCOR and BCORL1 mutations both had the highest mutation frequency in the M5 subtype. Further explorations of these associations with prognosis will be completed in our subsequent research.

Consistent with previous studies, TP53 mutation mainly appeared in a high-risk group and also had a trend of inferior OS in our study. In terms of response to chemotherapy, NRAS had high mutation frequency in NR-group and predicted lower CR rate, but mainly appeared in the favorable-risk group. It was common in CD11b+, CD14+, CD34-, and CD117- patients. Further study is needed to determine the impact of NRAS mutation on the prognosis. RUNX1 mutation mainly appeared in a high-risk group, might be an independent unfavorable factor for RFS, and was more common in patients with CD33- than CD33+. Considering the adverse impacts of RUNX1 mutation, we should pay more attention to patients with CD33-. In our study, consistent with previous studies, U2AF1 mutation had also the same trend of inferior OS, which was more common in patients with MPO- than MPO+. And we found it had the highest mutation frequency in the M5 subtype, so we should pay more attention to these associations. Zheng et al suggested that WT1 mutation predicted lower CR rate, inferior OS, and RFS. In our study, WT1 mutation was associated with CD56- and CD123+. In addition to the adverse impact on the prognosis of FLT3-ITD mutation, we found that FLT3-TKD mutation was also an independent prognostic risk factor.

In this study, multiple clinical features such as FAB classification, immunophenotype, and chromosome karyotype were associated with their gene mutations from multiple angles and dimensions. Then these associations were also analyzed with prognosis, and some results worthy of reference, such as our study first found that TET2 mutation with CD123+ or CD11b+ in elderly AML patients predicted a low CR rate, and CD64 may be associated with FLT3-ITD mutation, which predicts an
adverse prognosis. We also found that when SRSF2 was mutant in patients, CD123+ might predict longer OS than CD123-.
Therefore, we should pay attention to these three groups of elderly AML. We hope our work will contribute to prognostic prediction systems and improve the outcomes of elderly AML patients through targeted treatment.

**Conclusion**
The study showed specific correlations between elderly AML patients’ genetic mutations and clinical features, some of which impact prognosis.

**Data Sharing Statement**
The datasets used and analyzed in this study are available from the corresponding authors upon reasonable request.

**Ethics Approval and Informed Consent**
This study was approved, and the written informed consent was waived by the ethics committee of the first affiliated hospital of University of Science and Technology of China (2022-RE-064) due to the retrospective nature of the review, and confirmed that the data was anonymized and maintained with confidentiality. The study was conducted in accordance with the Declaration of Helsinki.

**Acknowledgments**
The authors thank the patients, their families, and all doctors who treated the patients.

**Funding**
This work was supported by the Key Research and Development Program Project of Anhui Province [No. 201904a07020102], the National Natural Science Foundation of China [No. 81600107] and the Anhui Provincial Natural Science Foundation [No. 1708085MH180].

**Disclosure**
The authors report no conflicts of interest in this work.

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