

# Leukemic stem cell phenotype is associated with mutational profile in acute myeloid leukemia

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**Background/Aims:** Understanding leukemic stem cell (LSC) is important for acute myeloid leukemia (AML) treatment. However, association of LSC with patient prognosis and genetic information in AML patients is unclear.

**Methods:** Here we investigated the associations between genetic information and the various LSC phenotypes, namely multipotent progenitor (MPP)-like, lymphoid primed multipotent progenitor (LMPP)-like and granulocyte-macrophage progenitors (GMP)-like LSC in 52 AML patients.

**Results:** In secondary AML patients, MPP-like LSC was significantly higher than *de novo* AML ( $p = 0.0037$ ). The proportion of MPP-like LSC was especially high in post-myeloproliferative neoplasm AML ( $p = 0.0485$ ). There was no correlation between age and LSC phenotype. Mutations of *KRAS* and *NRAS* were observed in MPP-like LSC dominant patients, *TP53* and *ASXL1* mutations in LMPP-like LSC dominant patients, and *CEBPA*, *DNMT3A* and *IDH1* mutations in GMP-like LSC dominant patients. Furthermore, *KRAS* mutation was significantly associated with MPP-like LSC expression ( $p = 0.0540$ ), and *TP53* mutation with LMPP-like LSC expression ( $p = 0.0276$ ). When the patients were separated according to the combined risk including next generation sequencing data, the poorer the prognosis, the higher the LMPP-like LSC expression ( $p = 0.0052$ ). This suggests that the dominant phenotype of LSC is one of the important factors in predicting the prognosis and treatment of AML.

**Conclusions:** LSC phenotype in AML is closely associated with the recurrent mutations which has prognostic implication. Further research to confirm the meaning of LSC phenotype in the context of genetic aberration is warranted.

**Keywords:** Acute myeloid leukemia; Leukemic stem cell; Next generation sequencing; Prognosis

## INTRODUCTION

Acute myeloid leukemia (AML) is a clinically heterogeneous disease characterized by multitudes of chromosomal abnormalities and gene mutations. With advances in next generation sequencing (NGS) techniques,

more sophisticated risk stratification models have been adapted for better management of the disease [1-3]. Unfortunately resistance to treatment and relapse remains a big challenge, and some feel that the mutation status alone is not enough to predict the prognosis of AML.

Leukemic stem cells (LSCs) are cells with self-renewal

capacity to repopulate a leukemia and often thought as the major cause of resistance and relapse despite their low frequency [4-6]. The origin of these cells, whether it be from normal hematopoietic stem cells or from more mature progenitors that gain stem-ness features, may differ among patients [7-9], but previous reports on the prognostic implications of LSCs support the importance of eliminating LSCs to improve AML clinical outcomes [4]. As such, understanding the relationship between LSC and genetically defined sub-clones can, in turn, help to understand the heterogeneity of AML [10]. However, to date, there are only a few reports specifically focusing on this topic. To this end, we conducted this study to (1) examine the phenotypic diversity of AML-LSC, (2) explore the association between AML-LSC phenotypes and gene mutations, and (3) investigate the prognostic implications of AML-LSCs.

## METHODS

### Study design and subjects

Newly diagnosed adult AML patients, defined as 18 years or older, were included for this study. Patients with acute promyelocytic leukemia were excluded. The patient selection and bone marrow sample acquisition was done between October 2016 and July 2018. At the end, a total of 52 patients were included for analyses. Their medical records were reviewed and analyzed for demographics, baseline disease characteristics, factors related to treatment, response to treatment, adverse events including treatment related mortality, and survival outcomes. Data available up to March 2019 were used. Their bone marrow samples collected at AML diagnosis were subjected to flow cytometry and NGS.

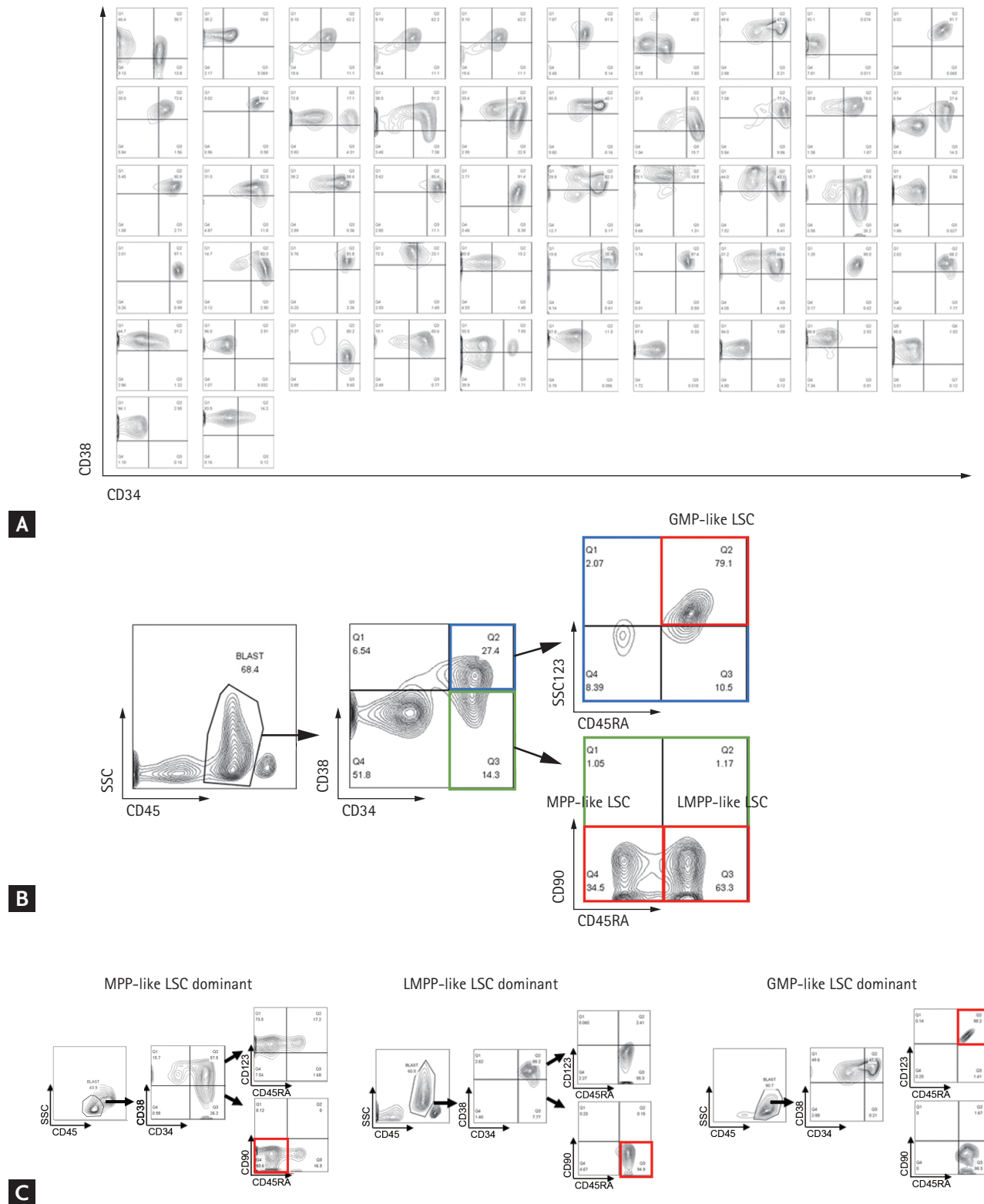
The diagnosis of AML was made according to the WHO classification of hematopoietic neoplasms [11], which requires identification of 20% or more leukemic blasts in the bone marrow. Secondary AML was defined as AML following myelodysplastic syndrome (MDS) or myeloproliferative neoplasm (MPN) confirmed prior to the diagnosis of AML. Cytogenetic studies were performed onsite, whose satisfactory performance was monitored by a national external quality assurance scheme. Bone marrow cells were cultured for 24 hours then karyotype was analyzed using the standard G-banding technique.

The karyotypes were constructed and chromosomal abnormalities were reported in accordance with the International System for Human Cytogenetic Nomenclature [12]. Prognostic grouping of cytogenetics was performed according to Southwest Oncology Group (SWOG) criteria [13]. Risk stratification was mainly based on cytogenetics, and NGS results for those with available data, according to 2019 National Comprehensive Cancer Network (NCCN) guidelines [14].

The study was conducted in compliance with all national and international ethical standards for research with humans and for research using radiopharmaceuticals. This study was conducted according to the Declaration of Helsinki and was approved by the Institutional Review Board of Seoul National University Hospital (IRB No. H-1902-140-101). Written informed consent was obtained from all patients. All authors had access to the study data and reviewed and approved this study.

### Flow cytometry

Mononuclear cells were isolated from the patient's bone marrow aspirates by ficoll gradient centrifugation and cryopreserved in serum-free medium. Stored cells were thawed to Iscove's Modified Dulbecco's Medium and washed with fluorescence-activated cell sorting (FACS) buffer (2% fetal bovine serum, Dulbecco's phosphate-buffered saline). Cells were stained with following anti-human monoclonal antibodies: CD45-APC/cy7 (557833), CD34-APC (555824), CD38-BV421 (562445), CD90-PE (555596), CD123-PE/Cy7 (560826), and CD45RA-PerCP/Cy5.5 (563429) (BD Bioscience). Analyses were performed on a FACSCanto II (High Throughput Sampler, BD Bioscience, Franklin Lakes, NJ, USA) and FlowJo version 10.0 program (BD Bioscience). First, the expression profiles of CD34 and CD38 markers were analyzed (Fig. 1A). Then, the phenotypes of LSCs were further analyzed per methods proposed by Goardon et al. [15]; multipotent progenitor (MPP)-like LSC (CD34<sup>+</sup>CD38<sup>-</sup>CD90<sup>-</sup>CD45RA<sup>-</sup>), lymphoid primed multipotent progenitor (LMPP)-like LSC (CD34<sup>+</sup>CD38<sup>+</sup>CD90<sup>-</sup>CD45RA<sup>+</sup>) and granulocyte-macrophage progenitors (GMP)-like LSC (CD34<sup>+</sup>CD38<sup>+</sup>CD123<sup>+</sup>CD45RA<sup>+</sup>). The CD45 and side scatter properties were used for gating bulk leukemic blast cells and sub-gated from the cells [16], as shown in Fig. 1B. End results showed that there was a dominantly expressed LSC phenotype in each patient (Fig. 1C). In some



**Figure 1.** Phenotypic diversity of acute myeloid leukemia (AML) leukemic stem cells (LSCs). (A) CD34 and CD38 surface marker profiles of all 52 patients. (B) Gating strategy for AML-LSCs. (C) Three distinct populations were analyzed as LSC phenotypes: left, multipotent progenitor (MPP)-like LSC; middle, lymphoid primed multipotent progenitor (LMPP)-like LSC; right, granulocyte-macrophage progenitors (GMP)-like LSC.

cases, two or more LSC phenotypes co-existed. The percentage of cell fractions satisfying each LSC phenotype markers in CD34<sup>+</sup> cells was calculated.

### Targeted capture and massive parallel sequencing

The DNA capture probes for 76 target genes were designed using the Agilent SureDesign web-based application (<https://earray.chem.agilent.com/suredesign/>, Agilent Technologies, Santa Clara, CA, USA). The target regions included protein coding exons with 10 bp intron flanking regions and hot spot regions of the 20 genes involved in recurrent translocations. DNA was extracted on a Chemagic 360 instrument (Perkin Elmer, Baesweiler, Germany). The genomic DNA was sheared using Covaris S220 focused-ultrasonicator (Covaris, Woburn, MA, USA). We used 50 ng of total input genomic DNA. A library preparation was performed according to Agilent's SureSelectQXT Target Enrichment protocol (Agilent Technologies). Paired-end 150-bp sequencing was using NextSeq 550 Dx platform (Illumina, San Diego, CA, USA). Targeted sequencing raw data was obtained in FASTQ format.

### Secondary analysis using in-house bioinformatics pipeline

The produced sequencing data was analyzed using the Seoul National University Hospital First Hemic Panel Analysis Pipeline (FHPAP). First, we performed the quality control of the FASTQ file and analyzed only the data that passed the criteria. Pair-end alignment to HG19 reference genome was performed using BWA-men and the GATK Best Practice [17,18]. After finishing the alignment step, an "analysis-ready BAM" was produced and a second quality control is performed to determine if further variant calling is appropriate. In the pipeline, SNV, InDel, CNV, translocation, *FLT3* ITD, and *KMT2A* PTD were analyzed using at least more than two analysis tools including in-house and open-source software. Along with in-house software, GATK UnifiedGenotyper, SNVer, LoFreq were used for SNV/InDel detection [18-20], Delly and Manta for Translocation discovery [21,22], and THetA2 and CNVKit for purity estimation and CNV calling [23,24], respectively. *FLT3* ITD and *KMT2A* PTD were analyzed by only the in-house software. Detected variants was annotated by SnpEff with various database such as RefSeq, COSMIC, dbSNP, ClinVar, and gnomAD. Then germline variant was filtered using the pop-

ulation frequency of these databases (> 1% population frequency) [25-30]. Finally, the variants were confirmed throughout a comprehensive review of a multidisciplinary molecular tumor board.

### Statistical analysis

The relapse free survival (RFS) and overall survival (OS) curves were estimated using the Kaplan–Meier method. The RFS was derived from the date AML diagnosis to that of relapse or death from any cause, while the OS was defined as the time from AML diagnosis to death from any cause. If patients survived, RFS and OS was censored on the last date of follow-up when no death or relapse was confirmed. Complete remission rate, based on the revised recommendations of the Center for International Blood and Marrow Transplant Research (CIBMTR) group, with induction was compared. Induction TRM was also analyzed, and defined as mortality during induction chemotherapy. The differences between groups were assessed using a Mann–Whitney test or one-way analysis of variance for continuous variables, and Pearson chi-square test for categorical variables, as indicated. All data were analyzed using the SAS Enterprise Guide version 6.1 (SAS Institute, Cary, NC, USA) and the statistical software R ([www.r-project.org](http://www.r-project.org)). A *p* value of < 0.05 was considered statistically significant.

## RESULTS

### Patient characteristics, AML subtypes and LSC phenotypes

The baseline characteristics of the enrolled patients are presented in Table 1. There were 40 *de novo* AML and 12 secondary AML patients. Among secondary AML patients, there were four post-MPN AML patients and eight post-MDS AML patients. The median age of the whole group was 62 years.

When different LSC phenotypes were considered, secondary AML group showed higher fraction of CD34<sup>+</sup>CD38<sup>-</sup> cells (*p* = 0.0245) and MPP-like LSCs (*p* = 0.0037) compared to *de novo* AML group (Fig. 2A). When secondary AML was further classified into post-MPN versus post-MDS, there was a statistically significant increase in percentage of MPP-like LSCs in post-MPN AML compared to post-MDS AML (*p* = 0.0485) (Fig. 2B).



**Table 1. Baseline characteristics**

Characteristic	Total	<i>De novo</i> AML	Secondary AML	<i>p</i> value
Number of patients	52	40	12	NA
Age, yr	62 (23–88)	63 (23–88)	60 (40–81)	0.656
≤ 65	33 (63.5)	26 (65.0)	7 (58.3)	
> 65	19 (36.5)	14 (35.0)	5 (41.7)	
Sex, male	33 (63.5)	26 (65.0)	7 (58.3)	0.674
<b>Cytogenetic risk</b>				
Low	6 (11.5)	6 (15.0)	0 (0.0)	0.001
Intermediate	38 (73.1)	32 (80.0)	6 (50.0)	
High	8 (15.4)	2 (5.0)	6 (50.0)	
<b>Combined risk<sup>a</sup></b>				
Low	7 (13.5)	7 (17.5)	0	0.037
Intermediate	19 (36.5)	15 (37.5)	4 (33.3)	
High	12 (23.1)	6 (15.0)	6 (50.0)	
Unknown	14 (26.9)	12 (30.0)	2 (16.7)	
Extramedullary involve	0	0	0	NA
<b>Laboratory findings</b>				
Bone marrow blast, %	63.7 ± 22.4	66.5 ± 20.2	54.5 ± 27.6	0.281
WBC count, /L	29,169 ± 54,648	31,740 ± 61,259	20,599 ± 20,903	0.334
Platelet count, × 10 <sup>9</sup> /L	86.1 ± 72.0	81.3 ± 52.5	102.1 ± 117.9	0.564
Hemoglobin, g/dL	9.2 ± 2.0	9.0 ± 2.0	9.8 ± 2.2	0.292

Values are presented as median (range), number (%), or mean ± standard deviation.

AML, acute myeloid leukemia; NA, not applicable; WBC, white blood cell.

<sup>a</sup>Combined risk refers to risk stratification based on available next generation sequencing data including *TP53*, *ASXL1*, *RUNX1* mutation status.

Collectively, secondary AML, particularly post-MPN AML, was associated with higher fraction of the MPP-like LSC phenotype compared to its *de novo* counterpart.

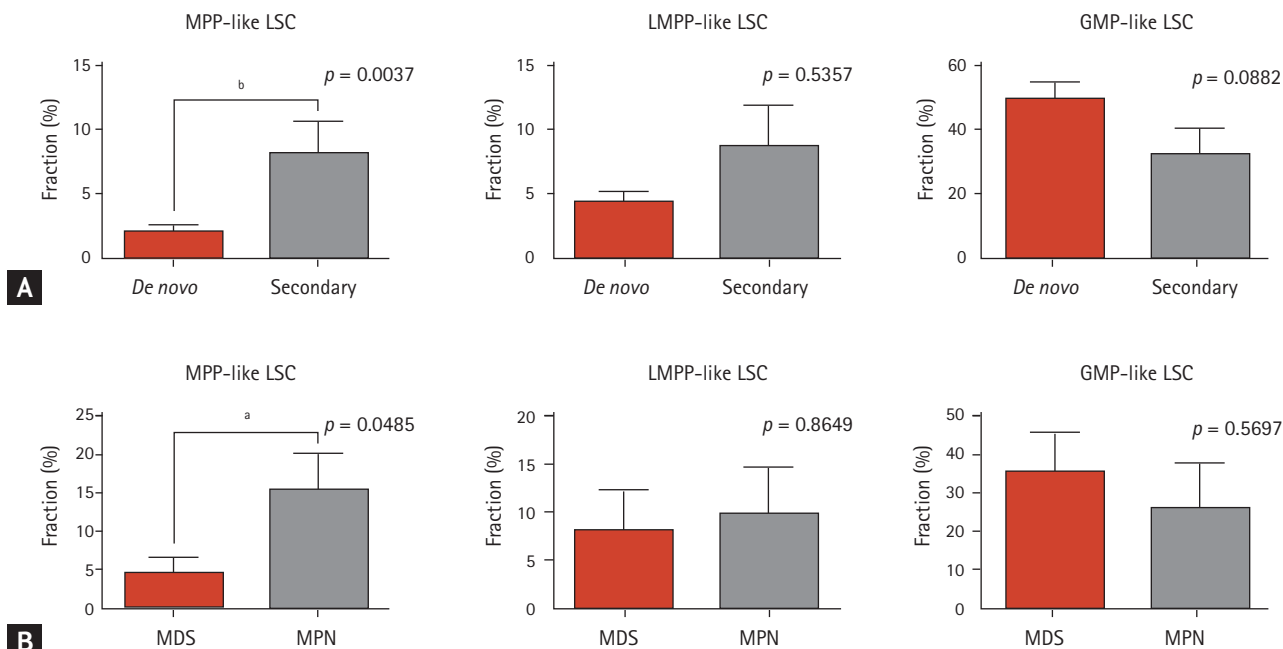
To determine the association between the age and LSC phenotype compositions, patients were divided into two group according to age at AML diagnosis at cutoff of 65 years (Supplementary Fig. 1). The CD34<sup>+</sup>CD38<sup>-</sup> cell fraction was higher for younger patients (i.e., ≤ 65 years), but there were no differences in LSC phenotypes between the two age groups.

### Risk stratification, survival and LSC phenotypes

While data on conventional cytogenetics was available in all patients, the NGS data was not available in 14 patients due to lack of quality-assured samples. The majority of the patients were classified as intermediate risk by both cytogenetics risk stratification and combined risk strat-

ification based on NGS data (Table 1). No correlation was seen between cytogenetics risk and LSC phenotypes (Supplementary Table 1). On the other hand, LMPP-like LSC was predominant in the poor risk group by combined risk stratification (*p* = 0.0052) (Fig. 3). This was true for both *de novo* AML (*n* = 28, *p* = 0.0401) and secondary AML (*n* = 10, *p* = 0.0871) patients (Supplementary Table 2).

For survival analyses, the patients were grouped according to their dominant LSC phenotypes (Fig. 4). There were ten patients in MPP-like LSC group, 11 in LMPP-like LSC group, and 31 in GMP-like LSC group. For the entire cohort, the mean RFS was 10.8 months and mean OS 12.3 months (Supplementary Table 3). Supplementary Fig. 2 represents the survival curves of the patients undergoing treatment per dominant LSC phenotype. In light of small number of patients with heterogeneous treatment schema, only those who un-



**Figure 2.** Leukemic stem cells (LSCs) phenotypes and acute myeloid leukemia (AML) subtypes. (A) The multipotent progenitor (MPP)-like LSCs are present in higher fraction in secondary AML patients compared to *de novo* AML patients. (B) Post-myeloproliferative neoplasm (MPN) AML is associated with higher fraction of MPP-like LSCs compared to post-myelodysplastic syndrome (MDS) AML. The graphs were drawn using a percentage that meets cell surface markers, which means each cell. Error bars indicate mean  $\pm$  standard deviation. LMPP, lymphoid primed multipotent progenitor; GMP, granulocyte-macrophage progenitor. <sup>a</sup> $p < 0.05$ , <sup>b</sup> $p < 0.01$ .

derwent treatment regardless of intensity were included for survival analyses. Furthermore, to compensate the innate differences in survival among different AMLs, *de novo* AML and secondary AML were analyzed separately. There was no correlation between RFS or OS and LSC phenotype in either group.

### Individual mutations and LSC phenotypes

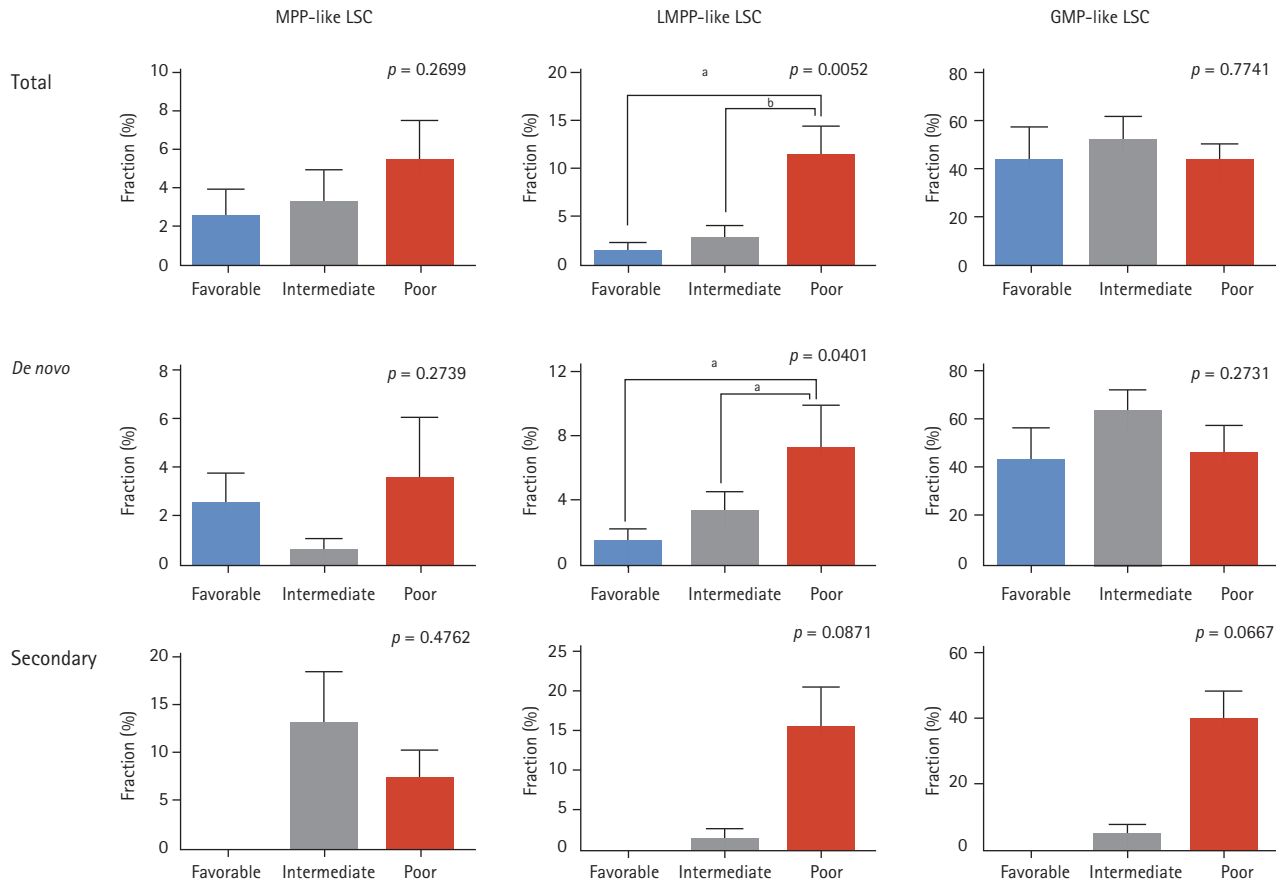
The mutational landscape of enrolled patients is shown in Fig. 4. We looked for specific mutations that were enriched in each LSC phenotype group. Mutations in *CEBPA* (6 out of 7), *DNMT3A* (4 out of 7), and *IDH1* (3 out of 5) were predominantly clustered in GMP-like LSC group. On the other hand, mutations in *NRAS* (4 out of 6) and *KRAS* (3 out of 4) were dominantly found in MPP-like LSC group, while mutations in *TP53* (4 out of 5) and *ASXL1* (5 out of 6) were often found in LMPP-like LSC group.

This observation was verified by comparing the LSC phenotype composition of specific mutation carriers versus non-carriers (i.e., wild-type). The differences in LSC expression per *CEBPA*, *DNMT3A*, and *IDH1* mu-

tations were not statistically significant. Interestingly, *KRAS* mutants showed higher fraction of MPP-like LSC compared to *KRAS* wild-types (Fig. 5). Also, the fraction of GMP-like LSC was significantly lower in *KRAS* mutants. *NRAS* mutants also showed similar tendencies, but the difference did not reach statistical significance. For *TP53* and *ASXL1* mutants, LMPP-like LSC was present in higher fraction. For *TP53* mutants, the expression of LMPP-like LSC was significantly higher by 3.4-folds ( $p = 0.0276$ ) compared to wild-type. *ASXL1* mutants also showed trends towards higher expression of LMPP-like LSC ( $p = 0.0654$ ).

### DISCUSSION

The importance of our study lies in that we showed for a given AML patients there is a dominant LSC phenotype and LSCs are associated with clinical outcomes, supporting the significance of cancer stem cell model for human AML. First of all, based on detailed characteriza-

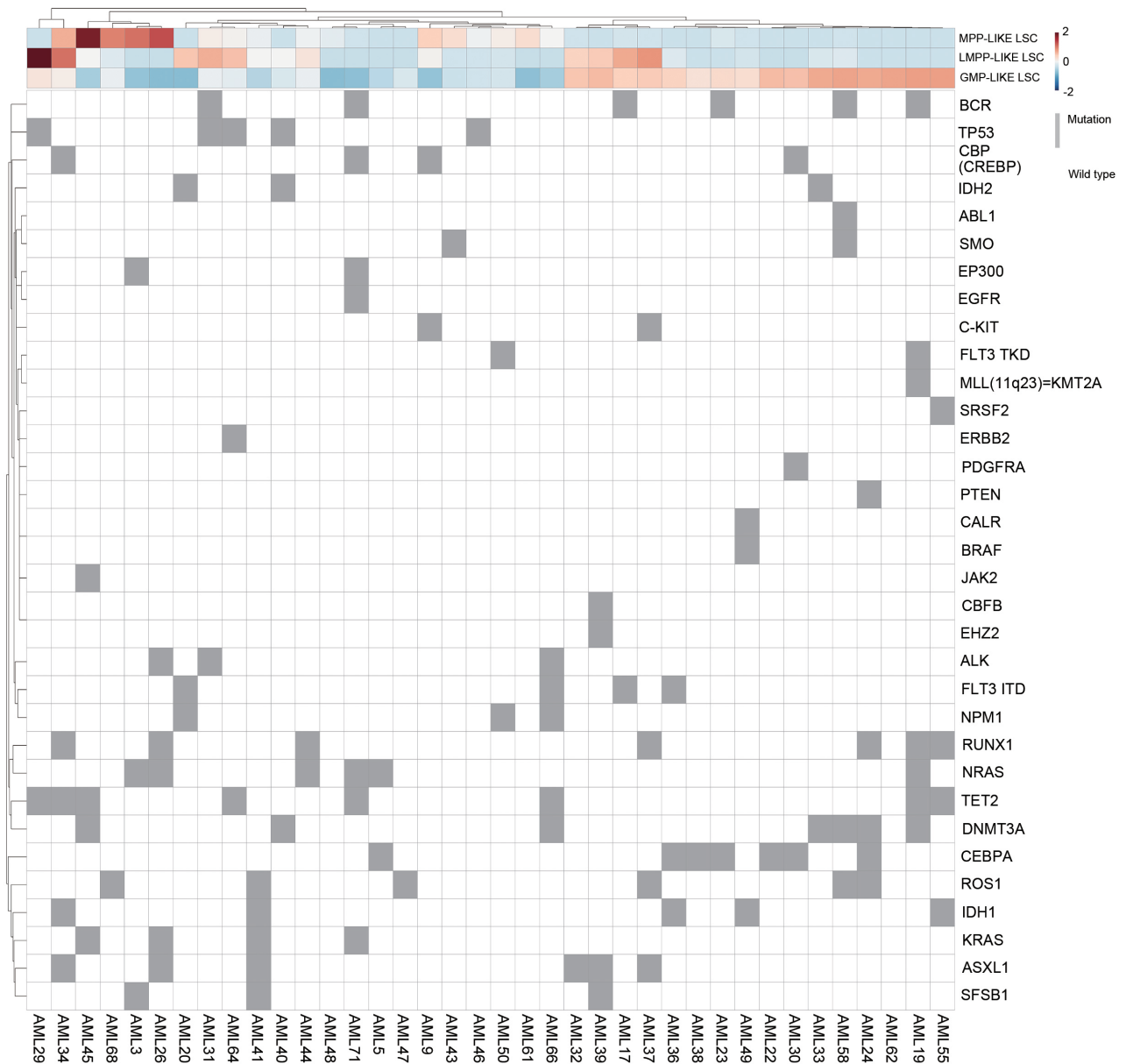


**Figure 3.** The association between acute myeloid leukemia risk stratification based on next generation sequencing data and leukemic stem cells (LSCs) phenotypes. Error bars indicate mean  $\pm$  standard deviations. MPP, multipotent progenitor; LMPP, lymphoid primed multipotent progenitor; GMP, granulocyte-macrophage progenitor. <sup>a</sup> $p < 0.05$ , <sup>b</sup> $p < 0.01$ .

tion of the surface immunophenotype of AML-LSCs we found that AML show evidence of a hierarchical cellular organization (Fig. 1). We also recognized that the composition of LSC phenotypes is associated with AML phenotypes. For example, secondary AML patients showed higher fraction of MPP-like LSCs compared to *de novo* AML patients (Fig. 2). Given that MPP-like LSCs are at the apex of LSC hierarchy, these cells are thought capable of initiating AML [4] and able to protect themselves from chemotherapy via ATP-binding cassette transporters [31-33]. In this regard, the higher expression of MPP-like LSCs could explain the poor response to standard treatments traditionally associated with secondary AML. Furthermore, the higher expression of MPP-like LSCs in post-MPN AML compared to post-MDS AML could explain the dismal prognosis associated with post-MPN AML, despite the relative indolent clinical course in

their chronic phase and the presence of druggable target.

Interestingly, age did not seem to affect the composition of LSCs (Supplementary Fig. 1). The expression of CD34<sup>+</sup>CD38<sup>-</sup> cell fraction was higher in younger patients, but this was probably due to the higher total number of cells in younger patients. When each LSC phenotype was considered, there was no difference between age groups. From clinical point of view, this finding implies two things: (1) this could explain the unsatisfactory RFS with hypomethylating agents (HMA) alone, and (2) this supports the role of allogeneic stem cell transplantation in elderly AML patients and at the same time highlights the unmet medical needs for better transplant techniques in such population. Whilst it is true that HMA has broadened the therapeutic options for elderly AML, since LSCs are reported to be more hypomethylated



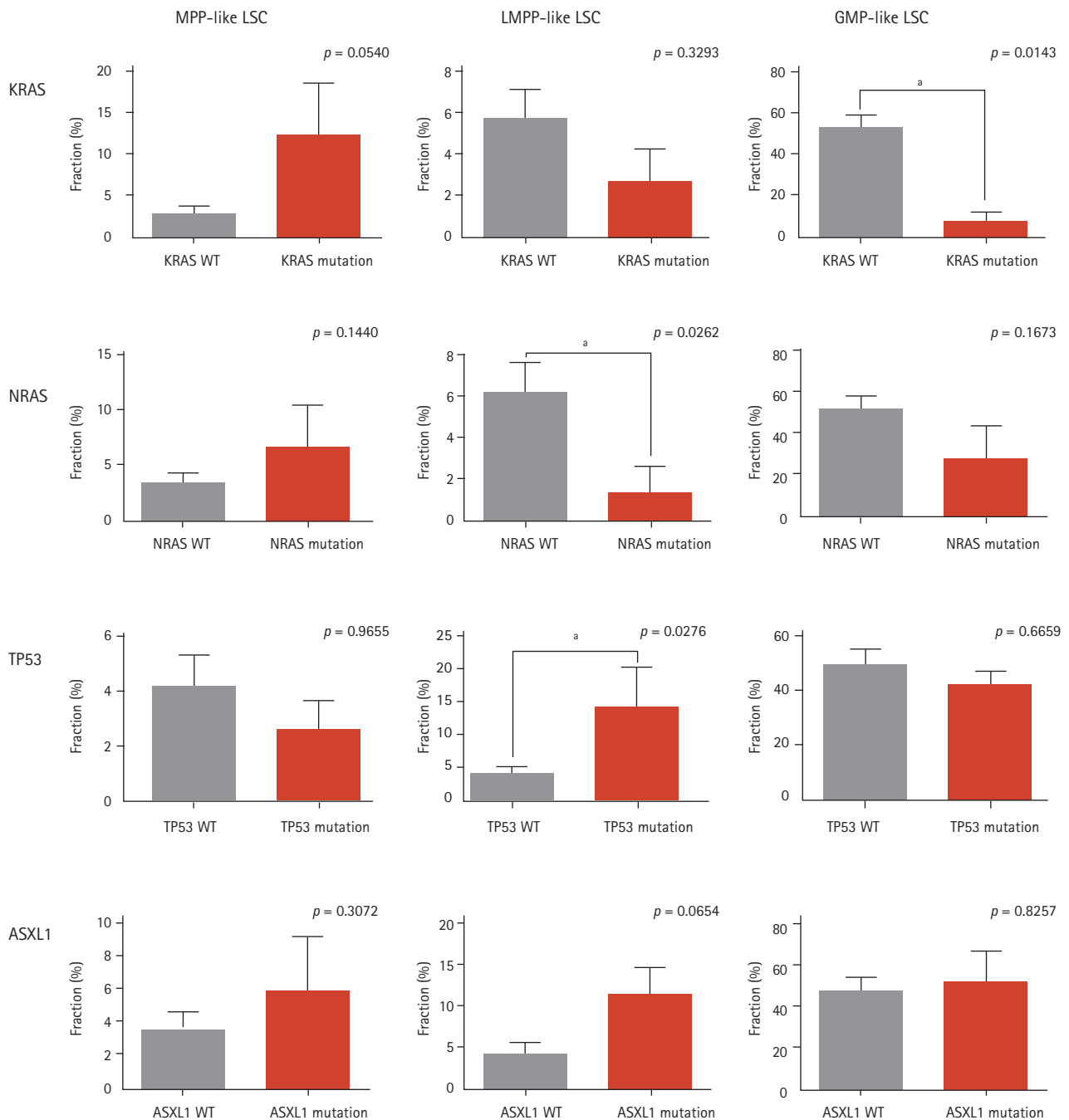
**Figure 4.** The dominant leukemic stem cell (LSC) phenotype and heatmap of mutations. MPP, multipotent progenitor; LMPP, lymphoid primed multipotent progenitor; GMP, granulocyte-macrophage progenitor; AML, acute myeloid leukemia.

compared to leukemic blasts to begin with, these agents cannot eradicate LSCs hence the short RFS [34]. Thus, allogeneic stem cell transplantation remains the only currently available option of eradicating LSCs [6].

Moving onto the association between LSCs, gene mutations and prognostic implications, we found that LSC phenotypes by themselves do not predict survival out-

comes but when taken in together with gene mutations they carry certain predictive values. It is true that AML risk stratification has been refined considerably with the advent of NGS. However, even within the same risk group some patients do much worse than others. *KRAS*/*NRAS* mutations deserve special attention in this regard. It has been found that there is a sequential order for the





**Figure 5.** The relationship between individual gene mutation and leukemic stem cells (LSCs) phenotypes. Error bars indicate mean  $\pm$  standard deviations. MPP, multipotent progenitor; LMPP, lymphoid primed multipotent progenitor; GMP, granulocyte-macrophage progenitor; WT, wild type. <sup>a</sup> $p < 0.05$ .

acquisition of AML defining gene mutations during leukemogenesis. Somatic mutations in signaling pathways that drive proliferation, such as *KRAS*/*NRAS*, are considered later events in AML transformation [35,36].

In our samples the variant allele frequencies of *KRAS*/*NRAS* mutations were less than 10% in most patients, confirming these mutations are acquired later during clinical course. Whether *RAS* mutations confer prog-

nostic values or not is yet to be settled [37-39], thus they are not incorporated in the current risk stratification schema [14]. Although the relationship is somewhat tenuous, we showed that *RAS* mutants might actually be associated with worse prognosis because they tend to carry higher fraction of MPP-like LSCs (Fig. 5).

The obvious limitation of our study includes the small number of patients included. Another pitfall of our study is that the NGS was carried out with DNA extracted from bone marrow samples at diagnosis, rather than with the gated LSCs. However, we believe the value of our study lies in that it contains adequate scientific clues to inspire future studies regarding the genetics of LSC and their prognostic values. Given that merging of data on genomic and epigenomic landscape of AML with biology of LSCs is still in its infancy, our data might provide certain clues for further investigations.

In conclusion, our findings provide better insights into the characteristics and clinical implications of LSC. Although in a small scale, we provide evidence that specific LSC phenotypes are associated with certain mutations thus should be in the AML risk stratification process.

### KEY MESSAGE

1. Leukemic stem cells (LSCs) are associated with clinical outcomes in acute myeloid leukemia (AML) patients, supporting the significance of cancer stem cell model for human AML.
2. The composition of LSC phenotypes is associated with AML phenotypes. For example, secondary AML patients showed higher fraction of multipotent progenitor-like LSCs compared to *de novo* AML patients.
3. LSC phenotypes, when taken in together with gene mutations, carry certain predictive values. This can explain why even within the same risk group some patients do much worse than others.

### Conflict of interest

No potential conflict of interest relevant to this article was reported.

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**Supplementary Table 1. LSC phenotype according to AML risk stratification by cytogenetics only**

Conventional risk	Low	Intermediate	High	<i>p</i> value
Total no.	6	38	8	NA
MPP-like LSC	4.30 ± 5.06	2.52 ± 4.59	7.52 ± 8.84	0.0675
LMPP-like LSC	8.42 ± 8.31	3.89 ± 5.80	10.07 ± 11.58	0.0273
GMP-like LSC	28.46 ± 38.62	49.17 ± 36.30	41.10 ± 15.12	0.3196
<i>De novo</i> AML no.	6	32	2	NA
MPP-like LSC	4.30 ± 5.06	1.19 ± 1.95	9.54 ± 8.56	0.0500
LMPP-like LSC	8.42 ± 8.31	3.74 ± 5.29	2.04 ± 1.54	0.1748
GMP-like LSC	28.46 ± 38.62	54.30 ± 35.15	36.30 ± 8.78	0.1742
Secondary AML no.	0	6	6	NA
MPP-like LSC	NA	9.62 ± 7.80	6.85 ± 9.62	0.4848
LMPP-like LSC	NA	4.68 ± 8.67	12.75 ± 12.36	0.0921
GMP-like LSC	NA	21.78 ± 31.90	42.70 ± 17.10	0.2403

Values are presented as mean ± SD.

LSC, leukemic stem cell; AML, acute myeloid leukemia; NA, not applicable; MPP, multipotent progenitor; LMPP, lymphoid primed multipotent progenitor; GMP, granulocyte-monocyte progenitor.



**Supplementary Table 2. LSC phenotype according to AML risk stratification based on NGS data**

Combined risk <sup>a</sup>	Low	Intermediate	High	p value
Total no.	7	19	12	NA
MPP-like LSC	2.58 ± 3.46	3.32 ± 6.90	5.55 ± 6.57	0.2699
LMPP-like LSC	1.54 ± 2.23	2.95 ± 4.72	11.47 ± 10.20	0.0052
GMP-like LSC	44.46 ± 34.23	52.18 ± 39.18	43.59 ± 23.24	0.7741
<i>De novo</i> AML no.	7	15	6	NA
MPP-like LSC	2.58 ± 3.46	0.73 ± 1.69	3.74 ± 5.99	0.2739
LMPP-like LSC	1.54 ± 2.23	3.37 ± 5.14	7.51 ± 6.36	0.0401
GMP-like LSC	44.46 ± 34.23	64.89 ± 33.84	47.31 ± 27.00	0.2731
Secondary AML no.	0	4	6	NA
MPP-like LSC	NA	13.04 ± 10.61	7.36 ± 7.15	0.4762
LMPP-like LSC	NA	1.37 ± 2.49	15.44 ± 12.29	0.0871
GMP-like LSC	NA	4.50 ± 5.10	39.87 ± 20.63	0.0667

Values are presented as mean ± SD.

LSC, leukemic stem cell; AML, acute myeloid leukemia; NGS, next generation sequencing; NA, not applicable; MPP, multipotent progenitor; LMPP, lymphoid primed multipotent progenitor; GMP, granulocyte-monocyte progenitor.

<sup>a</sup>Combined risk refers to risk stratification based on available next generation sequencing data including TP53, ASXL1, RUNX1 mutation status.

**Supplementary Table 3. Treatment schema and outcomes**

Variable	Total	<i>De novo</i> AML	Secondary AML
<b>Initial treatment</b>			
No treatment	3 (5.8)	1 (2.5)	2 (16.7)
Low intensity treatment <sup>a</sup>	18 (34.6)	14 (35.0)	4 (33.3)
Intensive induction <sup>a</sup>	31 (59.6)	25 (62.5)	6 (50.0)
CR achievement with induction <sup>b</sup>	23 (74.2)	20 (80.0)	3 (50.0)
Induction TRM	4 (12.9)	4 (16.0)	0
Upfront HSCT	13 (25.0)	8 (20.0)	5 (41.7)
<b>HSCT donor</b>			
Matched related	5 (38.5)	5 (62.5)	0
Matched unrelated	4 (30.8)	3 (37.5)	1 (20.0)
Others <sup>c</sup>	4 (30.8)	0	4 (80.0)
<b>HSCT conditioning</b>			
Myeloablative	7 (53.8)	3 (37.5)	4 (80.0)
Reduced intensity	6 (46.2)	5 (62.5)	1 (20.0)
Relapse	29 (55.8)	23 (57.5)	6 (50.0)
Relapse free survival, mon	10.8 ± 8.7	11.3 ± 8.8	9.0 ± 8.6
Death	18 (34.6)	14 (35.0)	4 (33.3)
Overall survival, mon	12.3 ± 8.8	13.2 ± 8.8	9.3 ± 8.4

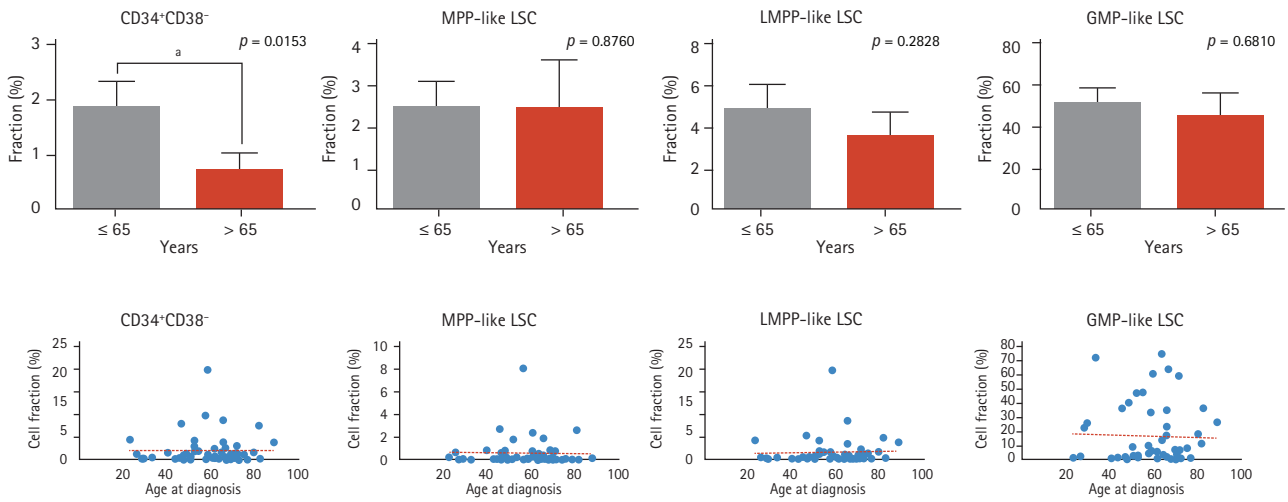
Values are presented as number (%) or mean ± SD.

AML, acute myeloid leukemia; CR, complete remission; TRM, treatment related mortality; HSCT, hematopoietic stem cell transplantation; NA, not applicable.

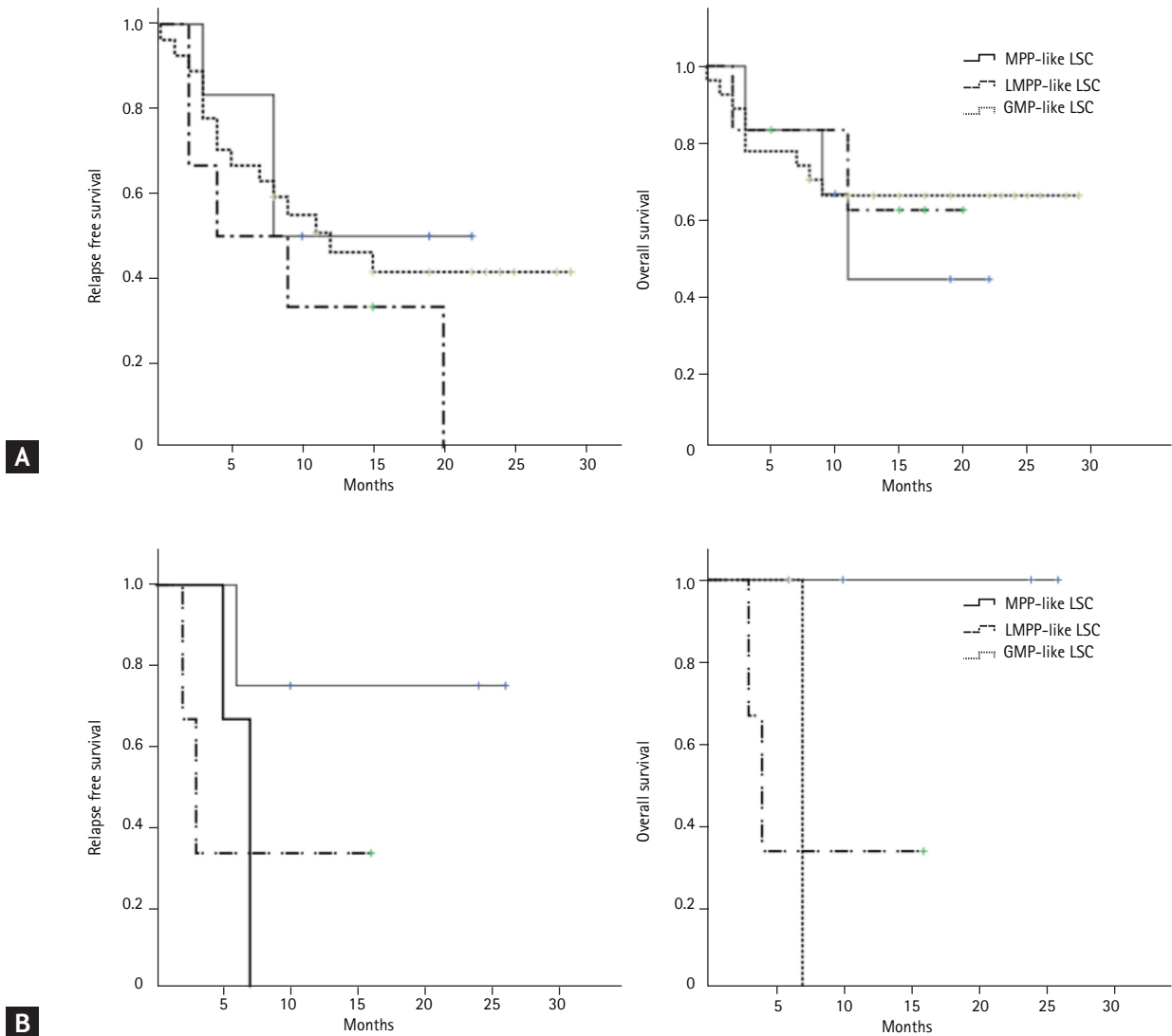
<sup>a</sup>Low intensity treatment refers to hypomethylating agents, venetoclax, low dose cytarabine; while intensive induction refers to standard 3 + 7 induction chemotherapy.

<sup>b</sup>For patients undergoing standard induction chemotherapy.

<sup>c</sup>Others include mismatched related donors and cord blood.



**Supplementary Figure 1.** The age at acute myeloid leukemia diagnosis and leukemic stem cell (LSC) phenotypes. MPP, multi-potent progenitor; LMPP, lymphoid primed multipotent progenitor; GMP, granulocyte-monocyte progenitor.



**Supplementary Figure 2.** Survival analyses. (A) Relapse free survival (RFS) and overall survival (OS) of 39 *de novo* acute myeloid leukemia (AML) patients undergoing treatment. (B) RFS and OS of 10 secondary AML patients undergoing treatment. MPP, multipotent progenitor; LSC, leukemic stem cell; LMPP, lymphoid primed multipotent progenitor; GMP, granulocyte-monocyte progenitor.