

Serum LC-MS based untargeted metabolomics machine learning identified CLL/SLL patients with different metabolic features and predict TTFT

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OBJECTIVES

- Use serum LC-MS based untargeted metabolomics to reveal serum metabolites features in CLL/SLL patients.
- Reveal the relationship between biological and clinical characteristics and serum metabolites.
- Explore prognosis of CLL/SLL patients with different metabolic modules.

CONCLUSIONS

- Serum metabolome differs among CLL/SLL patients and can be divided into three metabolic clusters.
- Patients assigned to different metabolic clusters showed different biological features.
- Patients belong to cluster2 showed better prognosis in terms of TTFT free survival in watch and wait cohort.

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INTRODUCTION

- Metabolic reprogramming is a hallmark of cancer and may play a pivotal role in driving disease progression. However, studies focused on serum metabolite profiles of chronic lymphocytic leukemia/small lymphocytic lymphoma (CLL/SLL) patients are scarce.
- Besides, heterogenous metabolite profiles and whether biological characteristics will influence serum metabolome is also uncertained.
- Prognosis of metabolome in watch and wait cohort were explored.

METHODS

- Between Oct 2020 and Nov 2024, newly diagnosed CLL/SLL patients in Jiangsu Province Hospital with available serum samples were enrolled in the cohort (N=182) among whom 68 pts were asymptomatic and 114 pts were symptomatic according to iwCLL2018 criteria.
- All serum samples were collected prior to any treatment and LC-MS based untargeted metabolomics analysis were performed.
- Unsupervised Learning was used to demonstrate heterogenous composition of serum metabolome.
- PCA(principal component analysis) and Partial least squares discriminant analysis (PLS-DA) was used to visualize differences among metabolic clusters.
- Baseline clinical and biological features were compared between different clusters.

RESULTS

- Serum untargeted metabolomics based on LC-MS method totally identified 1157 metabolites.
- Baseline clinical and biological characteristics is shown in Table1. 37.4% patients are asymptomatic and 62.6% patients are asymptomatic are symptomatic. 38.4% patients are IGHV unmutated. 6.6% patients have 17p deletion, 20.0% patients have TP53 mutation.
- By unsupervised learning method, patients enrolled in this study were divided into 3 metabolic clusters(Figure2). Both PCA and PLSDA visualized the distinct separation among different metabolic clusters(Figure3A-B).Metabolite set enrichment analysis (MESA) showed these metabolites enriched in glycerophospholipid metabolism, nitrogen metabolism, purine metabolism and arginine biosynthesis.
- Clinical and biological characteristics were compared between different clusters and clinical characteristics showed no difference among 3 clusters. However, cluster 1 had higher proportion of patients with KMT2D mutation and EGR2 mutation. Cluster 3 had higher proportion of patients with TP53 mutation(Table2).
- Median follow-up was 652 days in asymptomatic cohort(watch and wait cohort) . Patients featured with metabolic cluster 2 had higher 2 year TTFT free survival, suggesting serum metabolome may reflect metabolic clusters with biological features and prognosis.

ACKNOWLEDGMENTS

We would like to thank the physicians and staff who participated in this study. This study was supported by the National Natural Science Foundation of China (Grant No. 82170166, 82100207), the Youth Talent support Project in Jiangsu Province (Grant No. YNRCQN035), the Specialized Diseases Clinical Research Fund of Jiangsu Province Hospital (Grant No. 303100160AA25).

DISCLOSURES

None

Figure1. Study design.

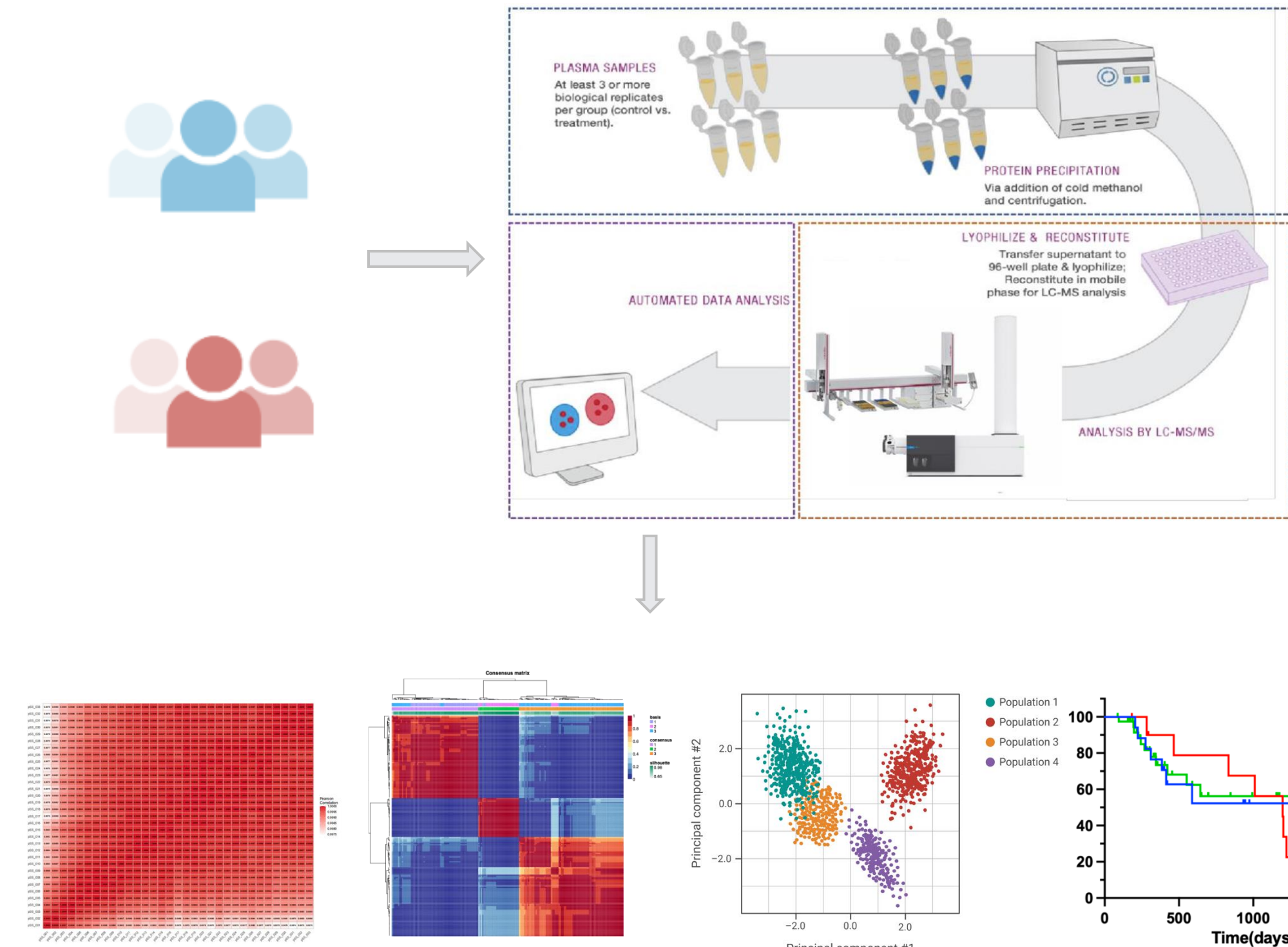


Table1.Clinical and biological characteristics of the cohort.

Characteristics	Overall (N = 182)
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Age, Mean ± SD	59 ± 12
Sex, n (%)	
Male	112 (61.5%)
Female	70 (38.5%)
BMI, n (%)	
<18.5	14 (7.7%)
18.5–24	85 (46.7%)
24–28	61 (33.5%)
≥28	10 (5.5%)
Missing	12 (6.6%)
B2MG, Mean ± SD	3.66 ± 1.84
LDH, Mean ± SD	220 ± 123
WBC, Mean ± SD	66 ± 72
ALC, Mean ± SD	55 ± 63
Disease status, n (%)	
PD	114/182 (62.6%)
WW	68/182 (37.4%)
Binet stage, n (%)	
A	45/182 (24.7%)
B	66/182 (36.3%)
C	71/182 (39.0%)
Bulky(>5cm), n (%)	
Yes	98/159 (61.6%)
No	61/159 (38.4%)
IGHV mutation, n (%)	
mutated	156/167 (93.4%)
unmutated	11/167 (6.6%)
Del(17p), n (%)	
No	87/154 (56.5%)
Yes	67/154 (43.5%)
Del(13q14), n (%)	
No	138/157 (87.9%)
Yes	19/157 (12.1%)
Del(11q), n (%)	
No	113/150 (75.3%)
Yes	37/150 (24.7%)
Trisomy12, n (%)	
No	113 (62.1%)
Yes	37 (20.3%)
Del(6q23), n (%)	
No	123/128 (96.1%)
Yes	5/128 (3.9%)
CK, n (%)	
No	139/179 (77.7%)
Yes	40/179 (22.3%)
TP53mut, n (%)	
No	128/160 (80.0%)
Yes	32/160 (20.0%)
NOTCH1mut, n (%)	
No	132/156 (84.6%)
Yes	24/156 (15.4%)
SF3B1mut, n (%)	
No	141/156 (90.4%)
Yes	15/156 (9.6%)
MYD88mut, n (%)	
No	138/157 (87.9%)
Yes	19/157 (12.1%)
ATMmut, n (%)	
No	135/156 (86.5%)
Yes	21/156 (13.5%)
KMT2Dmut, n (%)	
No	139/157 (88.5%)
Yes	18/157 (11.5%)
BIRC3mut, n (%)	
No	149/156 (95.5%)
Yes	7/156 (4.5%)
EGR2mut, n (%)	
No	145/156 (92.9%)
Yes	11/156 (7.1%)

Figure2.CLL/SLL patients can be divided into three metabolic clusters by consensus NMF.

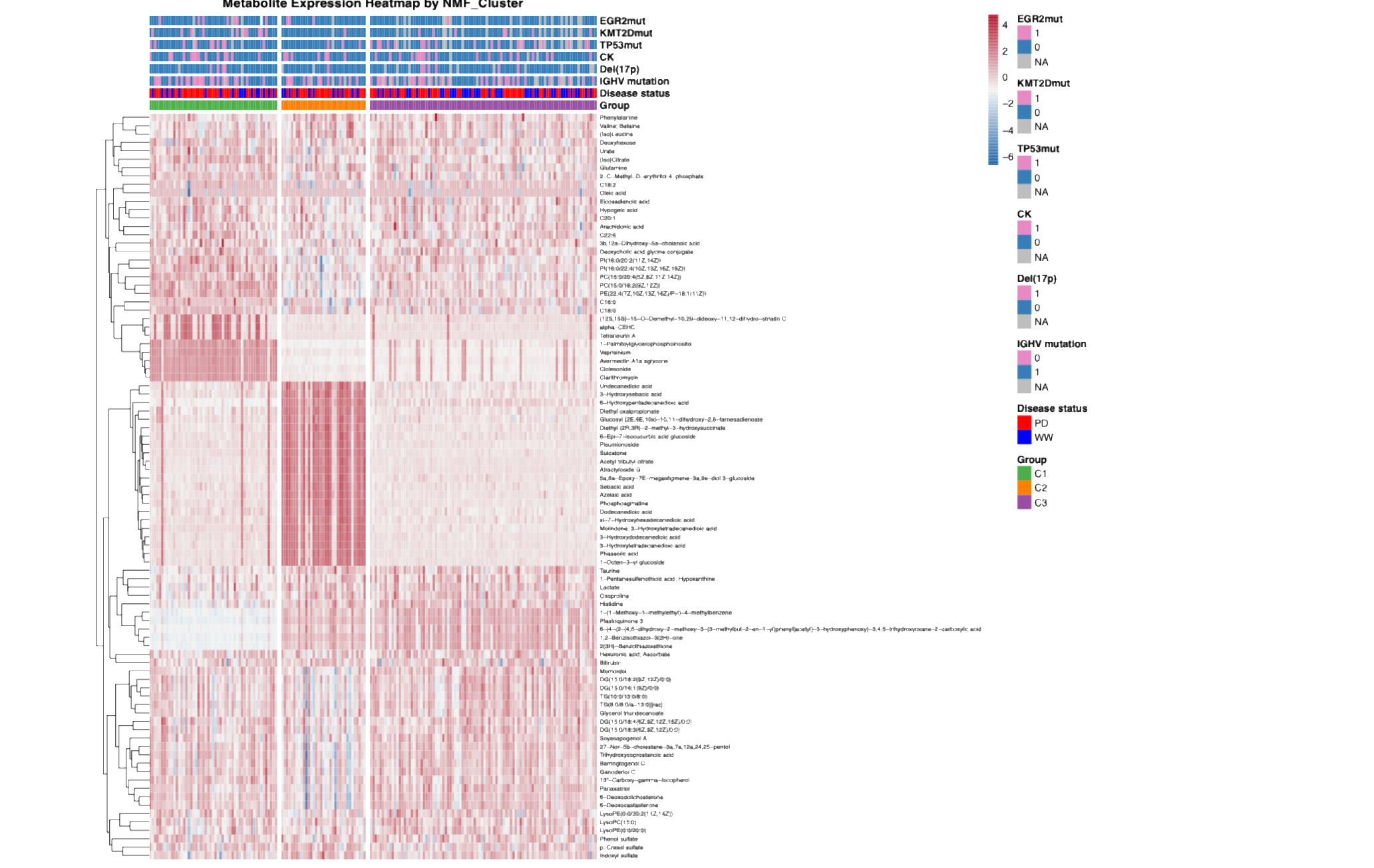


Figure3. Serum Metabolome differs between different metabolic clusters.

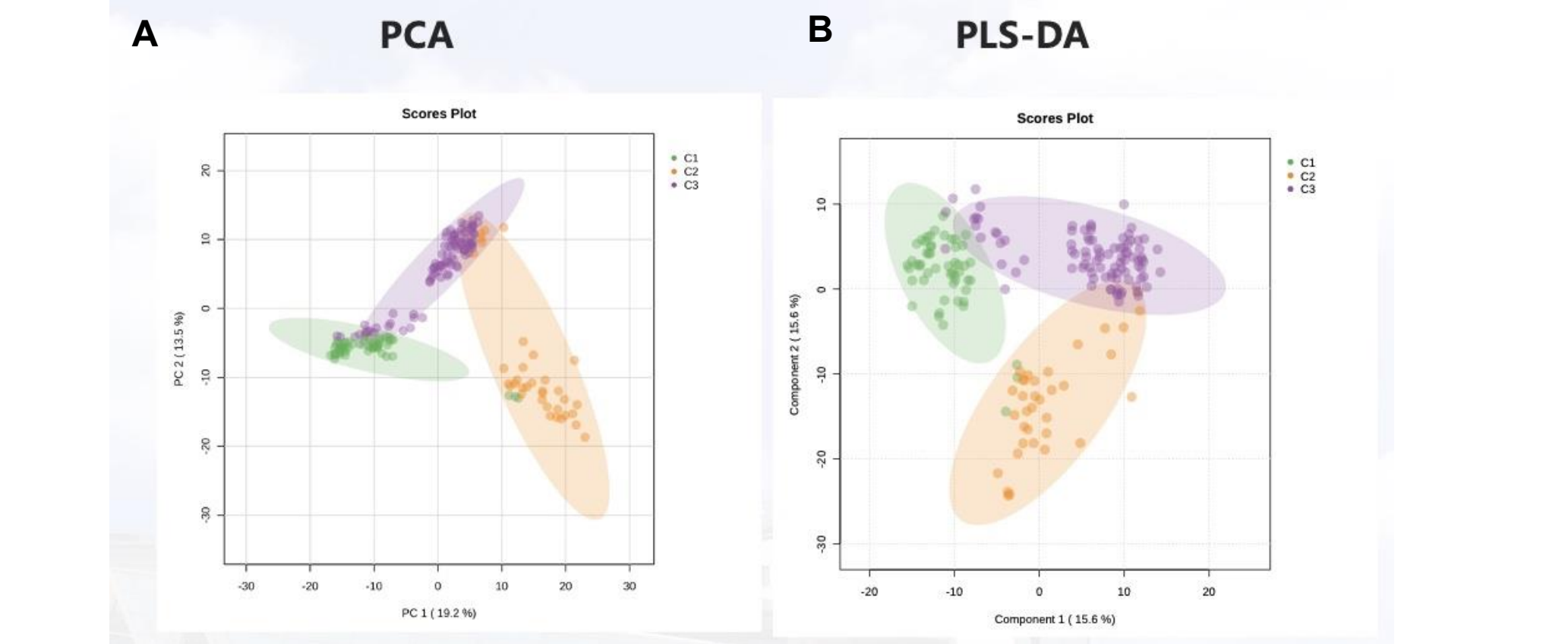


Table2. Biological characteristics differ between clusters

Percentage(%)	C1(N=53)	C2(N=35)	C3(N=94)	C1 vs C2	C1 vs C3	C2 vs C3
IGHV-UM	43.1%(22/51)	34.4%(11/32)	36.8%(28/76)	0.427 ¹	0.579 ¹	0.807 ¹
TP53mut	20%(10/50)	8.8%(3/34)	25%(19/76)	0.165 ¹	0.514 ¹	0.050 ¹
Del(17p)	7.8%(4/51)	3.0%(1/33)	7.2%(6/83)	0.644 ²	0.896 ¹	0.671 ²
Trisomy	28.6%(14/49)	20.0%(6/30)	23.9%(17/71)	0.395 ¹	0.569 ¹	0.666 ¹
CK	26.4%(14/53)	20.0%(7/35)	20.9%(19/91)	0.490 ¹	0.446 ¹	0.913 ¹
NOTCH1mut	14.6%(7/48)	6.1%(2/33)	20.0%(15/75)	0.298 ²	0.445 ¹	0.067 ¹
KMT2Dmut	20.0%(10/49)	2.9%(1/33)	9.2%(7/75)	0.025 ²	0.083 ¹	0.431 ²
EGR2mut	14.3%(7/48)	8.8%(3/33)	1.3%(1/75)	0.515 ²	0.006 ²	0.087 ²
ATMmut	14.6%(7/48)	6.1%(2/33)	16.0%(12/75)	0.298 ²	0.832 ¹	0.219 ²
MYD88mut	12.2%(6/49)	15.2%(5/33)	10.7%(8/75)	0.749 ²	0.786 ¹	0.531 ²
SF3B1mut	12.5%(6/48)	12.1%(4/33)	6.7%(5/75)	>0.999 ²	0.336 ²	0.451 ²

Figure4. Prognosis differs between different metabolic clusters in watch and wait cohort.

