Limits and Applications of Genomic Analysis of Circulating Tumor DNA as a Liquid Biopsy in Asymptomatic Forms of Multiple Myeloma

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The multi-step evolution of monoclonal gammopathies provides that a small clonal population of long-lived bone marrow (BM) plasma cells (PCs) arises with premalignant features (monoclonal gammopathy of undetermined significance -MGUS), expands its burden becoming an indolent cancer (smoldering multiple myeloma -SMM), and finally develops into an overt clinical form requiring treatment (multiple myeloma-MM). Based on long-term follow-up studies, MGUS progresses at a constant and low rate of 1%/yr. The risk of transforming from SMM to MM is more heterogeneous. Several risk factors have been identified to predict SMM progression at diagnosis, and scorers have been proposed to identify high-risk patients with >50% risk at 2 years. These would be patients misclassified as SMM while actually carrying active MM, just not recognized based on current diagnostic criteria. Furthermore, these studies focus on population averages and do not account for the variability of individual patients’ trajectories, nor for a change in their risk over years of observation. For example, a recent prospective sampling study highlighted how changes in the dynamics of monoclonal proteins or free serum light chains in MGUS can identify patients converting from low- to high-risk categories years before MM diagnosis. Similarly, in SMM a change in hemoglobin and/or monoclonal protein amount predicted increased risk of progression to active myeloma even in patients who were diagnosed as low risk. This highlights the need for better diagnostic stratification, and improved serial follow-up, to identify high-risk patients that may benefit from intensive follow-up or early treatment. Complementing the clinical observations described above, recent genome/exome-wide sequencing efforts in serial samples from SMM-MM patients allowed the identification of different clonal evolution patterns that recapitulate distinct clinical subsets of SMM patients, suggesting that genomic analysis can improve current risk scores to facilitate the identification of high-risk SMM patients with an “MM-like” genomic profile. Clearly, for this to be conceivable, genomic analysis of asymptomatic patients must be feasible in routine clinical practice, also allowing repeated sampling and accounting for MM spatial heterogeneity. Data produced by ours and other groups indicated that circulating cell-free DNA (cfDNA) is a valid source of tumor DNA and well reflects the mutational profile of neoplastic PCs in MM. Here, we tested whether this is also true in asymptomatic disease forms characterized by a lower tumor burden.

To this aim, we analyzed 104 samples using ultra-low pass whole-genome sequencing (ULP-WGS) and ultra-deep targeted next-generation sequencing (NGS) of 14 MM driver genes. Samples were derived from 65 patients, respectively 15 MGUS, 33 SMM, and 17 MM patients (11 of whom already analyzed). In addition, the analyses were also performed in the genomic DNA (gDNA) from BM CD138+ purified PCs of 39 patients (6/15 MGUS, 16/33 SMM and all the 17 MM cases).

We first investigated the 39 BM gDNA samples. Here our targeted panel, through sequencing at a depth of 5435×, identified a median of 0 mutations per patient in MGUS (range: 0–2), 0.5 in SMM (range: 0–3) and 2 in MM (range: 0–4). Overall, we found at least one mutation in 33%, 50% and 82% of MGUS, SMM and MM patients, respectively (SDC Fig. 1A, Supporting information). This is an open access article distributed under the terms of the Creative Commons Attribution-Non Commercial-No Derivatives License 4.0 (CCBY-NC-ND), where it is permissible to download and share the work provided it is properly cited. The work cannot be changed in any way or used commercially without permission from the journal. HemaSphere (2020) 4:4(e402). http://dx.doi.org/10.1097/HSS.000000000000402. Received: 11 February 2020 / Accepted: 27 April 2020
MGUS, 49% in SMM and 58% in MM, confirming how BM PCs are progressively more enriched in tumor cells as the disease evolves (SDC Fig. 1B, http://links.lww.com/HS/A87). IchorCNA returned genome-wide copy number profiles consistent with the known CNAs in the majority of samples (91% confirmed by FISH). Overall, 50% MGUS, 100% SMM and 94% MM showed at least one CNA (SDC Fig. 1A, http://links.lww.com/HS/A87).

We next turned our attention on cfDNA from peripheral blood (PB) of all the 65 patients included in the study. From an average of 6 ml of plasma, we recovered a mean of about 3326 haploid genome equivalents/mL. cfDNA concentrations were comparably distributed between MGUS and SMM, and significantly (on average 2.8-fold) lower than in MM (Kruskal-Wallis rank sum test \( p = 0.03 \); Kruskal-Wallis rank-sum test). B: Percentage of BM PC infiltration (pink bars) and tumor fraction estimates of matched cfDNA (green bars) and CD138+ gDNA (blue bars) samples in the 22 patients subjected to BM biopsy. The black and yellow lines track the 12% BM tumor fraction and 10% BM PC infiltration thresholds, respectively. C: Overview of identified gene mutations. Gray squares indicate no mutations. Mutations detected in plasma cfDNA and confirmed in tumor gDNA are filled in green; mutations detected in plasma cfDNA only are filled in blue. Each column represents one tumor sample and each row represents one gene. The fraction of tumors with mutations in each gene is plotted (right). MGUS samples are indicated in yellow, SMM samples in lilac. IDs of patients in whom ULP-WGS of cfDNA samples did not detect any circulating tumor DNA are framed.

For the 22 MGUS/SMM patients with matched BM-PB samples, we asked whether a correlation exists between tumor load in the 2 compartments. We did not identify a significant linear correlation, but we found that a greater tumor fraction of gDNA from BM PCs corresponded to an increase in the likelihood of finding tumor-specific cfDNA in PB. In particular, a ROC analysis indicated that ULP-WGS had the best performance in detecting circulating tumor DNA when the tumor fraction in the matched BM biopsy was greater than 12%; indeed, above this threshold, ichorCNA detected tumor material in the cfDNA of 12 out of 16 patients (Fig. 1B). Similarly, tumor-specific cfDNA was found in 13/16 asymptomatic patients with a total BMPC infiltrate \( \geq 10\% \), suggesting that plasma cfDNA can be a possible source of tumor DNA in SMM. As expected, BM PC infiltration and BM tumor fraction were linked by a positive (albeit weak) correlation (SDC Fig. 3, http://links.lww.com/HS/A87).

We performed mutational analysis in PB cfDNA of the 48 asymptomatic patients (Fig. 1C). Samples were sequenced to an average of 4523 \( \times \) . Considering the overall results of cfDNA/BM PC gDNA sequencing, 20 patients (42%) carried at least one non-synonymous somatic mutation. The most frequently mutated genes were \( E G R 1 \) and \( T R A F 3 \) (each in 4 patients), followed by \( C Y L D, B R A F \) and \( F A M 46 C \) (3/48), \( I R F 4, N R A S \) and \( D I S 3 \) (2/48 each), and \( K R A S, P R D M 1, T P 5 3, S P I 4 6, \) and \( C C N D 1 \) (1/48 each). \( Z N F 4 6 2 \) was never found mutated. Identified substitutions and indels comprised both previously described oncogenic mutations and novel variants. It is worth noting that gene mutations in cfDNA were found in eight of the 22/48 patients in whom ULP-WGS did not detect circulating tumor DNA. In all
likelihood, this occurred because the ultra-deep targeted sequencing approach has a much greater sensitivity than ULP-WGS. Focusing on the 22 patients of whom we sequenced also tumor gDNA, all the mutations found in the BM were confirmed in plasma with the exception of 4 mutations identified only in the BM biopsy (FAM46C T354Hfs*12 in ID86, KRAS G12D and Q61H in patient ID37, and CYLD W847fs* in ID14) (Fig. 1C, blue). Of note, among all the variants called in the tumor tissue, these four mutations had at the lowest variant allelic frequency (VAF) (VAF < 10%; SDC Fig. 4A, http://links.lww.com/HS/A87); therefore, considering how cellular DNA is diluted in PB, their failed detection in cfDNA was expected. We also observed four mutations in cfDNA not confirmed in BM gDNA (EGFR A243T, IRF4 P227S and NRAS S17Afs* in patient ID28, and BRAF L79Q in ID89). Presumably, cfDNA carries mutations arising in foci distant from the BM biopsy site, as myeloma notoriously displays spatial heterogeneity.17 Overall, for these 22 cases, we did not find correlation between the VAFs computed in tumor gDNA and those in plasma cfDNA (SDC Fig. 4B, http://links.lww.com/HS/A87).

We finally asked whether ULP-WGS could identify genome-wide CNAs in PB cfDNA. However, in this setting the concordance between PB and BM results was much lower than that of mutation analysis (Fig. 2A and SDC Fig. 5, http://links.lww.com/HS/A87): taking any CNA independently, we found a total of 241 abnormalities in BM and PB combined, of which only 14% were shared in the asymptomatic samples. Interestingly, the excess of private variants was mostly found as BM-only. Conversely, the same analysis performed in paired PB-BM samples from MM patients showed a more than three times higher rate of concordance of 51% (Fig. 2B), suggesting that the low amount of tumor-specific cfDNA may be a limiting factor for the use of PB ULP-WGS in the asymptomatic setting.

In conclusion, here we report on the promises and limitations of mini-invasive sampling of PB cfDNA for an extensive genomic characterization of asymptomatic MM stages. Our data support...
the notion that the lower tumor mass of asymptomatic stages likely correlates with a lower cfDNA tumor fraction. This is in keeping with the better performance of deep mutational analysis than CNA analysis performed at low depth with ULP-WGS in this setting. Even with the more sensitive method, more than half of asymptomatic patients did not show any mutation in cfDNA, likely reflecting a lower mutational load of these stages (clearly, by analyzing a larger gene panel the mutation rate may be higher, but reasonably not to a substantial extent, since the 14 sequenced genes are among the most recurrently mutated in MM). Our data confirm that ichorCNA performance is limited in the setting of very low tumor fractions. However, since ichorCNA estimates the percentage of tumor-derived cfDNA based on the detection of somatic CNAs (as non-hyperdiploid, IGH-translocated asymptomatic patients is within reach to improve prognostication in a minimvasive manner.

To optimize the use of liquid biopsy, filling the limits shown by the adopted experimental strategy and increasing sensitivity, a deeper genome-wide approach could be used. However, the cost-benefit ratio should be assessed also at the economic level. Alternatively, in addition to mutational hotspots, even DNA gain and loss hotspots could be targeted by a capture approach and sequenced at higher depth, as already reported on MM PC gDNA.18,19 Finally, to fully exploit the potential of the liquid biopsy as a surrogate of BM sampling, a further upgrade would be required to interrogate plasma also for IGH translocations (by capturing the whole IGH locus or a portion thereof where most of the breakpoints are clustered).18,19 Although the tested cfDNA analysis approach is not ready for prime time to evaluate MGUS and SMM, we believe that with the refinement of these aspects, a thorough genomic assessment of asymptomatic patients is within reach to improve prognostication in a minimvasive manner.

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