

Different Genomic Clusters Impact on Responses in Advanced Biliary Tract Cancer Treated with Cisplatin Plus Gemcitabine Plus Durvalumab

Margherita Rimini^{1,2} · Eleonora Loi³ · Mario Domenico Rizzato^{4,5} · Tiziana Pressiani⁶ · Caterina Vivaldi^{7,8} · Eleonora Gusmaroli⁹ · Lorenzo Antonuzzo^{10,11} · Erika Martinelli¹² · Ingrid Garajova¹³ · Guido Giordano^{14,15} · Jessica Lucchetti¹⁶ · Marta Schirripa¹⁷ · Noemi Cornara² · Federico Rossari² · Francesco Vitiello² · Elisabeth Amadeo² · Mara Persano³ · Vittoria Matilde Piva^{4,5} · Rita Balsano⁶ · Francesca Salani^{18,19} · Chiara Pircher⁹ · Stefano Cascinu^{1,2} · Monica Niger⁹ · Lorenzo Fornaro⁷ · Lorenza Rimassa^{6,20} · Sara Lonardi⁴ · Mario Scartozzi³ · Patrizia Zavattari³ · Andrea Casadei-Gardini^{1,2}

Abstract

Background The results reported in the TOPAZ-1 phase III trial led to the approval of the combination of cisplatin and gemcitabine with durvalumab as the new first-line standard of care for patients with locally advanced or metastatic cholangiocarcinoma.

Objective We performed a clustering analysis to classify patients into different groups based on their mutation profile, correlating the results of the analysis with clinical outcomes.

Methods We selected 51 patients with cholangiocarcinoma who were treated with the combination of chemotherapy and durvalumab and who were screened using the next-generation sequencing-based FoundationOne gene panel. We conducted mutation-based clustering of tumors and a survival analysis.

Results Three main clusters were identified. Cluster 1 is mostly characterized by mutations in genes belonging to the chromatin modification pathway, altered in 100% of patients. Cluster 2 is characterized by the alteration of several pathways, among which DNA damage control, chromatin modification, RTK/RAS, cell-cycle apoptosis, TP53, and PI3K were the most affected. Finally, most altered pathways in cluster 3 were RTK/RAS and cell-cycle apoptosis. Overall response rate was 4/13 (31%), 12/24 (50%), and 0/10 (0%) in cluster 1, cluster 2, and cluster 3, respectively, and the difference between the three clusters was statistically significant ($p = 0.0188$).

Conclusions By grouping patients into three clusters with distinct molecular and genomic alterations, our analysis showed that patients included in cluster 2 had higher overall response rates, whereas patients included in cluster 3 had no objective response. Further investigations on larger and external cohorts are needed in order to validate our results.

Margherita Rimini, Eleonora Loi are Co-first authors.

Patrizia Zavattari and Andrea Casadei-Gardini are Co-last authors.

1 Introduction

Biliary tract cancers (BTCs) are a heterogeneous group of malignancies with a dismal prognosis and poor therapeutic options [1–4]. For early stage, surgery followed by chemotherapy is the only curative option, but the proportion of recurrences after radical treatment remains high [5]. For locally advanced and metastatic stages, platinum-based chemotherapy had constituted the standard of care since 2010, when the randomized ABC-02 phase III trial reported a survival benefit with the addition of cisplatin to gemcitabine compared with gemcitabine alone. Nevertheless, survival rates reported for the chemotherapy combination were about 2% at 5 years for patients with metastatic disease [6, 7]. Recently, the development of new high-throughput molecular analysis techniques has highlighted the genomic heterogeneity of BTC, and several potential molecular targets with therapeutic implications have been investigated [1, 8–12].

Moreover, immunotherapy has shown promise for these patients: the phase III TOPAZ-1 trial reported a benefit in terms of both overall survival (OS) and progression-free survival (PFS) for patients with locally advanced or metastatic BTC who received the anti-programmed cell death ligand 1 (anti-PD-L1) durvalumab in addition to the standard chemotherapy combination [13]. More precisely, the combination of cisplatin/gemcitabine and durvalumab achieved a median OS of 12.8 months compared with 11.5 months for chemotherapy alone, with a significant reduction in the risk of death of 20% in favor of the experimental arm [13]. These results led to the approval of the combination with durvalumab as the new first-line standard of care by the US Food and Drug Administration and European Medicines Agency. Another chemo-immunotherapy combination has been recently proposed for patients with advanced BTC:

the phase III KEYNOTE-966 trial demonstrated improved survival outcomes with the addition of the anti-programmed cell death 1 pembrolizumab to the standard chemotherapy backbone cisplatin/gemcitabine [15], thus confirming the benefit of chemotherapy combined with immunotherapy in this setting.

Despite the recent progress in this field, no biomarkers able to identify which patients could benefit most from immunotherapy have been identified, and only scarce data about the clinical impact of genomic features in this setting are available. Of note, the TOPAZ-1 trial reported no significant interaction between PD-L1 status and treatment efficacy, as the clinical benefit of the addition of durvalumab to chemotherapy was not significantly different in PD-L1-positive and PD-L1-negative patients. The impact of molecular features on efficacy outcomes in patients enrolled in the same trial has been recently presented: efficacy expressed in terms of OS, overall response rate (ORR), and PFS with the addition of durvalumab to standard chemotherapy has been reported consistently in all patients regardless of the mutational profile, including patients with actionable alterations [14]. A deeper knowledge on the molecular profile of these patients and on the therapeutic implications could lead to the identification of those patients who are more likely to respond to a treatment. Moreover, the identification of those patients who could be more responsive to immunotherapy could open the way to further investigations focused on other clinical settings, including the neoadjuvant and adjuvant settings.

To gain insight into the molecular heterogeneity of BTC and the therapeutic impact of genomic profiling, we performed a clustering analysis to classify patients into different groups based on their mutation profile. In details, we analyzed a cohort of patients with locally advanced or metastatic BTC who received durvalumab plus cisplatin/gemcitabine as first-line treatment in a real-world setting, correlating the results of the analysis with clinical outcomes.

2 Materials and Methods

2.1 Patients' Enrollment and Sample Collection

The overall population included patients with unresectable, locally advanced, or metastatic BTC, including intra-hepatic and extra-hepatic cholangiocarcinoma and gallbladder carcinoma. Data were prospectively collected from 11 Italian institutions. Formalin-fixed paraffin-embedded samples and hematoxylin-eosin staining slides of the patients included were collected from the pathology department of each institution. A genomic analysis of primary tumors was performed by the FOUNDATION Cdx technology (FoundationOne assay).

Patients included in the cohort received gemcitabine (1000 mg/m²) plus cisplatin (25 mg/m²) intravenously on days 1 and 8 of a 21-day cycle for up to eight cycles. Durvalumab (1500 mg) was administered on day 1 of each cycle, in combination with chemotherapy. After completion of up to eight cycles, durvalumab monotherapy was administered once every 4 weeks until clinical or imaging disease progression or unacceptable toxicity. Because durvalumab is not yet reimbursed by the Italian Medicines Agency (AIFA), it was provided free of charge at the request of physicians for each individual patient by AstraZeneca Italy within an early access program. AstraZeneca Italy had no role in planning this study or collecting or analyzing patient data. The present study was approved by the local ethics committee at each center, complied with the provisions of the Good Clinical Practice guidelines and the Declaration of Helsinki and local laws, and fulfilled the Regulation (European Union) 2016/679 of the European Parliament and of the Council of 27 April 2016 on the protection of natural persons with regard to the processing of personal data.

2.2 Clinical Data

Clinical data including patients' age, gender, and Eastern Cooperative Oncology Group Performance Status were retrospectively collected at baseline. Pathological data, including primary tumor location, histological grading, and TNM stage according to the 8th edition of the 2017 American Joint Committee on Cancer staging system were collected at baseline. Response to treatment was assessed using RECIST 1.1 criteria. For each patient, a follow-up and oncology assessment were planned as per standard of practice, according to guidelines and institutional protocols.

2.3 Identification of Genomic Alterations

Formalin-fixed paraffin-embedded tumor tissues containing at least 20% of tumor cells were collected at each center and sent for genomic analysis by the next-generation sequencing (NGS)-based FoundationOne assay (FoundationOne®; Foundation Medicine Inc., MA, USA) gene panel. Identified alterations included insertions/deletions (indel, 1–40 bp), base substitutions, copy number alterations/amplifications (ploidy < 4, amplification with copy number ≥ 8), copy number alterations/deletions (ploidy < 4, homozygous deletions), and fusion/rearrangements in 324 genes. Variants of uncertain significance were included in the analysis. In addition, microsatellite status (determined by assessing indel characteristics at 114 homopolymer repeat loci in or near the targeted gene regions of the FoundationOne test) was assessed. A descriptive analysis of the molecular landscape in the entire sample was performed.

2.4 Clustering Analysis

Genomic data were collected into electronic data files by each participating center and centrally reviewed at the coordinating center (IRCCS San Raffaele Hospital) in order to perform a clustering analysis. For each patient, the mutational status of 324 genes screened in the FoundationOne assay was annotated. Genes showing mutations in more than 5% of patients were selected for the clustering analysis resulting in 37 genes.

Mutation-based clustering analysis was performed with *ccpw* Model from Zhang et al. [16], using a manually curated cancer pathway list (*mcp* list), including 511 putative cancer genes belonging to 16 pathways, as described in our previous work [9]. In brief, binary values indicate the mutational status of cancer driver genes and are used as features in the Ward's hierarchical clustering process. The *mcp* list includes the following pathways: cell cycle and apoptosis (32 genes), chromatin modification (31 genes), DNA damage control (31 genes), HH (6 genes), HIPPO (38 genes), MAPK (8 genes), MYC (13 genes), NOTCH (73 genes), NRF2 (3 genes), PI3K (57 genes), RTK RAS (99 genes), STAT (11 genes), TGF- β (14 genes), TP53 (6 genes), transcriptional regulation (13 genes), and WNT (76 genes). It should be noted that the same genes could belong to multiple pathways.

2.5 Statistical Analysis

Categorical variables were presented as totals and frequencies and evaluated by the Chi-squared test or Fisher exact test, as appropriate. Genomic alterations observed in ≥5% of the entire cohort of patients were considered for the analysis of distribution of genomic alterations. A survival analysis according to the identified clusters was performed. In the survival analysis, PFS and OS were considered. Progression-free survival was measured from the date of the start of first-line therapy to the date of disease progression, death, or last follow-up. Overall survival was measured from the date of the first-line therapy start and the date of death or last follow-up. The last follow-up was not counted as an event, but patients were censored at that point.

Progression-free survival and OS from first-line therapy were calculated by the Kaplan–Meier method and assessed by a log-rank test for a univariate analysis. The results were recorded as hazard ratios (HRs) and 95 confidence intervals (CIs). A two-tailed *p*-value less than 0.05 was considered statistically significant. For the response rate analysis, both ORR and disease control rate were considered. Overall response rate was defined as the proportion of patients who achieved a complete response or partial response. Disease control rate was defined as the proportion of patients who achieved a complete response or partial response or a stable

disease. A MedCalc package (MedCalc® version 16.8.4) was used for the statistical analysis.

3 Results

3.1 Clinical Characteristics

Overall, the cohort of patients included 138 patients with advanced BTC who received treatment from January 2017 to November 2022; 51 of them were studied with a 324-gene NGS panel (FoundationOne), and thus were included in the survival and clustering analysis. Median age at diagnosis was 62 years (range: 36–80 years). Overall, 74.5% of the sample was affected by an intrahepatic cholangiocarcinoma, whereas less than 20% presented with an extrahepatic cholangiocarcinoma and 6% with a gallbladder carcinoma. Thirty-five percent of the patients had received previous surgery, and 21.5% of patients were treated with adjuvant chemotherapy. Most of the cohort (96%) had metastatic disease at baseline, with only two patients with locally advanced disease. Fifty-seven percent of the entire population presented an Eastern Cooperative Oncology Group Performance Status of 0–1, and 55% were classified as normal weight. A significant proportion of patients (65%) had high blood levels of CA 19-9 at baseline, whereas approximately 50% presented with a neutrophil-lymphocyte ratio >3. Baseline characteristics are reported in Table 1.

3.2 Selection of Genomic Alterations

First, we performed a descriptive molecular analysis and selected the genomic alterations observed in at least 5% of patients. Overall, the 324-gene NGS panel allowed the identification of 183 genomic alterations observed in at least 5% of the entire cohort. These genomic alterations involved 37 genes, with a median of 5.0 genomic alterations per gene (range 3–15) and a median of genomic alterations for patients of 3.6. The most common genomic alterations were found in *ARID1A*, (29%) *CDKN2A/2B* (21.5%), *MLL2* (17.5%), *BRCA2* (15.5%), *PBRM1* (15.5%), *NRAS/KRAS* (15.5%), *BAP1* (15.5%), *TP53* (14%), *IDH1* (12%), *MTAP* (12%), *MDM2* (10%), *ATM* (10%), *MSH3* (10%), *SMAD4* (10%), and *PIK3C2B* (10%) (Fig. 1). The entire list of genomic alterations observed in ≥5% of patients is reported in Table 2.

3.3 Clustering Analysis

Overall, 30 out of the 37 genes with a mutation frequency of ≥5% were present in the mcp list and therefore included in the clustering analysis (Table 2). Figure 2A shows two main clusters: one including 27 patients (cluster 2) and the second

including 24 patients, further subdivided into one cluster of 13 patients (cluster 1) and one of 11 patients (cluster 3). The three main clusters are characterized by alterations in genes belonging to different pathways (Fig. 2B–D). Patients included in the three different clusters had similar baseline characteristics, except for baseline blood levels of CA 19-9, as patients in cluster 3 showed fewer patients with elevated CA 10-9 levels compared with those included in cluster 1 and 2 (18% vs 59% vs 62%, respectively, $p = 0.0395$) [Table 1].

By comparing the number of mutated genes in the three clusters, cluster 2 had the highest number of mutated genes (36/37 analyzed genes), while patients belonging to cluster 1 and cluster 3 showed genomic alterations in a similar number of genes (15/37 and 13/37 analyzed genes, respectively). Cluster 1 is mostly characterized by mutations in genes belonging to the chromatin modification pathway, altered in 100% of patients (Fig. 2B). Cluster 2 is characterized by the alteration of several pathways, among which DNA damage control (81%), chromatin modification (78%), RTK/RAS (63%), cell-cycle apoptosis (59%), TP53 (48%), and PI3K (30%) were the most affected (Fig. 2C). Finally, the most altered pathways in cluster 3 were RTK/RAS (36%) and cell-cycle apoptosis (27%) (Fig. 2D).

By considering the single pathways, several differences were found in terms of altered genes and frequencies in cluster 1, cluster 2, and cluster 3. Eight genes belonging to the chromatin modification pathway were found to be altered in our sample (*ARID1A*, *PBRM1*, *MLL2*, *IDH1*, *IDH2*, *SETD2*, *TET2*, and *ASXL1*). This pathway was altered in both cluster 1 and cluster 2 (Fig. S1A of the Electronic Supplementary Material [ESM]). Overall, 7/8 genes belonging to this pathway were mutated in cluster 1 and cluster 2, in particular, *IDH2* was mutated only in cluster 1, while *SETD2* was mutated only in cluster 2. Mutations in *ARID1A* (6/13, 46% of patients), *IDH1* (4/13, 31% of patients), and *IDH2* (3/13, 23% of patients) defined almost the entire cluster 1 (12/13 patients, 92%) and were mutually exclusive except for one patient who presented with mutations in both *ARID1A* and *IDH2*. In contrast, mutations in *ARID1A* (9/27, 33% of patients), *PBRM1* (7/27, 26% of patients), and *MLL2* (7/27, 26% of patients) defined 67% of cluster 2. No genes included in the chromatin modification pathway were altered in cluster 3.

Cluster 2 and cluster 3 shared alterations in the same pathways not altered in cluster 1, i.e., DNA damage control, RTK/RAS, cell cycle and apoptosis, and TP53, but with striking differences. Four genes included in the DNA damage control pathway were found to be altered in our sample (*BRCA2*, *BAP1*, *TP53*, and *ATM*) (Fig. S1B of the ESM). This pathway was altered almost only in cluster 2, showing mutations in 4/4 genes included in this pathway: *BRCA2* (8/27, 30% of patients), *TP53* (7/27, 26% of

Table 1 Patients' characteristics in the whole population, in cluster 1, cluster 2 and cluster 3

Characteristics	Total population <i>N</i> (%) <i>N</i> = 51	Cluster 1 (<i>N</i> = 13)	Cluster 2 (<i>N</i> = 27)	Cluster 3 (<i>N</i> = 11)	<i>P</i> -value
Gender					
Male	27 (53)	7 (54)	16 (59)	4 (36)	0.4382
Female	24 (47)	6 (46)	11 (41)	7 (64)	
Age, years					
≥ 70	17 (33)	4 (31)	8 (30)	5 (45)	0.6274
< 70	34 (67)	9 (69)	19 (70)	6 (55)	
Primary tumor site					
Intrahepatic	38 (74.5)	11 (85)	20 (74)	7 (63)	0.7575
Extrahepatic	10 (19.5)	2 (15)	5 (18.5)	3 (27)	
Gallbladder	3 (6)	0 (0)	2 (7.5)	1 (10)	
Hepatitis					
HBV	3 (6)	1 (8)	2 (7.5)	2 (7.5)	0.8247
HCV	2 (4)	0 (0)	2 (7.5)	2 (7.5)	
Negative	46 (90)	12 (92)	23 (85)	23 (85)	
Previous surgery					
Yes	18 (35)	4 (31)	7 (26)	7 (64)	0.0811
No	33 (65)	9 (69)	20 (74)	4 (36)	
Previous adjuvant therapy					
Yes	11 (21.5)	3 (23)	3 (11)	5 (45)	0.5531
No	8 (15.5)	2 (15)	4 (15)	2 (18)	
Unknown	32 (63)	8 (62)	20 (74)	4 (37)	
Drainage or stent					
Yes	11 (21.5)	4 (31)	4 (5)	3 (27)	0.4516
No	40 (78.5)	9 (69)	23 (95)	8 (73)	
Disease status					
Locally advanced	2 (4)	0 (0)	1 (4)	1 (9)	0.5184
Metastatic	49 (96)	13 (100)	26 (94)	10 (91)	
BMI					
Underweight	2 (4)	0 (0)	0 (0)	1 (9)	0.3938
Normal-weight	28 (55)	8 (62)	16 (59)	5 (45.5)	
Overweight	21 (41)	5 (38)	9 (41)	5 (45.5)	
ECOG PS					
0	29 (57)	5 (38)	17 (63)	7 (64)	0.2997
> 0	22 (43)	8 (62)	10 (37)	4 (36)	
CA 19.9					
< Normal value	17 (33)	5 (38)	10 (37)	9 (82)	0.0395
> Normal value	33 (65)	8 (62)	16 (59)	2 (18)	
Not reported	1 (2)	0 (0)	1 (4)	0 (0)	
CEA					
< Normal value	36 (70.5)	9 (69)	21 (78)	1 (9)	<0.0001
> Normal value	3 (6)	0 (0)	2 (7)	6 (55)	
Not reported	12 (23.5)	4 (31)	4 (15)	4 (36)	
NLR					
< 3	25 (49)	5 (38)	12 (44)	8 (73)	0.1940
> 3	26 (51)	8 (62)	15 (56)	3 (27)	

CEA and CA 19.9 are the two main tumor markers used in this setting of patients, and are normally used in clinical practice with that names

BMI body mass index, *ECOG PS* Eastern Cooperative Oncology Group Performance Status, *HBV* hepatitis B virus, *HCV* hepatitis C virus, *NLR* Neutrophil-lymphocyte ratio

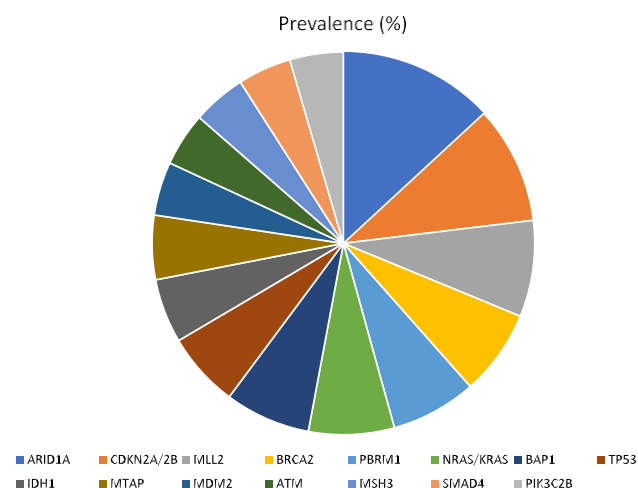


Fig. 1 Diagram of the most frequently altered genes according to the next-generation sequencing test

patients), *BAP1* (6/27, 22% of patients), and *ATM* (5/27, 19% of patients). Only two patients included in cluster 3 were reported to have mutations in *BAP1* (2/11, 18% of patients).

Six genes included in the RAS/RTK pathway were found to be altered in our patient sample (*KRAS*, *NFK1*, *ERBB3*, *FGFR3*, *IGF1R*, and *MAP3K1*) [Fig. S1C of the ESM]. Mutations in 6/6 genes of RAS/RTK pathway were observed in cluster 2: *RAS* (5/27, 19% of patients), *NF1* (4/27, 15% of patients), *FGFR3* (3/27, 11% of patients), *IGF1R* (3/27, 11% of patients), *MAP3K1* (3/27, 11% of patients), and *ERBB3* (2/27, 7% of patients). The only altered genes included in this pathway in cluster 3 were *RAS* (3/11, 27% of patients) and *ERBB3* (1/11, 9% of patients).

TP53, *CDKN2A/2B*, and *CCND1* genes included in the cell-cycle apoptosis pathway were found to be altered in our patient sample (Fig. S1D of the ESM). While *TP53* was mutated in cluster 2 (7/27, 26% of patients), *CDKN2A/B* and *CCND1* were mutated in both clusters with similar frequencies (*CDKN2A/B*: 8/27, 30% of patients in cluster 2 and 3/11, 27% of patients in cluster 3; *CCND1*: 2/27, 7% of patients in cluster 2 and 1/11, 9% of patients in cluster 3).

Finally, three genes included in the TP53 pathway were found to be altered in our patient sample (*TP53*, *MDM2*, and *ATM*). A high percentage of the alterations in this pathway were observed in cluster 2 with mutations in all three genes: *TP53* (7/27, 27% of patients), *ATM* (5/27, 19% of patients), and *MDM2* (3/27, 11% of patients). In contrast, the only mutated gene in cluster 3 was *MDM2* (2/11, 18% of patients). Of note, mutations in *CDKN2A/B* and *RAS* defined 86% (6/7) of mutated patients in cluster 3.

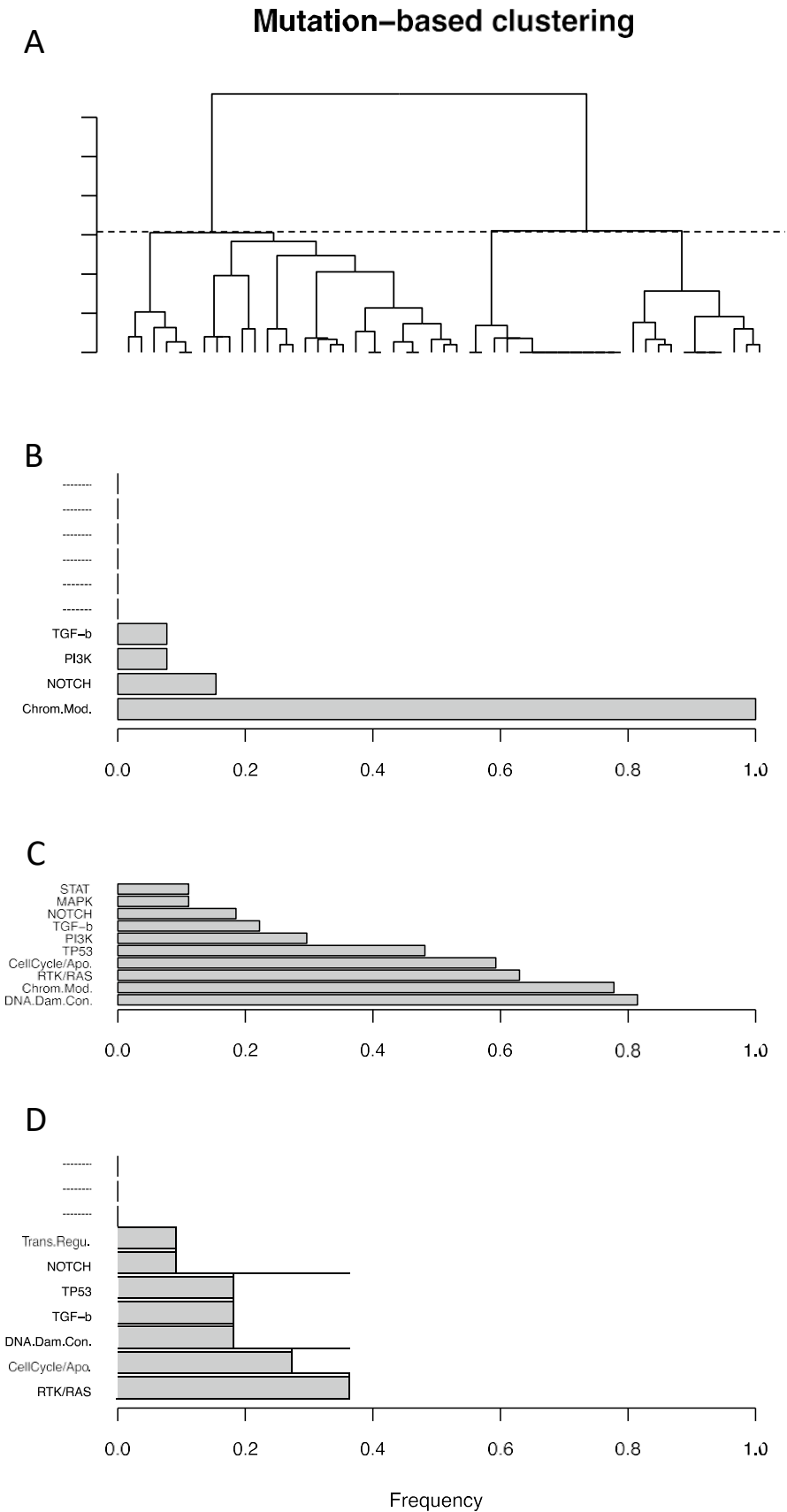
Table 2 List of the main altered genes with the respective pathway and frequency

Altered gene	Pathway(s)	Frequency, N (%)
ARID1A	Chromatin Modification	15 (29)
CDKN2A/2B	Cell cycle_Apoptosis	11 (21.5)
MLL2	Chromatin Modification	9 (17.5)
BRCA2	DNA damage control	8 (15.5)
PBRM1	Chromatin Modification	8 (15.5)
NRAS/KRAS	RTK RAS	8 (15.5)
BAP1	DNA damage control	8 (15.5)
TP53	Cell cycle_Apoptosis; DNA damage control; TP53	7 (14)
IDH1	Chromatin Modification	6 (12)
MTAP	–	6 (12)
MDM2	TP53	5 (10)
ATM	DNA damage control; TP53	5 (10)
MSH3	–	5 (10)
SMAD4	TGF-β	5 (10)
PIK3C2B	–	5 (10)
TGFBR2	TGF-β	4 (8)
NF1	RTK RAS	4 (8)
PIK3CA	PI3K	4 (8)
DIS3	–	4 (8)
NOTCH3	NOTCH	4 (8)
POLE	–	4 (8)
CTNNB1	WNT	3 (6)
SETD2	Chromatin Modification	3 (6)
ERBB3	RTK RAS	3 (6)
CCND1	Cell cycle_Apoptosis	3 (6)
IDH2	Chromatin Modification	3 (6)
SF3B1	Transcriptional Regulation	3 (6)
TET2	Chromatin Modification	3 (6)
FGFR3	PI3K; RTK RAS; STAT	3 (6)
IGF1R	RTK RAS	3 (6)
ATR	–	3 (6)
ASXL1	Chromatin Modification	3 (6)
SPEN	NOTCH	3 (6)
ZNF703	–	3 (6)
NOTCH1	NOTCH	3 (6)
RICTOR	PI3K	3 (6)
MAP3K1	RTK RAS; MAPK	3 (6)

3.4 An Algorithm to Stratify Patients in Clinical Practice According to the Clustering Analysis

We propose an easy-to-use algorithm able to stratify patients in clinical practice according to our clustering analysis. We chose the following genes as nodal points of our algorithm based on the presence of genomic alterations: *TP53*, *BRCA2*, *IDH1/2* or *ARID1A*, *CDKN2A/B* or *RAS* and *SMAD4*, *BAP1*, or *PBRM1* (Fig. 3).

Fig. 2 Classification of patients according to their mutation profile. **A** Mutation-based clustering. **B** Main altered pathways in the three clusters. *Bar plots* indicate the mutation frequency for the most altered pathways in the three clusters. Cluster 1 is mostly characterized by mutations in genes belonging to the chromatin modification pathway, altered in 100% of patients (**B**). Cluster 2 is characterized by the alteration of several pathways, among which DNA damage control (81%), chromatin modification (78%), RTK/RAS (63%), cell-cycle apoptosis (59%), TP53 (48%), and PI3K (30%) were the most affected (**C**). Finally, the most altered pathways in cluster 3 were RTK/RAS (36%) and cell-cycle apoptosis (27%) (**D**)



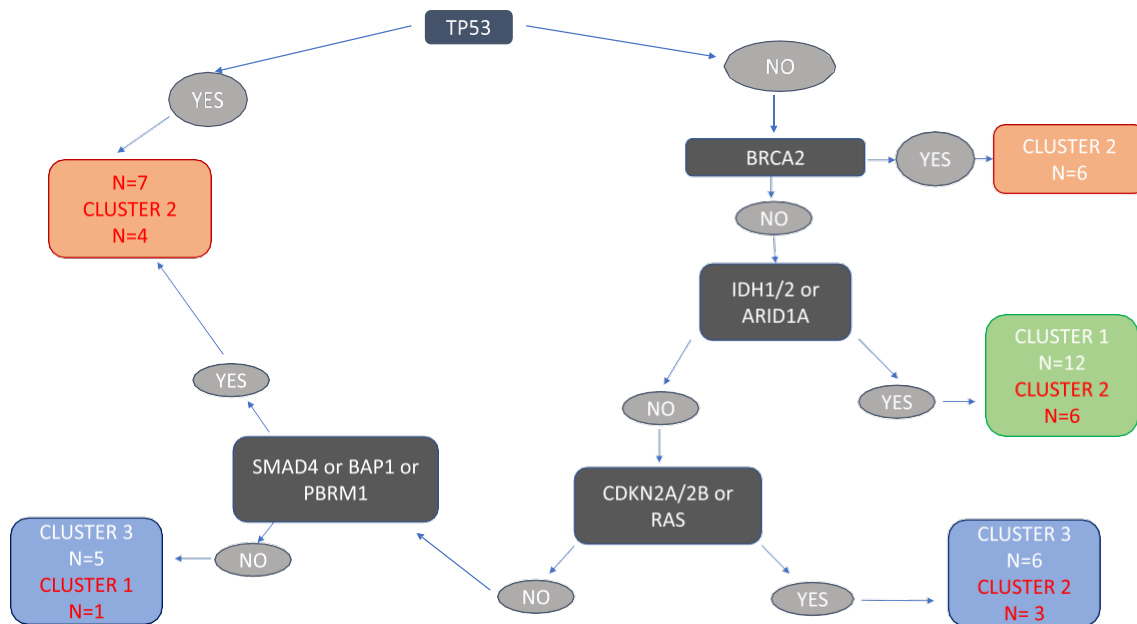


Fig. 3 Schematic of the algorithm designed to stratify patients in the three clusters based on a clustering analysis

3.5 Outcome Analysis According to the Identified Clusters

A survival analysis based on the clustering analysis was performed. In terms of PFS, a trend toward a better outcome was reported for cluster 2, without reaching statistical significance with a median PFS of 8.88 months versus 6.45 and 6.25 months for cluster 1 and cluster 3, respectively (cluster 2 reference HR 1, cluster 1 HR 1.79, cluster 3 HR 1.44, $p = 0.5605$) (Fig. 4A). In terms of OS, no statistically significant differences were found between cluster 1, cluster

2, and cluster 3 (cluster 3 reference HR 1, cluster 2 HR 2.01, cluster 3 HR 2.09, $p = 0.7834$) (Fig. 4B).

Overall, 13, 24, and 10 patients were available for the response rate analysis in cluster 1, cluster 2, and cluster 3, respectively. Overall response rate was 4/13 (31%), 12/24 (50%), and 0/10 (0%) in cluster 1, cluster 2, and cluster 3, respectively, and the difference was statistically significant ($p = 0.0188$). In terms of the disease control rate, 8/13 (61.5%), 12/24 (50%), and 9/10 (90%) patients had disease control in cluster 1, cluster 2, and cluster 3, respectively ($p = 0.0900$). In terms of adverse events, no statistically

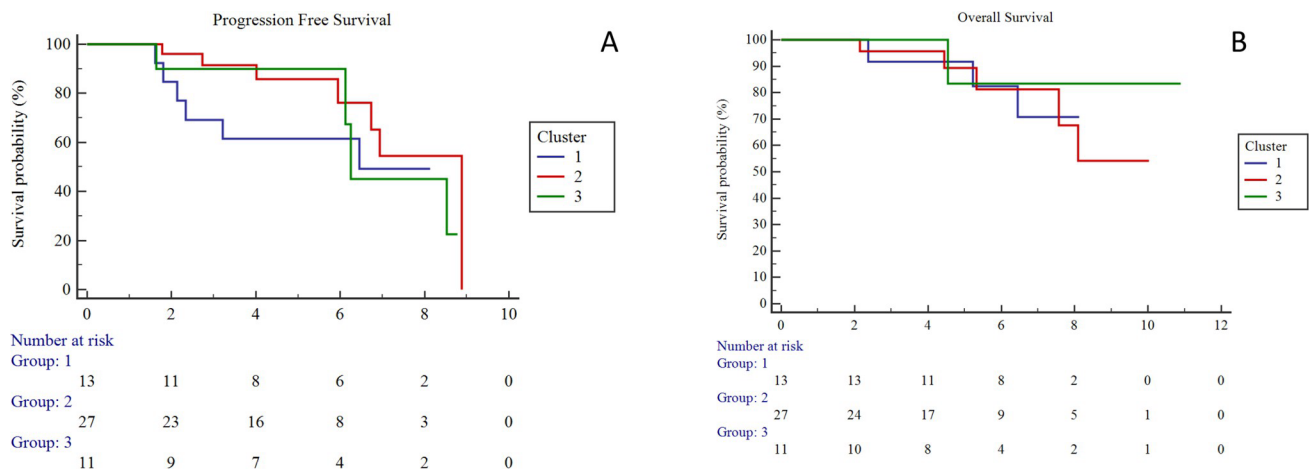


Fig. 4 Kaplan-Meier curves for progression-free survival (A) and overall survival (B) in months according to the three genomic clusters.

significant differences were reported in the three clusters of patients (Table 3).

4 Discussion

To the best of our knowledge, the present work reported the first clustering analysis performed on a sample of patients with advanced BTC treated with durvalumab plus cisplatin/gemcitabine as first-line treatment in a real-world setting. Our analysis highlighted three clusters characterized by different molecular and genomic features. Patients included in the three clusters showed significant differences in terms of response rate, with patients in cluster 2 showing the highest ORR (50%) and no patients included in cluster 3 achieving an objective response. Of note, the negative impact in terms of response rate was observed despite more favorable levels of CA 19-9. Finally, we developed an easy-to-use algorithm in order to transfer our results into clinical practice, thus

providing a useful tool to stratify patients who could benefit from treatment with cisplatin/gemcitabine and durvalumab in terms of ORR.

Several considerations arise from our results. Cluster 1 has been shown to be mainly characterized by genomic alterations in genes involved in the chromatin modification pathway, thus including *IDH1* and *ARID1A*. In contrast, cluster 2 has been highlighted as the most mutated, probably due to frequent mutations in genes involved in DNA damage repair (primarily, *BRCA2* and *ATM*), thus conferring a sort of genomic instability on this group of patients. Finally, cluster 3 was shown to be characterized by genomic alterations in genes involved in RAS/RTK, the cell cycle, and apoptosis. In the survival analysis, no statistically significant differences in OS or PFS were found between the three clusters, possibly because of the small sample size and the low rate of events. However, despite the small sample size, this result is of particular interest in the selection of patient who could benefit

Table 3 Safety Profile in the whole population, in cluster 1, cluster 2 and cluster 3

Toxicities	Total population <i>N</i> (%) <i>N</i> = 51	Cluster 1 (<i>N</i> = 13)	Cluster 2 (<i>N</i> = 27)	Cluster 3 (<i>N</i> = 11)	<i>P</i> -value
Neutrophils decreased					
Yes	3 (6)	1 (8)	2 (7.5)	0 (0)	0.6447
No	48 (94)	12 (92)	25 (92.5)	11 (100)	
Pruritus					
Yes	3 (6)	0 (0)	3 (11)	0 (0)	0.2425
No	48 (94)	13 (100)	24 (89)	11 (100)	
Colitis					
Yes	1 (2)	0 (0)	1 (4)	0 (0)	0.6355
No	50 (98)	13 (100)	26 (96)	11 (100)	
Hypothyroidism					
Yes	2 (4)	0 (0)	2 (7.5)	0 (0)	0.3965
No	49 (96)	13 (100)	25 (92.5)	11 (100)	
Decreased appetite					
Yes	5 (10)	1 (8)	2 (7.5)	2 (18)	0.5729
No	46 (90)	12 (92)	25 (92.5)	9 (82)	
Fever					
Yes	9 (17.5)	4 (31)	2 (7.5)	3 (27)	0.1231
No	42 (82.5)	9 (69)	25 (92.5)	8 (73)	
Fatigue					
Yes	27 (53)	8 (61.5)	14 (52)	5 (45.5)	0.7240
No	24 (47)	5 (38.5)	13 (48)	6 (54.5)	
Immune-related toxicities					
Yes	7 (14)	1 (8)	5 (18.5)	1 (9)	0.5704
No	44 (86)	12 (92)	22 (81.5)	10 (91)	
Other toxicities					
Yes	3 (6)	1 (8)	1 (4)	1 (9)	0.7738
No	48 (94)	12 (92)	26 (96)	10 (91)	

from neoadjuvant and/or conversion treatment. To date, no standard neoadjuvant treatment has been approved for BTC, data are scarce, and the selection of patients who could benefit from neoadjuvant or conversion treatment is still challenging. Indeed, in other cancer types, systemic treatment has been shown to reduce the risk of recurrence after surgery and make patients who were not resectable at baseline resectable. From the TOPAZ-1 trial, patients with locally advanced disease were found to have a larger survival benefit with cisplatin/gemcitabine plus durvalumab compared with patients with metastatic disease. Furthermore, the chemotherapy triplet cisplatin/gemcitabine plus nab-paclitaxel was shown to confer a survival benefit in patients with locally advanced disease and not in patients with metastatic disease [19]. We found distinct response rates by molecular cluster in the advanced setting may lead to some speculations on the impact of cisplatin/gemcitabine plus durvalumab also in a setting of conversion/neoadjuvant treatment, where shrinkage is even more needed to reach better outcomes. Our results might suggest that patients included in cluster 3 might not benefit from conversion/neoadjuvant treatment before surgery. In contrast, patients with locally advanced disease included in cluster 2 could be the best candidates for conversion treatment with cisplatin/gemcitabine plus durvalumab. More investigations are needed in order to verify our hypothesis. The improved response rate reported in cluster 2 could be ascribed to the molecular profile. Indeed, cluster 2 is enriched with mutations in genes like *BRCA2* and *ATM*, which are known to confer the “BRCAness” phenotype. As already demonstrated in several oncological settings, including breast, ovarian, and pancreatic cancer, the “BRCAness” phenotype confers a special susceptibility to platinum compounds and PARP inhibitors [11, 20–24]. In a previous retrospective analysis, our group demonstrated a significant benefit in terms of PFS and a trend toward better OS in a cohort of patients with advanced BTC with the “BRCAness” phenotype treated with cisplatin plus gemcitabine as first-line therapy compared with “BRCAness wild-type” patients [25]. In addition, a growing body of evidence suggests that changes in the DNA damage repair system could alter genomic stability [26], as they lead to the accumulation of DNA damage thus promoting local antigen release and enhancing immunogenicity in tumors [27–29]. The immunogenicity promoted by alterations in DNA damage repair systems may underlie the improved response to immune checkpoint inhibitors, which has been reported in various oncology settings [30]. Thus, “BRCAness” patients seem to be biologically prone to respond well not only to cisplatin and PARP inhibitors, but also to immunotherapy, which is consistent with our findings. Interestingly, a molecular analysis performed on

patients included in the TOPAZ-1 study showed a benefit on the ORR regardless of any genomic alteration.

From a molecular point of view, our work has shown that the most commonly altered genes are *ARID1A*, (29%) *CDKN2A/2B* (21.5%), *MLL2* (17.5%), *BRCA2* (15.5%), *PBRM1* (15.5%), *NRAS/KRAS* (15.5%), *BAP1* (15.5%), *TP53* (14%), *IDH1* (12%), *MTAP* (12%), *MDM2* (10%), *ATM* (10%), *MSH3* (10%), *SMAD4* (10%), and *PIK3C2B* (10%). Recently, Valle and colleagues presented the results of the genomic analysis from the TOPAZ-1 trial: this analysis showed that *TP53* (48.8%), *CDKN2A/B* (25.2%), *KRAS* (24.0%), *ARID1A* (20.9%), *SMAD4* (14.5%), *IDH1* (8.8%), and *PIK3CA* (8.2%) were the most frequently altered genes, with differences depending on the primary tumor site and geographic location. Moreover, they presented the impact of clinically actionable genomic alterations such as in *KRAS*, *IDH1*, *ERBB2*, *BRCA1/2*, *BRAF*, and *FGFR2* on the objective response rate. Interestingly, in the cisplatin/gemcitabine plus durvalumab arm, patients with *BRCA1/2* mutations reported a higher ORR compared with wild-type patients, which is consistent with our results.

Our research has several limitations. First, this is a retrospective analysis, thus selection biases cannot be excluded owing to the nature of the investigation. Second, the small sample size and short follow-up period do not allow the drawing of definitive conclusions, thus validation on a larger cohort of patients is needed. Nevertheless, durvalumab has been recently introduced in the therapeutic armamentarium for BTC, and our cohort is the largest cohort of patients with advanced BTC who received durvalumab in a real-world setting and who were screened with a comprehensive NGS panel at baseline, thus making our results, although preliminary, of interest. Another limitation is related to the NGS panel used: no whole exome sequencing analysis was performed in our cohort, but all patients were tested with the FoundationOne assay, to ensure consistent data quality and, consequently, more reliable results. In addition, the choice to include the variants of uncertain significance has to be considered and justified. Because scarce data are already available concerning the role of molecular alteration and clinical outcomes to cisplatin/gemcitabine plus durvalumab in this setting of patients, as well as the role of many genomic alterations in advanced BTC not yet conclusively defined, we decided to include all the genomic alterations highlighted, thus including the variants of uncertain significance. If it could be considered a methodological choice, we need to consider that in the interpretation of the present results. Moreover, no complete data about the variant allele frequency were available, so this important information has not been considered in the present analysis, thus configuring an important limitation. Finally, our data resulting from a sophisticated clustering analysis need to be confirmed and validated on larger cohorts of patients.

5 Conclusions

The present analysis is one of the first comprehensive genomic analyses performed on a cohort of patients with advanced BTC who received cisplatin/gemcitabine plus durvalumab. The clustering analysis highlighted the presence of three clusters characterized by different genomic profiles with interesting clinical impacts, mainly in terms of response rates. No prognostic value has been shown, and further investigations are needed. The present analysis aimed to identify molecular and genomic biomarkers to select which patients are more likely to respond to the combination of immunotherapy plus chemotherapy. By grouping patients into three clusters with distinct molecular and genomic alterations, despite the limitation of the small sample size, our analysis showed that patients included in cluster 2 had a higher ORR, whereas patients included in cluster 3 had no objective response. Further investigations with larger and external cohorts are needed in order to validate our results. Nevertheless, this preliminary insight into the molecular heterogeneity of patients with BTC treated with cisplatin/gemcitabine plus durvalumab as first-line therapy could open new avenues of research focusing on mechanisms of primary resistance to treatments, with the ultimate goal of improving the stratification of patients and clinical outcomes.

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Declarations

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Ethics Approval The ethics review board of each institutional hospital approved the present study. This study was performed in line with the principles of the Declaration of Helsinki.

Consent to Participate Written informed consent for treatment was obtained for all patients.

Consent for Publication Not applicable.

Availability of Data and Material Data are available on request from the authors.

Code Availability Not applicable.

Authors' Contributions Conception and design: EL, MR, PZ, AC-G; acquisition of data (acquired and managed patients): all authors. analysis and interpretation of data: EL, MR, PZi, AC-G; writing, review, and/or revision of the manuscript: EL, MR, PZ, AC-G; final approval of manuscript: all authors.

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Authors and Affiliations

Margherita Rimini^{1,2}  · Eleonora Loi³ · Mario Domenico Rizzato^{4,5} · Tiziana Pressiani⁶ · Caterina Vivaldi^{7,8} · Eleonora Gusmaroli⁹ · Lorenzo Antonuzzo^{10,11} · Erika Martinelli¹² · Ingrid Garajova¹³ · Guido Giordano^{14,15} · Jessica Lucchetti¹⁶ · Marta Schirripa¹⁷ · Noemi Cornara² · Federico Rossari² · Francesco Vitiello² · Elisabeth Amadeo² · Mara Persano³ · Vittoria Matilde Piva^{4,5} · Rita Balsano⁶ · Francesca Salani^{18,19} · Chiara Pircher⁹ · Stefano Cascinu^{1,2} · Monica Niger⁹ · Lorenzo Fornaro⁷ · Lorenza Rimassa^{6,20} · Sara Lonardi⁴ · Mario Scartozzi³ · Patrizia Zavattari³ · Andrea Casadei-Gardini^{1,2}

✉ Margherita Rimini
margherita.rimini@gmail.com

¹ IRCCS San Raffaele Scientific Institute Hospital, Vita-Salute San Raffaele University, Milan, Italy

² Department of Oncology, IRCCS San Raffaele Hospital, Via Olgettina n. 60, Milan, Italy

³ Department of Biomedical Sciences, Unit of Biology and Genetics, University of Cagliari, Cagliari, Italy

⁴ Department of Oncology, Veneto Institute of Oncology IOV, IRCCS, Padua, Italy

⁵ Department of Surgery, Oncology and Gastroenterology, University of Padua, Padua, Italy

⁶ Medical Oncology and Hematology Unit, Humanitas Cancer Center, IRCCS Humanitas Research Hospital, Rozzano, Milan, Italy

⁷ Unit of Medical Oncology 2, Azienda Ospedaliero-Universitaria Pisana, Pisa, Italy

⁸ Department of Translational Research and New Technologies in Medicine and Surgery, University of Pisa, Pisa, Italy

⁹ Medical Oncology Department, Fondazione IRCCS Istituto Nazionale dei Tumori di Milano, Milan, Italy

¹⁰ Clinical Oncology Unit, Careggi University Hospital, Florence, Italy

¹¹ Department of Experimental and Clinical Medicine, University of Florence, Florence, Italy

¹² Division of Medical Oncology, Department of Precision Medicine, University of Campania Luigi Vanvitelli, Naples, Italy

¹³ Medical Oncology Unit, University Hospital of Parma, Parma, Italy

¹⁴ Unit of Medical Oncology and Biomolecular Therapy, Policlinico Riuniti, Foggia, Italy

¹⁵ Department of Medical and Surgical Sciences, University of Foggia, Foggia, Italy

¹⁶ Division of Medical Oncology, Fondazione Policlinico Universitario Campus Bio-Medico, Rome, Italy

¹⁷ Medical Oncology Unit, Department of Oncology and Hematology, Central Hospital of Belcolle, Strada Sarmartinese Snc, Viterbo, Italy

¹⁸ Unit of Medical Oncology 2, Azienda Ospedaliero-Universitaria Pisana, Pisa, Italy

¹⁹ Institute of Interdisciplinary Research “Health Science”, Scuola Superiore Sant’Anna, Pisa, Italy

²⁰ Department of Biomedical Sciences, Humanitas University, Milan, Italy