# Lack of clinical utility of serum macrophage migration inhibitory factor (MIF) for monitoring therapy response and estimating prognosis in advanced lung cancer

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Received 19 February 2023 Accepted 7 June 2023

#### Abstract.

**BACKGROUND:** Lung cancer is a major burden to global health and is still among the most frequent and most lethal malignant diseases. Macrophage migration inhibitory factor (MIF) is a proinflammatory cytokine involved in a variety of processes including tumorigenesis, formation of a tumor microenvironment and metastasis. It is therefore a potential prognostic biomarker in malignant diseases.

**OBJECTIVE:** In this study, we investigated the applicability of MIF in serum samples as a biomarker in lung cancer.

**METHODS:** In a retrospective approach, we analyzed the sera of 79 patients with non-small-cell lung cancer (NSCLC) and 14 patients with small-cell lung cancer (SCLC) before the start of chemotherapy, as well as before the second and third chemotherapy cycle, respectively. Serum MIF levels were measured using a sandwich immunoassay with a sulfo-tag-labelled detection antibody, while pro-gastrin releasing peptide (proGRP) levels were determined with an enzyme-linked immunosorbent assay.

**RESULTS:** No difference in serum MIF levels between responders and non-responders to chemotherapy was observed at all time points, while proGRP levels were significantly lower in responders before the second chemotherapy cycle (p = 0.012). No differences in biomarker levels depending on the histopathological classification of NSCLC patients was found. Moreover, in ROC curve analyses MIF was not able to distinguish between responders and non-responders to therapy. proGRP could differentiate between responders and non-responders before the second chemotherapy cycle (p = 0.015) with sensitivities of 43% at 90% and 95% specificity, respectively. Likewise, proGRP yielded significantly longer survival times of patients with low proGRP concentrations before the second chemotherapy cycle (p = 0.015) in Kaplan-Meier analyses, yet MIF showed no significant differences in survival times at all time points. Comparison with the biomarkers CEA and CYFRA 21-1 in the same cohort showed that these established biomarkers clearly performed superior to MIF and proGRP.

**CONCLUSIONS:** From the present results, there is no indication that serum MIF may serve as a biomarker in prognosis and monitoring of response to therapy in lung cancer. Limitations of this study include its retrospective design, the inclusion of a larger NSCLC and a smaller SCLC subgroup, the classical chemotherapeutic treatment, the use of a non-diagnostic immunoassay (RUO-test) for MIF measurement and the lack of a validation cohort. Strengths of the study are its highly standardized procedures concerning sample collection, preanalytic treatment, measurements and quality control of the laboratory assays.

Keywords: Biomarkers, MIF, lung cancer, proGRP

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# 1. Introduction

Lung cancer is a major health burden worldwide and is being held accountable for 18% of all cancer deaths [1]. The major cause of lung cancer is tobacco smoking [2]. Although new therapeutic possibilities such as immunotherapy have considerably improved the treatment of lung cancer, overall survival remains relatively low, especially in advanced stages [3–5].

Blood-based biomarkers can play an important role in the diagnosis, prognosis and monitoring of therapy response of malignant diseases, as they possess several advantages compared to other diagnostic tools. They are readily available in most laboratory medicine environments without the need for additional specialized equipment. On the other hand, blood-based biomarkers are easily obtained on a virtually non-invasive basis, in contrast to procedures such as tissue biopsies or lumbar punctures.

Due to the great impact of lung cancer on public health, a multitude of biomarkers has been studied. Some of the most widely studied biomarkers in lung cancer are cytokeratin fragment 19 (CYFRA 21-1), carcinoembryonic antigen (CEA) and pro-gastrin releasing peptide (proGRP) [6–8]. In studies with lung cancer patients, these markers showed promising sensitivities and specificities, but did not reach a level of reliability sufficient for routine application in a clinical setting. Moreover, when a potential use as a screening tool is considered, positive and negative predictive values are low due to the overall low prevalence of lung cancer, which is estimated to be in the range of 1–4% in high risk groups [9]. Although combinations of biomarkers can further increase sensitivities and specificities [8, 10, 11], the search for new biomarkers and their clinical evaluation remain crucial factors to facilitate early diagnosis of lung cancer.

The macrophage migration inhibitory factor (MIF) is a proinflammatory cytokine with an additional tautomerase activity which is expressed by a variety of cells including T cells, monocytes, macrophages, dendritic cells, B cells, granulocytes, epi- and endothelial cells, fibroblasts and cells of the anterior pituitary gland [12, 13]. Its active physiological form is a homotrimer with a molecular weight of 37.5 kDa [14, 15]. MIF is highly pleiotropic and hence exerts a multitude of effects on the immune system and the general inflammatory response. It can counteract the immunosuppressive effects of glucocorticoids [16], is an activator of macrophages and T cells [12], stimulates the release of proinflammatory molecules such as TNF, IFN-y, interleukins, PGE<sub>2</sub> or COX2 [17-19], and has been linked to the pathogenesis of sepsis, ARDS, atherosclerosis, thrombosis, ACS, depression and rheumatologic diseases, for instance [12, 20-26]. Besides these pivotal functions, MIF is involved in tumorigenesis, tumor growth, angiogenesis and metastasis in cancer [13]. It is able to suppress the tumor suppressor p53, leading to a perturbance of the normal cell cycle and a reduced ability to enter apoptosis [17]. Additionally, MIF upregulates HIF-1 $\alpha$  and promotes the Warburg effect in lung cancer cell lines, induces VEGF secretion, reduces E-cadherin expression and activates both PI3K/Akt and ERK1/2 signaling pathways, which are known contributors to tumor formation [18, 27]. Furthermore, MIF supports the formation of a tumor microenvironment by M2 polarization of tumor-associated macrophages [28, 29]. In NSCLCs, MIF was found to exert at least part of its effects such as secretion of angiogenic mediators through CD74 receptors [30]. Inhibiting the tautomerase activity of MIF led to growth attenuation of lung cancer both *in vitro* and in a mouse model [31]. Kamimura et al. showed that the intracellular distribution of MIF predicted prognosis of patients with lung adenocarcinomas [32]. In another study with 96 patients with squamous cell carcinomas of the lung, high MIF levels were associated with lymph node metastases, shorter disease-free survival and shorter diseasespecific survival [33]. Altogether, MIF may not only provide additional insight into the linkage between inflammation and tumor growth, but also has the potential to function as a novel biomarker in lung cancer.

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In this study, we sought to evaluate the potential of MIF serum levels as a predictive factor of therapy response and a prognostic factor for survival. Moreover, the results are compared to established biomarkers of lung cancer so as to better assess the potential role of MIF.

# 2. Patients, materials and methods

The cohort reported in this study and the applied chemotherapeutic agents have already been described elsewhere [34], and is therefore covered only briefly here. In a retrospective approach, serum samples of 93 patients suffering from newly diagnosed or recurring lung cancer were investigated, comprising 79 patients with NSCLC (53 male, 26 female) and 14 SCLC (8 male, 6 female). NSCLC patients were classified as UICC stage 3 (N=21, 27%) or 4 (N=56, 71%); in two cases (2%), no staging information was available. SCLC cases comprised six patients (43%) with UICC stage 3 and eight patients (57%) with UICC stage 4. Samples were collected in a clinical routine setting in the Lungenfachklinik Gauting (Germany) between January 1999 and June 2002 prior to the start of chemotherapy as well as prior to cycles 2 and 3. Predominant histopathology in NSCLCs was adenocarcinoma (N=30, 38%) and squamous cell carcinoma (N=25, 33%), large cell carcinoma (N=6, 8%) and unspecified (N=17, 21%). NSCLC and SCLC groups did not differ significantly in terms of age (U test, p=0.776) or sex ( $\chi^2$  test, p=0.546) [34].

Response to chemotherapy was determined by CT-based staging after cycle 2 according to the WHO classification, thereby defining *partial remission* as a reduction in tumor mass of at least 50%, *progression* as an increase in tumor mass of at least 25%, and *no change* if neither condition was met [35, 36]. Therapy response in patients classified as *no change* was reevaluated after cycle 4 by another CT-based staging. Patients classified as *partial remission* or *no change* were combined into the group *response to therapy*, while patients classified as *progression* were defined as *no response to therapy*.

MIF concentrations were determined using the Human MIF Assay (Meso Scale Diagnostics, Rockville, MD, USA) on a Meso Quickplex SQ120 imager (Meso Scale Diagnostics, Rockville, MD, USA). This method is based on a sandwich immunoassay with a sulfo-tag-labelled detection antibody. Detection is conducted by electroluminescence using tripropylamine and [Ru(bipyridyl)<sub>3</sub>]<sup>2+</sup> as catalyst. Regeneration of the Ru(II) complex is controlled by an anodic current. Serum levels of CEA and CYFRA 21-1 were measured by electrochemiluminescence on the Cobas Elecsys E 411 platform (Roche Diagnostics, Mannheim, Germany), ProGRP by enzyme-linked immunosorbent assay (ELISA, IBL, Hamburg, Germany).

Statistical analyses were performed with IBM SPSS Statistics (version 23, International Business Machines Corporation, Armonk, NY, USA). Medians, interquartile ranges (IQRs) and ranges are used in tables and diagrams as measures of central tendency and measures of variation, respectively. Differences in MIF and proGRP between groups were calculated with Mann-Whitney-U tests. Correlations between biomarkers were assessed with Spearman's rank-order correlation, whereas correlations of biomarkers with therapy response were assessed with point-biserial correlations. To evaluate the diagnostic power of MIF and proGRP, receiver operating characteristic (ROC) curves with tumor progression as state variable were computed. Areas under the curves (AUCs) as well as sensitivities at specificities of 90% (Sens<sub>90</sub>) and 95% (Sens<sub>95</sub>), were calculated from ROC curves. Prognostic value of MIF and proGRP was investigated by univariate log-rank tests and visualized with Kaplan-Meier curves. The significance level was set to p < 0.05 for all tests. Considering the explorative approach of this study, no correction for multiple testing was performed. Results of the analysis are compared to established tumor markers CEA and CYFRA 21-1, which have also been measured in the study cohort. The results from those measurements have already been reported elsewhere and are therefore not analyzed in detail here [34].

# 3. Results

## 3.1. Serum levels of MIF and proGRP

Pretherapeutic concentrations of MIF and proGRP in responders and non-responders with NSCLCs are given in Table 1, while the respective concentrations in SCLCs are shown in Table 2. MIF levels did not differ significantly between responders and non-responders at all time points both in NSCLCs and SCLCs, as determined by Mann-Whitney-U tests. Concentrations of proGRP only differed between responders and non-responders at the second chemotherapy cycle, with responders possessing a lower concentration than non-responders.

Spearman correlation of pretherapeutic levels of MIF with pretherapeutic levels of proGRP in NSCLC cases showed a positive correlation coefficient ( $\rho$ =0.195), but failed to reach the defined significance level (p=0.171). Results from point-biserial correlation between baseline concentrations and therapy response yielded no significant correlations as well (MIF:  $\rho$ =-0.058, p=0.613; proGRP:  $\rho$ =0.055, p=0.703). Respective correlations in the SCLC subgroup revealed a correlation coefficient between pretherapeutic levels of MIF and proGRP of 0.532 (p=0.075) as well as point-biserial corre-

Pretherapeutic         R         41 $5.91$ $7.25$ $2.27 - 92.0$ $0.6$ MIF         R         38 $6.12$ $5.93$ $1.48 - 85.2$ proGRP         R         31 $16.0$ $12.0$ $3.0 - 48.0$ $0.6$ NR         20 $17.0$ $15.3$ $4.0 - 59.0$ $0.6$ Cycle 2           NR $20$ $17.0$ $15.3$ $4.0 - 59.0$ Cycle 2           NR $20$ $17.0$ $15.3$ $4.0 - 59.0$ Cycle 2           NR $20$ $17.0$ $15.3$ $4.0 - 59.0$ proGRP         R         16 $10.5$ $13.8$ $3.0 - 24.0$ $0.6$ proGRP         R         16 $10.5$ $13.8$ $3.0 - 24.0$ $0.6$ MIF         R         40 $7.09$ $6.44$ $2.61 - 36.3$ $0.7$ proGRP         R         20 $12.5$ $11.8$ $3.0 - 193$ $0.4$	NSCLC	Response	Ν	Median	IQR	Range	<i>p</i> -value
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	Pretherapeutic						
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	MIF	R	41	5.91	7.25	2.27 - 92.0	0.610
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		NR	38	6.12	5.93	1.48 - 85.2	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	proGRP	R	31	16.0	12.0	3.0 - 48.0	0.699
$\begin{array}{c c c c c c c c c c c c c c c c c c c $		NR	20	17.0	15.3	4.0 - 59.0	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	Cycle 2						
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	MIF	R	40	4.94	8.56	1.77 – 51.6	0.638
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		NR	38	6.62	7.37	1.42 - 39.0	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	proGRP	R	16	10.5	13.8	3.0-24.0	0.012
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		NR	7	22.0	32.0	9.0-174	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	Cycle 3						
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	MIF	R	40	7.09	6.44	2.61-36.3	0.753
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		NR	34	7.20	6.71	1.7-51.51	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	proGRP	R	20	12.5	11.8	3.0-193	0.403
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		NR	11	18.0	26.0	3.0-198	
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	Changes Cycles 1-2 (%)						
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	MIF	R	41	82.3	118	0.0 - 816	0.259
proGRP         R         12         113.6         203.1         8.6 - 700         0.4           NR         3         135.3         -         44.4 - 916         0.4           Changes Cycles 1-3 (%)         NR         3         135.3         -         44.4 - 916           MIF         R         41         94.4         182         0.0 - 977         0.9           NR         38         102         154         0.0 - 750         0.9           proGRP         R         31         0.00         93.8         0.0 - 1206         0.5		NR	38	96.0	108	24.0 - 884	
NR         3         135.3         -         44.4 - 916           Changes Cycles 1-3 (%)         R         41         94.4         182         0.0 - 977         0.9           MIF         R         41         94.4         182         0.0 - 977         0.9           NR         38         102         154         0.0 - 750         0.9           proGRP         R         31         0.00         93.8         0.0 - 1206         0.5           NR         20         0.00         46.3         0.0 - 1042         0.5	proGRP	R	12	113.6	203.1	8.6 - 700	0.448
Changes Cycles 1–3 (%)       R       41       94.4       182 $0.0-977$ $0.5$ MIF       R       38       102       154 $0.0-750$ proGRP       R       31 $0.00$ 93.8 $0.0-1206$ $0.5$ NR       20 $0.00$ 46.3 $0.0-1042$ $0.5$		NR	3	135.3	_	44.4 - 916	
MIF         R         41         94.4         182         0.0 - 977         0.9           NR         38         102         154         0.0 - 750         0.9           proGRP         R         31         0.00         93.8         0.0 - 1206         0.5           NR         20         0.00         46.3         0.0 - 1042         0.5	Changes Cycles 1-3 (%)						
NR         38         102         154         0.0 - 750           proGRP         R         31         0.00         93.8         0.0 - 1206         0.5           NR         20         0.00         46.3         0.0 - 1042         0.5	MIF	R	41	94.4	182	0.0 - 977	0.969
proGRP R 31 0.00 93.8 0.0 – 1206 0.5 NR 20 0.00 46.3 0.0 – 1042		NR	38	102	154	0.0 - 750	
NR 20 0.00 46.3 $0.0 - 1042$	proGRP	R	31	0.00	93.8	0.0 - 1206	0.569
100  20  0.00  10.5  0.0 - 1042		NR	20	0.00	46.3	0.0 - 1042	

Table 1 Serum concentrations of MIF and proGRP in responders and non-responders with NSCLC

R: responders, NR: non-responders, IQR: interquartile range. Concentrations are given in ng/mL (MIF) and pg/mL (proGRP), respectively.

SCLC	Response	Ν	Median	IQR	Range	<i>p</i> -value
Pretherapeutic						
MIF	R	10	3.92	4.83	1.1 - 10.13	0.635
	NR	4	5.37	43.08	2.2 - 57.84	
proGRP	R	8	33.0	241	0.0 - 1261	0.461
	NR	4	341	740	0.0 - 923	
Cycle 2						
MIF	R	10	3.33	5.25	0.68 - 16.6	0.733
	NR	4	4.66	7.58	1.77 - 10.5	
proGRP	R	8	97.5	164	7.0 - 345	0.497
	NR	3	364	_	0.0 - 1355	
Cycle 3						
MIF	R	10	3.34	5.01	0.95 - 34.5	0.240
	NR	4	5.63	30.92	3.1 - 43.02	
proGRP	R	9	16.0	33.5	3.0 - 83.0	0.260
	NR	4	430	2423	0.0 - 2977	
Changes Cycles 1–2 (%)						
MIF	R	10	101	117	33.3 - 433	0.733
	NR	4	81.8	292	4.5 - 393	
proGRP	R	6	61.0	96.6	12.1 - 150	0.429
	NR	2	115	_	83.3 - 146.8	
Changes Cycles 1–3 (%)						
MIF	R	10	122	73.2	15.1 – 485	0.839
	NR	4	116	1476	5.4 - 1957	
ProGRP	R	7	13.1	31.5	0.0 - 200	0.117
	NR	3	185	_	20.4 - 323	

 Table 2

 Serum concentrations of MIF and proGRP in responders and non-responders with SCLC

R: responders, NR: non-responders, IQR: interquartile range. Concentrations are given in ng/mL (MIF) and pg/mL (proGRP), respectively.

lation coefficients of 0.157 (p = 0.592) and 0.231 (p = 0.470) for correlations of therapy response with MIF and proGRP, respectively.

## 3.2. Changes in MIF and proGRP during chemotherapy

As given in Tables 1 and 2, changes in biomarker concentrations from the start of chemotherapy to the second and third chemotherapeutic cycle did not differ significantly between responders and non-responders for both biomarkers and in both subgroups. In SCLC, median levels of proGRP were considerably higher in non-responding patients and kinetics differed between both response groups, too. However, the numbers of patients investigated were quite low preventing an acceptable level of significance.

### 3.3. Influence of histopathology in NSCLC

Pretherapeutic serum concentrations of MIF and proGRP in the most prominent histopathological entities, i.e. adenocarcinomas and squamous cell carcinomas, were compared to the remaining NSCLC cases, respectively. According to the results shown in Table 3, no significant differences were found.

Table 3

Pretherapeutic serum concentrations of MIF and proGRP stratified for histopathological entities in NSCLC cases						
Histopathology	MIF (ng/mL)	proGRP (pg/mL)				
Adenocarcinomas	6.12	11.5				
Non-adenocarcinomas	5.76	17.0				
<i>p</i> -value	0.455	0.158				
Squamous cell carcinomas	6.46	17.0				
Non-squamous cell carcinomas	5.84	16.0				
<i>p</i> -value	0.602	0.879				

Table 4 Results from ROC curve analyses in NSCLC patients with tumor progression as state variable

	AUC	95%-CI	Sens <sub>90</sub> (%)	Sens <sub>95</sub> (%)	<i>p</i> -value
Pretherapeutic					
MIF	0.467	0.338 - 0.595	10.5	2.6	0.610
proGRP	0.532	0.369 - 0.696	5.0	5.0	0.700
Cycle 2					
MIF	0.531	0.401 - 0.661	7.9	0.0	0.638
proGRP	0.826	0.645 - 1.000	42.9	42.9	0.015
Cycle 3					
MIF	0.521	0.387 - 0.655	17.6	8.8	0.753
proGRP	0.595	0.363 - 0.828	18.2	9.1	0.386

AUC: area under the curve, CI: confidence interval.

As MIF expression levels in tissue samples of squamous cell carcinomas (SCC) have been found to be associated with tumor progression [33], further analyses of MIF in patients with SCC were performed. No differences in MIF concentrations between responders and non-responders to chemotherapy within the SCC cohort were observed in pretherapeutic samples (p = 0.919), at cycle 2 (p = 1.000) or at cycle 3 (p = 0.546). Additionally, no significant differentiation between responders and non-responders was found in ROC curve analyses of pretherapeutic MIF levels (p = 0.897), at cycle 2 (p = 0.979) or at cycle 3 (p=0.520). Likewise, Kaplan-Meier analyses yielded no significant differences in survival times of SCC patients after stratification according to the respective median serum MIF concentrations pretherapeutically (p = 0.190), at cycle 2 (p = 0.586) or at cycle 3 (p = 0.314).

#### 3.4. ROC curve analyses of therapy response in NSCLC

Results from ROC curve analyses of all three time points are shown in Table 4, while ROC curves are depicted in Fig. 1. While proGRP was able to significantly discriminate between therapy response and tumor progression at cycle 2, MIF failed to distinguish responders from non-responders at all time points. Accordingly, Sens<sub>90</sub> and Sens<sub>95</sub> values were well below 20% and 10%, respectively, with the exception of proGRP at cycle 2 reaching values above 40%.



Fig. 1. ROC curves of MIF (blue) and proGRP (green) in NSCLC patients with tumor progression as state variable. a) pretherapeutic biomarker concentrations, b) concentrations at cycle 2 and c) concentrations at cycle 3.

## 3.5. Prognostic value of MIF and proGRP in NSCLC

To assess the prognostic value of MIF and proGRP in NSCLC patients, the NSCLC group was stratified according to the respective median serum concentrations. Results of subsequent analyses of overall survival with Kaplan-Meier curves are given in Table 5 and Fig. 2. At cycle 2, proGRP levels above the median value of 13.0 pg/mL were associated with significantly shorter overall survival. At other time points, and at all time points in the case of MIF, no significant differences in survival times between low and high biomarker concentrations were obtained.

#### 3.6. Analyses in the SCLC subgroup

In ROC curve analyses, neither proGRP nor MIF could differentiate significantly between responders and non-responders before the start of chemotherapy (proGRP: p = 0.445/MIF: p = 0.572), at the second chemotherapy cycle (p = 0.414/p=0.671) or at the third chemotherapy cycle (p = 0.217/p=0.203).

In Kaplan-Meier analyses after stratification by median serum concentrations of proGRP, no difference in survival times before chemotherapy (p = 0.071), at cycle 2 (p = 0.759) or at cycle 3 (p = 0.816) was found. When the SCLC subgroup was stratified according to the respective median MIF concen-

Biomarker			Above threshold		Below threshold			
	Threshold	N	Median	95%-CI	N	Median	95%-CI	<i>p</i> -value
Pretherapeutic								
MIF	5.91	40	211	120 - 302	39	195	138 - 252	0.655
proGRP	16.0	24	190	68 - 312	27	291	19 – 563	0.216
Cycle 2								
MIF	6.16	39	211	136 - 286	39	205	90 - 320	0.858
proGRP	13.0	11	199	155 - 243	12	676	_	0.013
Cycle 3								
MIF	7.19	37	184	113 – 255	37	279	177 – 381	0.183
proGRP	13.0	15	216	150 - 282	16	429	0 - 876	0.254

Table 5 Kaplan-Meier analyses of NSCLC patients

CI: confidence interval.

trations, again no significant difference in survival times between patients with high and those with low serum MIF levels before chemotherapy (p=0.856) or at cycle 2 (p=0.198) was obtained. At cycle 3, Kaplan-Meier analysis resulted in significantly longer survival times for patients with serum MIF concentrations below the median of 3.62 ng/ml (median survival 596 d vs. 144 d, 95% confidence interval: 213–980 d vs. 130–158 d, p=0.029).

## 4. Discussion

Lung cancer is still a severe burden to public and individual health. It is among the most frequent and most lethal types of cancer both in industrial and in developing countries. Although novel treatment options, such as immune checkpoint inhibitors expand therapeutic strategies, success rates remain relatively low in advanced stages [37]. Biomarkers have the potential to guide therapeutic decision making as well as provide a basis for prognosis. A biomarker with promising biochemical properties and a vital position within oncological pathways is MIF. Not only does MIF influence crucial processes such as apoptosis, it also serves as a link between tumorigenesis, inflammation and the immune system [38]. MIF has been shown to promote the tumor microenvironment, for instance, by stimulating tumorassociated macrophages [28, 29]. Additionally, MIF may affect the cell cycle and cell growth by interacting with pathways involving p53, Bcl-2 or PI3K/Akt [13, 17, 18]. Additionally, MIF supports immunoevasion of tumor cells by promoting suppression of T-cell activation [39–41] and activation of immunosuppressive functions of myeloid-derived suppressor cells [42, 43].

It has already been shown that MIF may facilitate prognosis in lung cancer [32, 33]. In the present study, we investigated the possible use of MIF and proGRP for prediction of therapy response and stratification of overall survival in a cohort of lung cancer patients. Pretherapeutic concentrations of MIF and proGRP in NSCLCs and SCLCs did not differ between responders and non-responders to subsequent chemotherapy. Therefore, it was assumed that initially, the biochemical status concerning MIF and proGRP was similar in responders and non-responders. Moreover, the histopathology within the NSCLC group did not influence pretherapeutic biomarker levels. In contrast to expectations, no differences in MIF concentrations between responders and non-responders occurred in the course of chemotherapy. As MIF promotes tumor growth and tumor microenvironment, tumor progression in non-responders was thought to increase MIF secretion as compared to responders.



Fig. 2. Kaplan-Meier curves of NSCLC patients of a) pretherapeutic MIF concentrations, b) MIF levels at cycle 2, c) MIF levels at cycle 3, d) pretherapeutic proGRP concentrations, e) proGRP levels at cycle 2, f) proGRP levels at cycle 3.

The only significant results were obtained for proGRP in the NSCLC group before the second chemotherapy cycle, with responders possessing lower serum concentrations of proGRP and accordingly longer survival times in patients with lower proGRP levels. A possible explanation for this interesting finding is that after the start of chemotherapy and the onset of therapeutic response (or lack thereof in non-responders) with reduction in tumor burden, an increasing difference in serum marker concentrations and better differentiation between responders and non-responders based on these concentrations may occur. However, these considerations do not account for the lack of statistical significance before the third chemotherapy cycle. A thorough explanation for this convergence of proGRP concentrations at cycle 3 cannot be given at the moment.

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## 4.1. Comparison with established biomarkers

We have already reported results for the established biomarkers carcinoembryonic antigen (CEA) and cytokeratin-19 fragments (CYFRA 21-1) as well as for the immunogenic biomarker high mobility group box 1 protein (HMGB1) in the present cohort [34]. CYFRA 21-1 was well able to discriminate between responders and non-responders to chemotherapy in the case of NSCLCs, with significantly lower concentrations of CYFRA 21-1 in responders after the second and third chemotherapy cycle. In contrast, CEA allowed for a good differentiation between responders and non-responders in SCLCs. Furthermore, not only did CYFRA 21-1 show a good performance in ROC curves of NSCLC patients with areas under the curve above 0.7, but also yielded a significant prognostic value in Kaplan-Meier curves, with overall survival being 3.7-4.8 times higher in patients with a low CYFRA 21-1 concentration than in those with high serum levels. Similar results have also been obtained by other groups (reviewed in [44]). Clearly, both MIF and proGRP could not compete with these biomarkers. This is a remarkable result, since MIF has been linked to inflammatory processes which also occur within a tumor environment. Additionally, our study comprised patients in advanced tumor stages, tantamount to relatively high tumor mass and therefore not only a high relative, but also a potentially large absolute reduction of tumor mass in the case of responders to therapy. Hence, it was assumed that response to therapy would be associated with a decrease in serum levels of MIF, and that MIF in turn may serve as a significant factor in overall survival as determined by Kaplan-Meier analyses. It is unclear whether this is an intrinsic characteristic of lung cancer in general or a specific behavior of the cohort investigated here. However, as outlined above, studies concerning the role of MIF in cancer have found somewhat contradicting results. Our current data indicate that the subtype of NSCLCs does not change the secretion of MIF significantly, thereby ruling out the possibility of a histopathological sampling effect. It is unclear whether the lack of a change in MIF concentrations is influenced by chemotherapy itself in this cohort. Chemotherapeutic agents such as cisplatin or doxorubicin are known to promote inflammatory signal transduction [45]. However, due to the heterogeneous therapies and the general design of the present study, it was not possible to determine if the respective therapies themselves were able to elevate MIF concentrations and counterbalance a possible decrease of MIF serum levels in responders. In the SCLC subgroup, low MIF concentrations before the third chemotherapy cycle were associated with significantly longer survival times. This may indicate that serum MIF levels could be useful as a prognostic tool in SCLC patients. However, it must be emphasized that the significance of such a result is clearly limited due to the very limited number of SCLC patients in this study.

Therefore, more research is needed towards the effects and interlinkage of MIF in biochemical pathways involved in lung cancer progression. At present, we conclude that there is no indication that serum MIF levels may serve as a biomarker in prognosis and monitoring of response to chemotherapy in NSCLC. From the present data, no conclusions can be drawn regarding the applicability of MIF in immunotherapy of lung cancer, for instance with PD-1/PD-L1 inhibitors in advanced stages of NSCLC. Due to the constant advancement in therapeutic options with biologicals, more research towards the role of MIF and similar biomarkers in patients treated with such novel therapies is needed.

Obviously, there are some limitations of the study like the retrospective design, the inclusion of a larger NSCLC and a smaller SCLC subgroup, the classical chemotherapeutic treatment, the use of a non-diagnostic immunoassay (RUO-test) for MIF measurement and the lack of a validation cohort. However, samples were analyzed in a standardized procedure regarding collection, storage, treatment prior to measurement and analysis. High quality standards were obtained by standardized calibration curves, internal controls, measurement of serial samples within the same runs and final crossplate checks to minimize inter-assay variations. Radiological imaging analyses were performed by experienced, independent radiologists and data interpretation was done independently from data acquisition.

## 5. Conclusion

The present study unfortunately revealed the lack of clinical utility of soluble MIF in serum samples of advanced lung cancer patients for monitoring therapy response and estimating prognosis in both NSCLC and SCLC subtypes.

## Acknowledgments

We thank the biobank of the Ludwig Maximilians-University (LMU) Munich for storage of serum samples.

## **Author contributions**

CONCEPTION: S.H. and J.v.P. DATA CURATION: S.H., J.v.P., S.B., N.T. and A.R. ANALYSIS OF DATA: S.B., A.R. and S.H. PREPARATION OF THE MANUSCRIPT: S.H. and A.R. REVISION FOR IMPORTANT INTELLECTUAL CONTENT: J.v.P., S.B. and N.T. SUPERVISION: S.H. and J.v.P.

# **Conflict of interest**

S. H. is one of the editors of the special issue "Lung Cancer Tumor Markers" but had no involvement in the peer review process of this article. The other authors declare no potential conflicts of interest with respect to the research, authorship, and/or publication of this article.

## **Ethical considerations**

The study was conducted according to the guidelines of the Declaration of Helsinki, and approved by the Ethics Committee (IRB) of the Ludwig Maximilians-University (LMU) Munich (UE-Nr 114-13, date of approval: 05.08.2013). Anonymized biobanked serum specimens of the LMU Munich were used for this retrospective study as specified in the IRB approval (waiver for informed consent).

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