

Journal Pre-proofs

Multi-target optimization of solid phase microextraction to analyse key flavour compounds in wort and beer

Ana C. Vieira, Ana C. Pereira, José C. Marques, Marco S. Reis

PII: S0308-8146(20)30328-9
DOI: <https://doi.org/10.1016/j.foodchem.2020.126466>
Reference: FOCH 126466

To appear in: *Food Chemistry*

Received Date: 29 November 2019
Revised Date: 19 February 2020
Accepted Date: 20 February 2020



Please cite this article as: Vieira, A.C., Pereira, A.C., Marques, J.C., Reis, M.S., Multi-target optimization of solid phase microextraction to analyse key flavour compounds in wort and beer, *Food Chemistry* (2020), doi: <https://doi.org/10.1016/j.foodchem.2020.126466>

This is a PDF file of an article that has undergone enhancements after acceptance, such as the addition of a cover page and metadata, and formatting for readability, but it is not yet the definitive version of record. This version will undergo additional copyediting, typesetting and review before it is published in its final form, but we are providing this version to give early visibility of the article. Please note that, during the production process, errors may be discovered which could affect the content, and all legal disclaimers that apply to the journal pertain.

Multi-target optimization of solid phase microextraction to analyse key flavour compounds in wort and beer

Ana C. Vieira ^a, Ana C. Pereira ^{a,b,c,*}, José C. Marques ^{a,c}, Marco S. Reis ^b

^a *Faculty of Exact Sciences and Engineering, University of Madeira, Campus da Penteadá, 9020-105 Portugal*

^b *Chemical Process Engineering and Forest Products Research Centre, Department of Chemical Engineering, University of Coimbra, Pólo II - Rua Sílvia Lima, 3030-790, Portugal*

^c *Institute of Nanostructures, Nanomodelling and Nanofabrication (I3N), University of Aveiro, 3810-193, Portugal*

* Corresponding author: Tel: +351 291705122; E-mail: apereira@eq.uc.pt or

ana.pereira@staff.uma.pt

Ana C. Vieira – 0000-0003-3701-9648

Ana C. Pereira – 0000-0003-3097-7704

Marco S. Reis - 0000-0002-4997-8865

José C. Marques - 0000-0002-4832-4341

Abstract

Despite the literature comprises numerous studies dealing with the analysis of wort and beer flavour-related compounds by HS-SPME followed by GC-MS quantification, no generalized consensus exists regarding the optimal conditions for the extraction procedure. The complex chemistry nature of these matrices, the number of analytes, as well as the number and interactions among parameters affecting the extraction performance, requires the adoption of optimal experimental design protocols. This aspect is often overlooked and often not properly addressed in practice. Therefore, in the present work, the optimal conditions under which a range of wort and beer analytes can be extracted and quantified were analyzed. The optimal extraction conditions were presented at two levels of aggregation: global (untargeted) and key-flavour analysis. Experimental data was generated by Definitive-Screening-Design, followed by model development and optimization. Both approaches were compared and critically analyzed. For vicinal-diketones group, a complete validation study for the optimal conditions is presented.

Keywords: Optimization of analytical processes; Design of experiments; HS-SPME; Wort and beer; Flavour; Principal Component Analysis; Definitive Screening Designs;

1. Introduction

The global production of beer has been experiencing a steady and robust increasing trend in the last decade, establishing it in the top rank of the most consumed and popular alcoholic beverages. The increasing consumption and production volume allied to the market demand and consumer's preferences have also undergone changes which call for innovative technologies and a more comprehensive knowledge of the production process in breweries, in order to understand and better respond to the drivers of demand. In this regard, several studies have been focusing on the analysis of the volatile organic compounds (VOCs) formation during beer fermentation and storage conditions in order to eliminate or promote specific flavours in the final product.

The headspace solid phase microextraction (HS-SPME) is the most used extraction technique before gas chromatography coupled to mass spectrometry (GC-MS) quantification of beer volatile compounds (Andrés-Iglesias, Montero, Sancho, & Blanco, 2015; Braga, Zielinski, Silva, de Souza, Pietrowski, Couto, et al., 2013), both for its efficiency as well as operability. Table 1 shows a systematic overview of the parameters and the conditions usually adopted for the extraction of VOCs in beer and wort, as well as the optimization procedure adopted to set the optimal value of HS-SPME parameters. This literature review reveals that, despite being a widely used technique, it is still quite difficult to establish the optimal conditions for analysing beer VOCs. For example, regarding fiber coating, there are five references to different coatings as being optimal to quantify beer VOCs. The optimal extraction time and temperatures also present significant differences, varying from 30 to 60 min and from 30 °C to 60 °C, for the same fiber coating. Significant differences can also be found in the optimal values of the remaining extraction parameters, namely the degas and salt addition effect, the sample/vial volume ratio, the use of stirring and sample pre-incubation. This lack of agreement and consensus may be caused by the adoption of sub-optimal optimization practices, such as one-factor-at-a-time (OFAT) strategies (Bezerra, dos Santos, Santos, Novaes, Ferreira, & de Souza, 2016; Granato & Calado, 2013; Pereira, Reis, Leça, Rodrigues, & Marques, 2018). Besides completely overlooking factor interactions, OFAT

occasionally induce some parameters to be fixed and not considered in the optimization procedure.

Statistical design of experiment methodologies (DoE) overcome OFAT limitations, guaranteeing statistically meaningful results with reduced experimental effort (Ferreira, Silva Junior, Felix, da Silva, Santos, Santos Neto, et al., 2019). The most common used DoE methodologies in HS-SPME optimization procedures are the Full Factorial Design (FFD) and CCD (Central Composite Design). In general, such procedures are applied to two, up to four parameters, and to quantitative variables, such as time, temperature and sample/vial volume ratio. Fiber coating optimization was only considered by Leça *et al.* (2015) in the optimization of HS-SPME for quantifying just two beer off-flavour compounds. A new relevant addition to the DoE toolkit was recently introduced with the development of Definitive Screening Designs (DSD) (Bradley Jones & Nachtsheim, 2011). DSDs are a new class of three-level screening designs, with the capability for estimating quadratic effects of the model (if more than six factors are contemplated). In these designs, the main effects are completely decoupled from second-order interactions, making their estimation very efficient (as expected for a screening methodology). Furthermore, second-order interactions are also not aliased with each other, making the pattern of aliasing of these designs very interesting when compared to other alternatives. DSDs only require two more treatments than twice the number of factors, making them highly competitive from a cost-benefit perspective. The design matrix for DSDs is generated from numerical methods applied to solve D-optimal designs formulations or, more simply, using conference matrices (Xiao, Lin, & Fengshan, 2012). For more details, please refer to (Bradley Jones & Nachtsheim, 2011; B. Jones & Nachtsheim, 2013).

In the present work, we addressed the task of systematically finding the optimal conditions for quantifying eight groups of compounds (multi-target optimization). These groups include the major chemical families of VOCs that can be found in beer, namely the higher alcohols and esters, and less abundant molecules but presenting great importance in beer-flavour such as aldehydes, volatile fatty acids and vicinal diketones. The latter compounds, the vicinal diketones (diacetyl and 2,3-pentanedione) are also key markers for fermentation process monitoring in brewers. We also test the alternative untargeted methodology, where a compromise solution for all the

compounds (28 overall) is sought. Both approaches are compared and critically analysed. We also fully validate the quantification for the particular case of VDK, given their particular interest to beer producers and the higher quality of the models obtained.

This article is organized as follows. In the following section, the materials and experimental methods employed in this work are described in detail. Then, the methodological workflow for data generation and analysis is introduced. The results obtained are reported in the Section 3, including the optimal operation conditions for HS-SPME regarding untargeted (global) and multi-target analysis of key-flavour groups of compounds. Specific optimal conditions for one family (VDK) are then fully validated, and the method is applied to independent samples ranging from fermenting wort to different packed beer brands. The paper closes with a final section summarizing the main contributions and conclusions.

2. Materials and methods

In this section we present a detailed description of the materials employed in the study and the analytical procedures followed.

2.1 Materials, Reagents, Chemicals and Samples

The SPME fibers tested were purchased from Supelco (Bellefonte, PA, USA): a stableflex fiber core coated with 50/30 μm divinylbenzene-carboxen-polydimethylsiloxane (DVB/Car/PDMS), a 85 μm carboxen-polydimethylsiloxane (Car/PDMS) and divinylbenzene- polydimethylsiloxane (DVB/PDMS). Absolute ethanol (>99.8%) was purchased from Sigma–Aldrich (Steinheim, Germany). Ultra-pure water (conductivity of 18 $\text{M}\Omega$) was obtained by the Simplicity®UV ultrapure water (type 1) apparatus from Millipore (USA). A calcium chloride ($\text{CaCl}_2 \cdot 2\text{H}_2\text{O}$, $\geq 99.5\%$, Chem-Lab, Zedelgem, Belgium) aqueous solution (50 g/L) was also prepared for sample conservation purposes when yeast was present. One millilitre was added per 50 ml of sample collected during fermentation and then stored at $-26\text{ }^\circ\text{C}$ until analysis. Sodium chloride (> 99.8%, Chem-Lab, Zedelgem, Belgium) was used in the HS-SPME extraction process. 1 g/L stock solution of 4-methyl-1-pentanol (97%, Sigma-Aldrich, Steinheim, Germany) was prepared by

dissolving this compound in synthetic beer and it was used as an internal standard (500 µg/L in the sample). The alkane solution (C7 to C30) was obtained from Supelco (Sigma Aldrich, St. Louis, MO, USA). Volatile organic compounds stock solutions were prepared in synthetic beer (SB) or in ethanol, whenever the solubility of the analytes was too low in aqueous solution. All the standards used have a purity grade of more than 97.0 %, except for trans-2-nonenal (95 %), adequate to GC-MS analyses. Trans-2-nonenal, ethyl isovalerate, 2,3-pentanedione, ethyl hexanoate, isoamyl acetate, isovaleric acid, ethyl caprylate, acetaldehyde and phenylethyl alcohol were purchased from Acros Organics (Geel, Belgium). Dimethyl sulphide, 2-methoxy-4-vinylphenol (vinyl guaiacol), phenylethyl acetate, isoamyl alcohol and isobutyraldehyde were purchased from Sigma-Aldrich (Steinheim, Germany). Diacetyl, ethyl butyrate, isobutyl acetate, octanoic acid, acetoin and isobutyl alcohol were purchased from TCI (Zwijndrecht, Belgium). Ethyl acetate was purchased from Fisher Scientific (Loughborough, United Kingdom).

The following matrixes were prepared to determine the presence of matrix effect: a) Lager beer from a local brewery with an ethanol content of 5.1%; b) a synthetic beer which was prepared by adding the adequate volume of absolute ethanol for a final concentration of 5.1% (v/v) in ultrapure water and by adjusting mixture pH to 4.0 with an aqueous solution of sodium hydroxide (98%, Panreac, Barcelone, Spain).

About 24 samples were considered in this study to evaluate the methodology applicability, including samples collected during the fermentation process and the final product. To be more precise, seven beer samples from different brands (labelled as samples 1 to 7) and several fermenting wort samples (samples 8 to 24), obtained from a local brewery were analysed.

2.2 HS-SPME conditions

Several qualitative and quantitative parameters were evaluated according to the literature review presented in Table 1. After this preliminary screening of factors, the following factors were selected to be contemplated in the analysis: the sample degasification influence (no degass and 15 minutes in ultrasonic bath), the sample/vial volume ratio (in particular, the sample volume in a 20 mL vial), the sample agitation (without and at 100 rpm), salt addition effect, the pre-

incubation and extraction time, as well as extraction temperature. The type of fiber was also considered, namely three different fiber coatings were studied (PDMS/DVB, CAR/PDMS and CAR/PDMS/DVB). A preliminary analysis using design of experiments revealed that, very clearly, CAR/PDMS was the best option among the different fiber coatings under analysis. Therefore, this factor was then kept fixed to CAR/PDMS during the study. In Table 2, the levels considered in the experimental design, for all parameters, are described. All SPME tests were carried out on a TriPlus autosampler, in SPME mode and were properly conditioned before using, according to the supplier indication.

Optimal extraction conditions were chosen after a careful statistical analysis of the experimental design results (Definitive Screening Design) from 18 randomly performed assays (see details in the next section). Once the optimized extraction conditions were found, in each vial, 5 μ l of 4-methyl-1-pentanol (1 g/L) were added to the sample before properly sealing and homogenizing it in a vortex. Also, according to the ethanolic content of each sample, either the volume of ethanol was added or a dilution with water was performed to adjust ethanol concentration to 5.1 % and, consequently, uniformize the interference of this compound in the efficiency of the extraction.

2.3 HS-SPME optimization: multi-targeted and untargeted methodologies

The optimization of the operational conditions for simultaneously quantifying multiple components in a complex mixture, such as beer or wort, is an involved process. Most often, the research efforts are directed for quantifying a single component or family or components, for which the operation conditions are to be optimized. In this work we considered the case where multiple targets are considered, for which the optimal conditions, possibly different, need to be found. This is a gap found in the literature that we aim to mitigate in this work. Furthermore, we also considered the untargeted case where all compounds are to be considered simultaneously and a single set of conditions is to be proposed. This approach is often found in the literature when multiple compounds are present and will be further explored and critically analyzed in this work. The proposed systematic methodology consists of a logical sequence of steps. First, factors are selected (as described in section 2.2). Then a DSD design is set and executed in random order

(generating two plus twice as many experiments as factors). The experimental results are analyzed using multiple linear regression, after aggregating (summing) responses in families of compounds (multi-targeted approach) or into a single response, by computing the scores of the first principal component (PC1), in a Principal Component Analysis (PCA) (untargeted approach, also called, global approach). The obtained models are finally optimized using a desirability-based approach (in this case, the goal is to maximize the peak area of the chromatogram), leading to the optimal factor levels.

2.3.1. Design of experiments

Being a data-driven methodology, it is very important to collect high quality data to support the subsequent steps of the workflow. In this sense, the Definitive Screening Design were applied using JMP-PRO ver. 12.1.0 (64-bit) (SAS Institute Inc.).

2.3.2 Multi-targeted and untargeted data analysis

In this work, two aggregation levels were considered. In multi-targeted analysis, the responses are the sums of compounds in each group (the most common chemical families of VOCs found in beer). In the untargeted or global methodology, the information of all the 28 compounds present are aggregated into a single response, by taking the linear combination of the compounds that most explains the original variability in all compounds. This is done through Principal Component Analysis (PCA). In the PCA nomenclature, the linear combinations are known as loading vectors or simply loadings (one such vector per principal component) and the result of the linear combinations applied to data are the scores (again, there is one score vector for each component). The importance of each component is analysed by the relative magnitude of the associated eigenvalue, which can also be used to compute the amount of variation of the original dataset it is able to explain (Granato, Santos, Escher, Ferreira, & Maggio, 2018). PCA summarizes the information of all analytes in uncorrelated Principal Components (scores) that explain most of the experimental variability – the global level of analysis.

2.3.3 Model building

In this stage, a model is derived that relates the experimental factors with the responses. As data was generated using DoE, polynomial models are to be estimated, consisting of combinations of main effects, second-order interactions and eventually quadratic factors. These models are estimated by Ordinary Least Squares (OLS), assisted with variable selection methods, such as the Forward Stepwise Variable Selection methodology (Draper & Smith, 1998) that sequentially selects the effects to include in the model that improve, in a statistically significant manner, its explanatory capability. At each round, irrelevant variables can also be discarded, if their effect ceases to be statistically relevant. The algorithm stops when no variables are added or removed from the model.

2.3.4 Optimization

Finally, the optimization of the estimated model is made by maximizing the selected desirability function (Ballus, Meinhart, de Souza Campos, Bruns, & Godoy, 2014; Candiotti, DeZan, Cámara, & Goicoechea, 2014). This leads to the optimal settings for the experimental factors (in our case, the seven HS-SPME factors presented in Table 3; the blanks cells in this table indicate that the corresponding factors (indicated in the rows) do not have a statistically tangible effect on the quantification of the specified compounds (indicated in the columns) within the experimental domain explored).

2.4 Chromatographic conditions

All analyses to identify and quantify VOCs were carried out using a GC-MS system, the TRACE GC Ultra gas chromatograph coupled with an ISQ single quadrupole from Thermo Scientific (Hudson, NH, USA). We have employed a capillary column, TRB-WAX 60 m × 0.25 mm ID DF = 0.25 (Teknokroma, Spain). The carrier gas was helium at a constant flow rate of 1 mL/min. The injector port was kept at 260 °C, in splitless mode, while the transfer line and the ion source were maintained at 260 and 240 °C, respectively. The oven temperature program started at 40 °C, hold for 2 min, then increased up to 250 °C at 4 °C/min and it was finally kept at 250 °C for 5 min. The total GC run time rounded about 60 min. The first chromatograms were obtained in full scan mode (total ion count) in order to obtain the retention time (t_{RS}) of each target compound. After

the confirmation of the t_{RS} , the analyses were always performed with the characteristic and major ions of each analyte and characteristic ions were used for quantification purposes (presented in Results and Discussion section, Table 4).

The mass spectrometer was operated in electron impact (EI) mode at 70 eV. The selective ion monitoring (SIM) operating mode was used with the characteristic ions for each target compound. Each calibration point was extracted in triplicate within the validation range of each analyte.

2.5 Validation of the analytical methodology

The optimal experimental condition for implementing HS-SPME-GC-MS for the quantifying family of VDKs was fully validated in terms of selectivity, linearity, sensitivity, matrix effect, precision and accuracy, for compounds found in beer samples and for which chemical standards are available. Each VOC compound area ratio (VOC area/IS area), from all increasing standard solutions, was plotted against the corresponding concentration in order to obtain the respective calibration curves. Selectivity was confirmed by the absence of chromatographic interferences at the t_{RS} of the VOCs in a local lager beer, used for the matrix-matched calibration and several other commercial beers and fermentation samples.

The linearity (R^2), limit of detection (LOD) and limit of quantification (LOQ) were computed based on a linear regression approach, following the IUPAC recommendations. The values obtained for these quantities were: $LOD = 3.3 \sigma/b$ and $LOQ = 10 \sigma/b$, where σ stands for the standard deviation of the regression and b is the slope. A matrix effect (ME) study was also carried out in order to verify the influence of beer matrix in the extraction of target compounds, when compared to SB. The ME was calculated according to Leça *et al* (2015).

Intra-day precision was determined by the quantification of 10 successive replicates of SB spiked with three different standard solutions (low, intermediate and high concentrations of each analyte) and inter-day precision was assessed by performing the same analysis in 3 different days over a week. The values were expressed in terms of relative standard deviation. A recovery study was carried out (at the above-mentioned three levels of concentrations) to determine the method's accuracy, by computing the percentage of variation between the expected theoretical

concentrations, in each level, and the mean values obtained for the concentrations, by applying the proposed methodology.

Carry-over was also tested by running a blank sample after extracting the highest concentrated working standard solution of each analyte, which allowed us to adjust the fiber conditioning temperature and time.

3. Results and discussion

In this section, the results obtained during the implementation of the workflow described in Section 2.3 are presented. The responses for the multi-targeted analysis consist of the sums of peak areas for all the compounds in each one of the eight groups. At the global (untargeted) level, the compounds are analysed simultaneously, as a whole, considering their mutual correlations. The analysis at the global level was conducted using PCA. The first component (PC1) explains over 40% of original variability whereas the first three components explain altogether 80%. The experimental results are scattered in the two-dimensional PC1-PC2 plane, not showing any obvious outlier or abnormal experiment (data not shown). The same was observed for the distribution of residuals around the PCA subspace (using the squared distance to the PCA subspace – not shown) which also did not reveal any outlying observations.

The loadings for the first three components are presented in Figure 1. These are the coefficients of the linear combinations for each PC and codify mutual correlation affinities. The first PC, representing the main overall correlation trend, has only positive coefficients, meaning that it is a sort of weighted mean of all analyte responses. The higher the coefficients, the stronger the correlations between the analyte responses. Esters, such as isobutyl acetate, isoamyl acetate, ethyl butyrate, ethyl hexanoate and ethyl caprylate (central part of the plot) are less represented in PC1 and more in PC2, which seem to represent a contrast between the quantification of these analytes and all the other components. 1-butanol shows a rather peculiar behaviour, widely different from any other analyte (it has a low magnitude loading coefficient in the first PC, meaning that the

variation of this analyte is not strongly correlated with the dominant sources of structured variability present in data). In summary, PC1 has a limited ability to represent all the compounds present, and any solution based on it may be limited in terms of the performance achieved for the less well-represented compounds. By seeking the best compromise between compounds with different patterns of interaction with the HS-SPME factors, PC1 also becomes a mixture of effects, with a rather erratic dependency upon experimental factors, potentially leading to poorer models.

3.1 Model building and optimization

For the multi-targeted analysis, a regression model was developed for explaining each one of the eight groups of compounds such as: VDKs, acetaldehyde, acetoin and 2,3-butanediol, majority and minority esters, higher alcohols and fatty acids. In Table 3, the model figures of merit and the optimal HS-SPME settings for each case are presented. Also presented are the results for the global (untargeted) approach (last column). The best model corresponds to the estimation of the VDKs (more specifically, the sum of diacetyl and 2,3-pentanedione), based on the seven experimental factors and two-way interactions. The resulting model is significant (p -value = 0.0028) with a good quality of fit ($R^2 = 0.99$; Predicted $R^2 = 0.92$; see Figure 2). The models targeting other components present less explanation power but are still significant (significance level of 0.01), as well as the factors identified, which enable the determination of the most suitable operation conditions.

Regarding the global level (last column of Table 3), the quality of the model is, as expected from the discussion above, quite limited (low R^2). The global response is PC1, which is a result of the combination of the behaviour of all analytes with possibly conflicting dependencies upon the experimental factors. Therefore, it represents a sort of compromise solution that does not optimally fulfil the conditions to quantify any analyte, with the exception perhaps of those that most structure the variation originated in the DOE.

In summary, Table 3 presents our proposal for setting the extraction conditions under multi-targeted and untargeted conditions, which are valid for the extraction of volatile organic compounds from beer using a Car-PDMS fiber. It was found that carbonation does not significantly influence SPME sampling, in accordance with previous results reported by Pizarro, Pérez-del-Notario (Pizarro et al., 2010). Concerning the sample volume, the results suggest that 10 mL in a vial with 20 mL of capacity is the best option, since the extraction was not favoured when smaller sample amounts were used. Regarding the extraction time and temperature, the results indicate that these factors have a significant impact for almost every chemical families that were analysed, and their settings were established according to VDKs best results. The agitation and incubation time factors were found to be relevant only for VDKs and acetoin, and therefore their settings were established based on the results of these compounds.

3.2 Validation and independent testing

In this section, we complete the cycle of developing an analytical procedure, by fully validating a pre-selected target. The proposed multi-targeted approach leads to optimal conditions for different families of compounds. Among these, VDKs are a family of analytes with considerable interest for beer producers. Therefore, we selected this target to illustrate the usefulness of the proposed methodology. In this case, we take the optimal conditions in Table 3 and conduct a complete validation study and compute the figures of merit for the optimal conditions. The analytical methodology validation was carried out for analytes which chemical standards are available. Additionally, three other compounds were also considered in this phase (dimethyl sulfide, isobutyraldehyde, trans-2-nonanal) since these are common off-flavours in beer and chemical standards are available.

3.2.1 Assessment of matrix effects

Following the extraction conditions described above, injections of standard solutions prepared in beer and in synthetic beer were analysed and calibration curves for all the analytes in study were determined in both cases. From these results, the matrix effect was studied by comparing the

deviations between the slopes of the calibration curves resultant from beer and synthetic beer. It was observed that this parameter was lower than 15% for most of the target compounds, confirming that matrix effect does not occur (Table 4). For that reason, SB was used for the quantification of VOCs and for the validation of this methodology

3.2.2 Figures of merit of the methodology proposed

The selectivity of the methodology was confirmed by the absence of interferences at the retention times associated to each VOC in the calibration solutions, as well as the absence of any coeluted compounds with target compounds when beer was analysed using the same methodology. The parameters of each analyte linear regression curve, as well as validation results can be found in Table 4. Chromatograms resultant from a standard solution and from a lager beer sample injection are shown in Supplementary material section. All the regression curves were found to exhibit good scores in terms of linearity, R^2 (higher than 0.999 for all compounds analysed). LOD and LOQ values (evaluated from the standard deviation of the regression line) were appropriate for the linear ranges established for the target compounds, which most of the times were found both in wort fermenting samples and in several commercial beer from different brands and types (Table 4). The developed quantification method shows a good linearity in the concentration range of interest and recovery mean values ranging from 84.76 (diacetyl) and 114.86% (ethyl decanoate), confirming the method accuracy. The methodology precision was evaluated in terms of intra-day and inter-day precision, where variations were expressed as curve residual standard deviation values, which were generally lower than 10%.

3.2.3 Application of methodology to wort and beer samples

The optimized HS-SPME methodology was successfully applied to the quantification of VOCs in 7 beers and in 17 fermentation samples (Supplementary material section). The focus of this work relies on the possibility of applying a single analytical method to quantify the VOCs that are typically resultant from alcoholic fermentation of beer wort and other significant aromas such as the off-flavours dimethyl sulphide, isobutyraldehyde, trans-2-nonenal and vinyl guaiacol.

Sporadically, some analytes are present below their LOQ in specific samples but, once these are fermentation compounds, it was not expected to detect them in early fermentation samples. This was observed in some higher alcohols and esters, namely phenylethyl alcohol, ethyl acetate and ethyl butyrate. Furthermore, analytes such as the vicinal diketones (diacetyl and 2,3-pentanedione), which monitoring is crucial during the fermentation process, were quantifiable in all the samples that were analysed. These results reinforce the adequate applicability and performance of the developed methodology, particularly in terms of the wide linear range in which samples collected during the entire brewing process can fit.

4. Conclusions

The present study reports a multi-targeted and global methodologies for establishing the optimal extraction conditions regarding HS-SPME for quantifying a wide variety of VOCs in beer, including wort to beer fermentation samples, as well finished beers.

The proposed methodology consists of a sequence of steps, starting with the identification of experimental factors, application of a suitable design of experiments methodology (Definitive Screening Design) to plan the experimental trials, followed by their randomized execution, multi-targeted and global analysis of results (i.e., at different levels of aggregation of analyte information), and finally optimization leading to the optimal operation settings. The optimal settings can be flexibly used in the future, depending on the specific purpose of the analysis.

In addition, for illustration purposes, one of the conditions was fully validated and tested.

The optimal HS-SPME settings proposed in this study can now be adopted for performing key-flavour quantification studies in wort and beer. Moreover, the procedure can be replicated in the future for different target analyte or families of analytes (or even for the global approach), to establish the optimal conditions for operating the analytical techniques.

Acknowledgements

This work was financed by FEDER, PROCiência 2020 under the SuperPRO I project (M1420-01-0247-FEDER-000007). The authors acknowledge *Empresa de Cervejas da Madeira* for kindly supplying most of the samples that were used. Ana C. Pereira is thankful to *ARDITI (Agência Regional para o Desenvolvimento da Investigação Tecnologia e Inovação)* for the funding of the research grant under the project M1420-09-5369-FSE.

References

- Alvim, R. P. R., Gomes, F. d. C. O., Garcia, C. F., Vieira, M. d. L. A., & Machado, A. M. d. R. (2017). Identification of volatile organic compounds extracted by headspace solid-phase microextraction in specialty beers produced in Brazil. *Journal of the Institute of Brewing*, *123*(2), 219-225.
- Andrés-Iglesias, C., Montero, O., Sancho, D., & Blanco, C. A. (2015). New trends in beer flavour compound analysis. *Journal of the Science of Food and Agriculture*, *95*(8), 1571-1576.
- Ballus, C. A., Meinhart, A. D., de Souza Campos, F. A., Bruns, R. E., & Godoy, H. T. (2014). Doehlert design-desirability function multi-criteria optimal separation of 17 phenolic compounds from extra-virgin olive oil by capillary zone electrophoresis. *Food Chemistry*, *146*, 558-568.
- Baxter, E. D., Hughes, P. S., & Hornsey, I. S. (2001). *Beer: Quality, Safety and Nutritional Aspects* (first ed.). Cambridge, UK: Royal Society of Chemistry.
- Bezerra, M. A., dos Santos, Q. O., Santos, A. G., Novaes, C. G., Ferreira, S. L. C., & de Souza, V. S. (2016). Simplex optimization: A tutorial approach and recent applications in analytical chemistry. *Microchemical Journal*, *124*, 45-54.
- Biazon, C. L., Brambilla, R., Rigacci, A., Pizzolato, T. M., & dos Santos, J. H. Z. (2009). Combining silica-based adsorbents and SPME fibers in the extraction of the volatiles of beer: an exploratory study. *Analytical and Bioanalytical Chemistry*, *394*(2), 549-556.
- Boulton, C., & Quain, D. (2001). *Brewing Yeast and Fermentation* (first ed.). Cornwall, UK: Blackwell Science.
- Braga, C. M., Zielinski, A. A. F., Silva, K. M. d., de Souza, F. K. F., Pietrowski, G. d. A. M., Couto, M., Granato, D., Wosiacki, G., & Nogueira, A. (2013). Classification of juices and fermented beverages made from unripe, ripe and senescent apples based on the aromatic profile using chemometrics. *Food Chemistry*, *141*(2), 967-974.
- Brányik, T., Vicente, A., Dostálek, P., & Teixeira, J. (2008). A Review of Flavour Formation in Continuous Beer Fermentations. *Journal of the Institute of Brewing*, *114*(1), 3-13.

- Briggs, D. E., Hough, J. S., Stevens, R., & Young, T. W. (1999). *Malting and Brewing Science* (first ed. Vol. Volume II Hopped Wort and Beer). Maryland, USA: Aspen Publishers.
- Cajka, T., Ridellova, K., Tomaniova, M., & Hajslova, J. (2010). Recognition of beer brand based on multivariate analysis of volatile fingerprint. *Journal of Chromatography A*, *1217*(25), 4195-4203.
- Candioti, L. V., DeZan, M. M., Cámara, M. S., & Goicoechea, H. C. (2014). Experimental design and multiple response optimization. Using the desirability function in analytical methods development. *Talanta*, *124*, 123–138.
- Carnol, L., Schummer, C., & Moris, G. (2017). Quantification of Six Phthalates and One Adipate in Luxembourgish Beer Using HS-SPME-GC/MS. *Food Analytical Methods*, *10*(2), 298-309.
- Charry-Parra, G., Dejesus-Echevarria, M., & Perez, F. J. (2011). Beer volatile analysis: optimization of HS/SPME coupled to GC/MS/FID. *Journal of Food Science*, *76*(2), 205-211.
- Cheong, C., Wackerbauer, K., & Kang, S. A. (2007). Influence of aeration during propagation of pitching yeast on fermentation and beer flavor. *Journal of Microbiology and Biotechnology*, *17*(2), 297-304.
- da Silva, G., Augusto, F., & Poppi, R. (2008). Exploratory analysis of the volatile profile of beers by HS-SPME-GC. *Food Chemistry*, *111*(4), 1057-1063.
- da Silva, G. C., da Silva, A. A., da Silva, L. S., Godoy, R. L., Nogueira, L. C., Quiterio, S. L., & Raices, R. S. (2015). Method development by GC-ECD and HS-SPME-GC-MS for beer volatile analysis. *Food Chemistry*, *167*, 71-77.
- De Schutter, D. P., Saison, D., Delvaux, F., Derdelinckx, G., Rock, J.-M., Neven, H., & Delvaux, F. R. (2008). Optimisation of wort volatile analysis by headspace solid-phase microextraction in combination with gas chromatography and mass spectrometry. *Journal of Chromatography A*, *1179*(2), 75-80.

- Ferreira, S. L. C., Silva Junior, M. M., Felix, C. S. A., da Silva, D. L. F., Santos, A. S., Santos Neto, J. H., de Souza, C. T., Cruz Junior, R. A., & Souza, A. S. (2019). Multivariate optimization techniques in food analysis – A review. *Food Chemistry*, 273, 3-8.
- Giannetti, V., Boccacci Mariani, M., Torrelli, P., & Marini, F. (2019). Flavour component analysis by HS-SPME/GC–MS and chemometric modeling to characterize Pilsner-style Lager craft beers. *Microchemical Journal*, 149, 103991.
- González-Álvarez, J., Blanco-Gomis, D., Arias-Abrodo, P., Pello-Palma, J., Ríos-Lombardía, N., Busto, E., Gotor-Fernández, V., & Gutiérrez-Álvarez, M. D. (2013). Analysis of beer volatiles by polymeric imidazolium-solid phase microextraction coatings: Synthesis and characterization of polymeric imidazolium ionic liquids. *Journal of Chromatography A*, 1305, 35-40.
- Granato, D., & Calado, V. (2013). The use and importance of design of experiments (DOE) in process modelling in food science and technology. In, (pp. 1-18).
- Granato, D., Santos, J. S., Escher, G. B., Ferreira, B. L., & Maggio, R. M. (2018). Use of principal component analysis (PCA) and hierarchical cluster analysis (HCA) for multivariate association between bioactive compounds and functional properties in foods: A critical perspective. *Trends in Food Science & Technology*, 72, 83-90.
- Hill, P. G., & Smith, R. M. (2000). Determination of sulphur compounds in beer using headspace solid-phase microextraction and gas chromatographic analysis with pulsed flame photometric detection. *Journal of Chromatography A*, 872(1-2), 203-213.
- Hiralal, L., Olaniran, A. O., & Pillay, B. (2014). Aroma-active ester profile of ale beer produced under different fermentation and nutritional conditions. *Journal of Bioscience and Bioengineering*, 117(1), 57-64.
- Hiralal, L., Pillay, B., & Olaniran, A. O. (2013). Stability profile of flavour-active ester compounds in ale and lager beer during storage. *African Journal of Biotechnology*, 12(5), 491-498.

- Horak, T., Culik, J., Kellner, V., Jurková, M., Čejka, P., Hašková, D., & Dvořák, J. (2010). Analysis of Selected Esters in Beer: Comparison of Solid-Phase Microextraction and Stir Bar Sorptive Extraction. *Journal of the Institute of Brewing*, 116(1), 81-85.
- Huimin, L., Hongjun, L., Xiuhua, L., & Bing, C. Q. (2012). Analysis of volatile flavor compounds in top fermented wheat beer by headspace sampling-gas chromatography. *International Journal of Agricultural and Biological Engineering*, 5(2), 67.
- Jackson, J. F., & Linskens, H. F. (2002). Testing for Taste and Flavour of Beer. In T. Yonezawa & T. Fushiki (Eds.), *Analysis of Taste and Aroma* first ed.). New York, USA: Springer.
- Jeleń, H. H., Wlazły, K., Waśowicz, E., & Kamiński, E. (1998). Solid-Phase Microextraction for the Analysis of Some Alcohols and Esters in Beer: Comparison with Static Headspace Method. *Journal of Agricultural and Food Chemistry*, 46(4), 1469-1473.
- Jiao, J., Ding, N., Shi, T., Chai, X., Cong, P., & Zhu, Z. (2011). Study of Chromatographic Fingerprint of the Flavor in Beer by HS-SPME-GC. *Analytical Letters*, 44(4), 648-655.
- Jones, B., & Nachtsheim, C. J. (2011). A Class of Three-Level Designs for Definitive Screening in the Presence of Second-Order Effects. *Journal of Quality Technology*, 43(1), 1-15.
- Jones, B., & Nachtsheim, C. J. (2013). Definitive Screening Designs with Added Two-Level Categorical Factors. *Journal of Quality Technology*, 45(2), 121-129.
- Kleinová, J., Geršl, M., & Mareček, J. (2015). Monitoring Volatile Substances in Beer in Relation to Beer Production Technology. *Journal of Advanced Agricultural Technologies*, 2(2), 134-137.
- Krogerus, K., & Gibson, B. R. (2013). Influence of valine and other amino acids on total diacetyl and 2,3-pentanedione levels during fermentation of brewer's wort. *Applied Microbiology and Biotechnology*, 97(15), 6919-6930.
- , W., & Manger, H. J. (2010). *Technology Brewing & Malting* (fourth ed.). Berlin, Germany: VLB Berlin.
- Leça, J. M., Pereira, A. C., Vieira, A. C., Reis, M. S., & Marques, J. C. (2015). Optimal design of experiments applied to headspace solid phase microextraction for the quantification of

- vicinal diketones in beer through gas chromatography-mass spectrometric detection. *Analytica Chimica Acta*, 887, 101-110.
- Li, H., Liu, F., Kun-Farkas, G., & Kiss, Z. (2015). Quantitative Analysis of Flavor Volatiles in Beer Using Headspace Solid-Phase Microextraction and Gas Chromatography-Flame Ionization Detection (HS-SPME-GC-FID). *Journal of the American Society of Brewing Chemists*, 73(3), 261-265.
- Malfliet, S., Van Opstaele, F., de Clippeleer, J., Syryn, E., Goiris, K., de Cooman, L., & Aerts, G. (2008). Flavour Instability of Pale Lager Beers: Determination of Analytical Markers in Relation to Sensory Ageing. *Journal of the Institute of Brewing*, 114(2), 180-192.
- Martins, C., Brandão, T., Almeida, A., & Rocha, S. M. (2015). Insights on beer volatile profile: Optimization of solid-phase microextraction procedure taking advantage of the comprehensive two-dimensional gas chromatography structured separation. *Journal of Separation Science*, 38(12), 2140-2148.
- Moreira, N., Meireles, S., Brandao, T., & de Pinho, P. G. (2013). Optimization of the HS-SPME-GC-IT/MS method using a central composite design for volatile carbonyl compounds determination in beers. *Talanta*, 117, 523-531.
- Nešpor, J., Karabín, M., Hanko, V., & Dostálek, P. (2018). Application of response surface design to optimise the chromatographic analysis of volatile compounds in beer. *Journal of the Institute of Brewing*, 124(3), 244-253.
- Nykänen, L., & Suomalainen, H. (1983). *Aroma of Beer, Wine and Distilled Alcoholic Beverages* (first ed.). Berlin, Germany: Akademie-Verlag.
- Pereira, A. C., Reis, M. S., Leça, J. M., Rodrigues, P. M., & Marques, J. C. (2018). Definitive Screening Designs and latent variable modelling for the optimization of solid phase microextraction (SPME): Case study - Quantification of volatile fatty acids in wines. *Chemometrics and Intelligent Laboratory Systems*, 179, 73-81.
- Phiarais, B. P. N., Mauch, A., Schehl, B. D., Zarnkow, M., Gastl, M., Herrmann, M., Zannini, E., & Arendt, E. K. (2010). Processing of a top fermented beer brewed from 100%

- buckwheat malt with sensory and analytical characterisation. *Journal of the Institute of Brewing*, 116(3), 265-274.
- Pinho, O., Ferreira, I., & Santos, L. (2006). Method optimization by solid-phase microextraction in combination with gas chromatography with mass spectrometry for analysis of beer volatile fraction. *Journal of Chromatography A*, 1121(2), 145-153.
- Pizarro, C., Pérez-del-Notario, N., & González-Sáiz, J. M. (2010). Optimisation of a simple and reliable method based on headspace solid-phase microextraction for the determination of volatile phenols in beer. *Journal of Chromatography A*, 1217(39), 6013-6021.
- Preedy, V. R. (2009). *Beer in Health and Disease Prevention* (first ed.). Oxford, UK: Elsevier Science.
- Richter, T. M., Eyres, G. T., Silcock, P., & Bremer, P. J. (2017). Comparison of four extraction methods for analysis of volatile hop-derived aroma compounds in beer. *Journal of Separation Science*, 40(22), 4366-4376.
- Riu-Aumatell, M., Miró, P., Serra-Cayuella, A., Buxaderas, S., & López-Tamames, E. (2014). Assessment of the aroma profiles of low-alcohol beers using HS-SPME-GC-MS. *Food Research International*, 57, 196-202.
- Rodrigues, F., Caldeira, M., & Câmara, J. S. (2008). Development of a dynamic headspace solid-phase microextraction procedure coupled to GC-qMSD for evaluation the chemical profile in alcoholic beverages. *Analytica Chimica Acta*, 609(1), 82-104.
- Rodriguez-Bencomo, J. J., Muñoz-González, C., Martín-Álvarez, P. J., Lázaro, E., Mancebo, R., Castañé, X., & Pozo-Bayón, M. A. (2012). Optimization of a HS-SPME-GC-MS Procedure for Beer Volatile Profiling Using Response Surface Methodology: Application to Follow Aroma Stability of Beers Under Different Storage Conditions. *Food Analytical Methods*, 5(6), 1386-1397.
- Rossi, S., Sileoni, V., Perretti, G., & Marconi, O. (2014). Characterization of the volatile profiles of beer using headspace solid-phase microextraction and gas chromatography-mass spectrometry. *Journal of the Science of Food and Agriculture*, 94(5), 919-928.

- Saerens, S. M. G., Delvaux, F. R., Verstrepen, K. J., & Thevelein, J. M. (2010). Production and biological function of volatile esters in *Saccharomyces cerevisiae*. *Microbial Biotechnology*, 3(2), 165-177.
- Saison, D., De Schutter, D. P., Delvaux, F., & Delvaux, F. R. (2008). Optimisation of a complete method for the analysis of volatiles involved in the flavour stability of beer by solid-phase microextraction in combination with gas chromatography and mass spectrometry. *Journal of Chromatography A*, 1190(1-2), 342-349.
- Santos, M. S. d., & Carasek, E. (2013). Development of a simple analytical method for determining trihalomethanes in beer using a headspace solid-phase microextraction technique. *Quimica Nova*, 36, 1052-1056.
- Scarlata, C. J., & Ebeler, S. E. (1999). Headspace Solid-Phase Microextraction for the Analysis of Dimethyl Sulfide in Beer. *Journal of Agricultural and Food Chemistry*, 47(7), 2505-2508.
- Scherer, R., Wagner, R., Kowalski, C. H., & Godoy, H. T. (2010). (E)-2-Nonenal determination in brazilian beers using headspace solid-phase microextraction and gas chromatographic coupled mass spectrometry (HS-SPME-GC-MS). *Journal of Food Science and Technology*, 30, 161-165.
- Stefanuto, P.-H., Perrault, K. A., Dubois, L. M., L'Homme, B., Allen, C., Loughnane, C., Ochiai, N., & Focant, J.-F. (2017). Advanced method optimization for volatile aroma profiling of beer using two-dimensional gas chromatography time-of-flight mass spectrometry. *Journal of Chromatography A*, 1507, 45-52.
- Svoboda, Z., Mikulíková, R., Běláková, S., & Benešová, K. (2017). Optimization of Determination of Dimethyl Sulfide in Wort and Beer. *Kvasny prumysl*, 63(3), 121-125.
- Svoboda, Z., Mikulíková, R., Běláková, S., Benešová, K., Marova, I., & Nesvadba, Z. (2010). Determination of Trans-2-Nonenal in Barley Grain, Malt and Beer. *Kvasny Prumysl*, 56(11/12), 428-432.

- Svoboda, Z., Mikulíková, R., Běláková, S., Benešová, K., Marova, I., & Nesvadba, Z. (2011). Optimization of Modern Analytical SPME and SPDE Methods for Determination of Trans -2-nonenal in Barley, Malt and Beer. *Chromatographia*, 73(1), 157-161.
- Techakriengkrai, I., Paterson, A., Taidi, B., & Piggott, J. (2006). Relationships of Overall Estery Aroma Character in Lagers with Volatile Headspace Congener Concentrations. *Journal of the Institute of Brewing*, 112(1), 41-49.
- Vanbeneden, N., Delvaux, F., & Delvaux, F. R. (2006). Determination of hydroxycinnamic acids and volatile phenols in wort and beer by isocratic high-performance liquid chromatography using electrochemical detection. *Journal of Chromatography A*, 1136(2), 237-242.
- Vanbeneden, N., Saison, D., Delvaux, F., & Delvaux, F. R. (2008). Decrease of 4-Vinylguaiacol during Beer Aging and Formation of Apocynol and Vanillin in Beer. *Journal of Agricultural and Food Chemistry*, 56(24), 11983-11988.
- Vesely, P., Lusk, L., Basarova, G., Seabrooks, J., & Ryder, D. (2003). Analysis of Aldehydes in Beer Using Solid-Phase Microextraction with On-Fiber Derivatization and Gas Chromatography/Mass Spectrometry. *Journal of Agricultural and Food Chemistry*, 51(24), 6941-6944.
- Xiao, L., Lin, D. K. J., & Fengshan, B. (2012). Constructing Definitive Screening Designs Using Conference Matrices. *Journal of Quality Technology*, 44, 1-7.

Figure Captions

Figure 1. Principal Component Analysis: loading vectors for PC1, PC2 and PC3.

Figure 2: Predicted versus Observed plot for the model developed to predict the VDKs, together with several figures of merit, including the R^2 , Predicted R^2 , RMSE and p-value for the ANOVA F-test.

Journal Pre-proofs

CRedit author statement

Ana C. Vieira – Investigation; Writing - Original Draft; Visualization

Ana C. Pereira – Conceptualization, Formal analysis, Resources; Writing - Original Draft; Writing - Review & Editing; Project administration; Funding acquisition.

José C. Marques – Resources, Writing - Review & Editing; Supervision; Project administration; Funding acquisition

Marco S. Reis – Methodology, Formal analysis, Resources; Writing - Review & Editing; Supervision

Highlights:

- Adoption of an optimal experimental design protocol to HS-SPME procedure
- Multi-targeted and global methodologies for establishing the optimal conditions
- Good results in terms of figures of merit of analytical methodology proposed
- Volatile organic compounds quantification in fermenting wort and beer samples

Table 1: Review of methodologies applied to beer VOC extraction by HS-SPME followed by GC analysis. The parameters that were optimized are highlighted in bold.

Main analytes	Optimization	De-gas	Sample/vial volume	Salt addition	Fiber	Stirring	Extraction temperature	Incubation time	Extraction time	Derivatization	Reference
beer volatiles	-	yes	5 mL	1.35g NaCl	PDMS/DVB	1200 rpm	50 °C	5 min	30 min	-	(G. da Silva, Augusto, & Poppi, 2008)
beer volatiles	-	-	10/20 mL	-	-	-	85 °C	-	30 min	-	(Huimin, Hongjun, Xiuhua, & Bing, 2012)
beer volatiles	-	-	15/20 mL	5g NaCl	PDMS	-	50 °C	10 min	30 min	-	(Kleinová, Geršl, & Mareček, 2015)
beer esters	-	-	10 mL	-	PA	-	37 °C	30 min	60 min	-	(Techakriengkrai, Paterson, Taidi, & Piggott, 2006)

beer volatiles	-	-	5/20 mL	-	PDMS/DVB	-	50 °C	26 min	2 min	yes	(Rossi, Sileoni, Perretti, & Marconi, 2014)
beer volatiles	-	yes	10/20 mL	-	PDMS/DVB	500 rpm	40 °C	-	10 min	-	(Jiao, Ding, Shi, Chai, Cong, & Zhu, 2011)
beer volatiles	-	-	5/20 mL	1.75g NaCl	DVB/CAR /PDMS	-	45 °C	20 min	40 min	-	(Riu-Aumatell, Miró, Serra-Cayuela, Buxaderas, & López-Tamames, 2014)
alcohols and esters	-	yes	10 mL	-	PA	-	60 °C	-	50 min	-	(Jeleń, Wlazły, Wałowicz, & Kamiński, 1998)
aldehydes	-	-	10/20 mL	-	PDMS	-	20 °C	-	20 min	yes	(Vesely, Lusk, Basarova, Seabrooks, & Ryder, 2003)
beer volatiles	-	yes	5g/20 mL	2g NaCl	DVB/CAR /PDMS	-	20 °C	5 min	10 min	-	(Giannetti, Boccacci Mariani, Torrelli, & Marini, 2019)
hop volatiles	-	-	8/20 mL	2.4g NaCl	DVB/CAR /PDMS	yes (incubation only)	40 °C	5 min	30 min	-	(Richter, Eyres, Silcock, & Bremer, 2017)
beer volatiles	-	yes	10/20 mL	2g NaCl	Car-PDMS	-	60 °C	-	20 min	-	(Stefanuto, Perrault, Dubois, L'Homme, Allen, Loughnane, et al., 2017)
beer volatiles	-	yes	10/20 mL	3g NaCl	DVB/CAR /PDMS	500 rpm	60 °C	10 min	30 min	-	(Alvim, Gomes, Garcia, Vieira, & Machado, 2017)
beer volatiles	OFAT	-	10/20 mL	3.5g NaCl	DVB/CAR /PDMS	500/250 rpm	40 °C	10 min	30 min	-	(Saison, De Schutter, Delvaux, & Delvaux, 2008)
beer volatiles	OFAT	-	10/40 mL	3g NaCl	DVB/CAR /PDMS	-	60 °C	60 min (70 °C)	30 min	-	(G. C. da Silva, da Silva, da Silva, Godoy, Nogueira, Quiterio, et al., 2015)
beer volatiles	OFAT	-	30/60 mL	9g NaCl	DVB/CAR /PDMS	-	30 °C	-	60 min	-	(Rodrigues, Caldeira, & Câmara, 2008)
beer volatiles	OFAT	yes	2 mL beer + 2 ml water /10 mL	1.7g NaCl	DVB/CAR /PDMS	500 rpm	30 °C	5 min	5 min	-	(Cajka, Riddellova, Tomaniova, & Hajslova, 2010)
beer volatiles	OFAT	yes	5g/15 mL	2g NaCl	Car/PDMS	yes	20 °C	30 min	30 min	-	(Pinho, Ferreira, & Santos, 2006)
wort volatiles	OFAT	-	10/20 mL	3.5g NaCl	DVB/CAR /PDMS	250 rpm	45 °C	10 min	30 min	-	(De Schutter, Saison, Delvaux, Derdelinckx, Rock, Neven, et al., 2008)
beer volatiles	OFAT	-	5/15 mL	2g NaCl	PDMS	870 rpm	24 °C	-	45 min	-	(Charry-Parra, Dejesus-Echevarria, & Perez, 2011)

beer volatiles	OFAT	-	10/20 mL	3g NaCl	PA	-	40 °C	-	30 min	-	(Li, Liu, Kun-Farkas, & Kiss, 2015)
beer volatiles	OFAT	yes	10/20 mL	2g NaCl	PDMS/DVB	400 rpm	40 °C	10 min	30 min	-	(Martins, Brandão, Almeida, & Rocha, 2015)
beer volatiles	OFAT	yes	15/30 mL	2g NaCl	DVB/CAR/PDMS	-	40 °C	10 min	40 min	-	(Biazon, Brambilla, Rigacci, Pizzolato, & dos Santos, 2009)
volatile phenols	OFAT	-	6 mL/20 ml	2.4g NaCl	DVB/CAR/PDMS	250 rpm	80 °C	5 min	55 min	-	(Pizarro, Pérez-del-Notario, & González-Sáiz, 2010)
sulphur compounds	OFAT	-	10/15 mL	-	Car/PDMS	-	45 °C	-	32 min	-	(Hill & Smith, 2000)
trans-2-nonenal	OFAT	-	5/20 mL	1.5g NaCl	PDMS/DVB	yes	60 °C	-	20 min	-	(Svoboda, Mikulíková, Běláková, Benešová, Marova, & Nesvadba, 2011)
trans-2-nonenal	OFAT	yes	10/23 mL	-	Car/PDMS	yes	50 °C	15 min	90 min	-	(Scherer, Wagner, Kowalski, & Godoy, 2010)
phthalates	OFAT	-	4/15 mL	-	PDMS/DVB	250 rpm	95 °C	-	100 min	-	(Carnol, Schummer, & Moris, 2017)
beer esters	OFAT	-	10/20 mL	5g NaCl	DVB/CAR/PDMS	800 rpm	40 °C	-	60 min	-	(Horak, Culik, Kellner, Jurková, Čejka, Hašková, et al., 2010)
beer trihalomethanes	OFAT	yes	20/40 mL	4g NaCl	Car/PDMS	1000 rpm	30 °C	8 min	15 min	-	(Santos & Carasek, 2013)
carbonyl compounds	CCD	-	5/20 mL	-	PDMS/DVB	250 rpm	45 °C	7 min	20 min	yes	(Moreira, Meireles, Brandao, & de Pinho, 2013)
vicinal diketones	O-DOE	-	5/20 mL	-	Car/PDMS	yes	30 °C	5 min	25 min	-	(Leça, Pereira, Vieira, Reis, & Marques, 2015)
beer volatiles	FFD	-	8/10 mL	3g NaCl	IL-1 butenyl	900 rpm	35 °C	10 min	15 min	-	(González-Álvarez, Blanco-Gomis, Arias-Abrodo, Pello-Palma, Ríos-Lombardía, Busto, et al., 2013)
beer volatiles	CCD	-	6/20 mL	1.8g NaCl	DVB/Car/PDMS	-	44.8 °C	10 min	46.8 min	-	(Rodríguez-Bencomo, Muñoz-González, Martín-Álvarez, Lázaro, Mancebo, Castañé, et al., 2012)
beer volatile	CCD	yes	10/20	2.6g	Car-PDMS	-	50 °C	-	45	-	(Nešpor, Karabín, Hanko, & Dostálek,

Table 2. Experimental domain explored to optimize the HS-SPME extraction performance.

Factor	Qualitative/Quantitative	Levels
Type of Fiber	Qualitative	L1 - PDMS/DVB; L2 -CAR/PDMS; L3 - CAR/PDMS/DVB
Degass	Qualitative	L1 - No; L2 - Yes (15min)
Sample volume	Quantitative	L1 - 5 mL; L2 - 10 mL
Addition of salt	Qualitative	L1 - No, L2 - Yes
Agitation	Qualitative	L1 - No; L2 - Yes
Pre-incubation time (min)	Quantitative	L1 - 0; L2 - 5; L3 - 10
Extraction time (min)	Quantitative	L1 - 20; L2 - 30; L3 - 40
Extraction temperature (°C)	Quantitative	L1 - 40; L2 - 50; L3 - 70

Table 3. Optimal levels for the experimental factors optimizing HS-SPME performance.

VDKs	Acetaldehyde	Acetoin	2-3 Butanediol (R,S)	Majority Esters	Minority Esters	Alcohols	Fatty Acids
p-value=0.0028 R ² =0.99	p-value =0.0158 R ² =0.59	p-value =0.0069 R ² =0.80	p-value =0.0211 R ² =0.56	p-value =0.0044 R ² =0.68	p-value =0.001 R ² =0.60	p-value =0.0116 R ² =0.61	p-value =0,001 R ² =0,70
No		No		No	No	No	No
10 mL		10 ml					
Yes (3.3g NaCl)	Yes		Yes	Yes	No	Yes	No
Car-PDMS	Car-PDMS	Car-PDMS	Car-PDMS	Car-PDMS	Car-PDMS	Car-PDMS	Car-PDMS
No		Yes	Yes	No	No		Yes
0 min				10 min			
20 min	40 min	40 min			40 min		20 min
40 °C	70 °C	40 °C	70 °C	40 °C	70 °C	70 °C	70 °C

Journal Pre-proofs

Table 4. Figures of merit for the HS-SPME/GC-MS methodology to quantify VOC in samples ranging from wort to beer (* indicates compounds not considered in DSD analysis).

t_R (minutes)	Analyte	Kovats index	Identification/quantification mode (m/z)	Concentration range in literature	Matrix effect (%)	Linear range	R ²	LOD	LOQ
4.66	Acetaldehyde (mg/L)	676	29, 42, 43 , 44, 45	1.7-40 (Brányik, Vicente, Dostálek, & Teixeira, 2008; Charry-Parra, Dejesus-Echevarria, & Perez, 2011; Kunze & Manger, 2010; Preedy, 2009)	27.21	1.26-150.89	0.999997	0.35	1.18
5.04	Dimethyl sulfide (µg/L)*	740	47, 61, 62	2-215 (Preedy, 2009; Scarlata & Ebeler, 1999; Svoboda, Mikulíková, Běláková, & Benešová, 2017)	11.02	1.00-112.31	0.999999	0.20	0.68
5.63	Isobutyraldehyde (µg/L)*	810	41, 43, 72	3.8-146.8 (Malfliet, Van Opstaele, de Clippeleer, Syryn, Goiris, de Cooman, et al., 2008; Moreira, Meireles, Brandao, & de Pinho, 2013; Saison, De Schutter, Delvaux, & Delvaux, 2008)	4.46	1.00-100.02	0.999997	0.23	0.77
6.71	Ethyl acetate (mg/L)	887	43 , 61, 70, 88	0.29-60.9 (Jeleń, Wlazły, Wałowicz, & Kamiński, 1998; Kleinová, Geršl, & Mareček, 2015; Kunze & Manger, 2010)	15.94	4.99-62.46	0.999921	729.24	2.43
8.67	Diacetyl (µg/L)	976	42, 43, 86	8-1180 (Brányik, Vicente, Dostálek, & Teixeira, 2008; Briggs, Hough, Stevens, & Young, 1999; Kunze & Manger, 2010)	2.34	10.00-999.78	0.999997	2.72	9.06
9.58	Isobutyl acetate (µg/L)	1014	43 , 56, 73	30-10120 (Jeleń, Wlazły, Wałowicz, & Kamiński, 1998; Kunze & Manger, 2010; Nykänen & Suomalainen, 1983)	13.10	100.05-2501.15	0.999899	33.54	111.79
10.24	Ethyl butyrate (µg/L)	1044	60, 71 , 88	40-300 (Boulton & Quain, 2001; Kunze & Manger, 2010; Phiarais, Mauch, Seehl, Zarnkow, Gastl, Herrmann, et al., 2010)	15.73	20.30-406.00	0.999824	7.04	23.46
10.94	2,3-pentanedione (µg/L)	1074	43, 45, 57, 100	10-900 (Briggs, Hough, Stevens, & Young, 1999; Krogerus & Gibson, 2013; Leça, Pereira, Vieira, Reis, & Marques, 2015)	17.43	25.05-500.56	0.999901	6.89	22.95
11.15	Ethyl isovalerate (µg/L)	1083	57, 85, 88	Not available; it was established by determining its concentration in wort, fermenting wort and beer samples by external addition method	0.76	1.00-25.06	0.999989	0.11	0.36
12.12	Isobutyl alcohol (mg/L)	1123	41, 42, 43 , 55, 74	0.3-59.92 (Charry-Parra, Dejesus-Echevarria, & Perez, 2011; Cheong, Wackerbauer, & Kang, 2007; Jeleń, Wlazły, Wałowicz, & Kamiński, 1998; Kunze & Manger, 2010)	3.46	1.00-49.86	0.999990	0.23	0.76
12.96	Isoamyl acetate (mg/L)	1157	61, 70 , 87	0.08-6.6 (Charry-Parra, Dejesus-Echevarria, & Perez, 2011; Kunze & Manger, 2010; Saison, De Schutter, Delvaux, & Delvaux, 2008)	0.77	0.20-5.00	0.999967	0.04	0.13
15.86	Isoamyl alcohol (mg/L)	1249	42, 55 , 70	2.5-70 (Charry-Parra, Dejesus-Echevarria, & Perez, 2011; Kunze & Manger, 2010; Preedy, 2009)	3.52	1.25-149.99	0.999997	0.37	1.24

16.67	Ethyl hexanoate ($\mu\text{g/L}$)	1270	60, 88 , 99	50-1500 (Jeleń, Wlazły, Wařowicz, & Kamiński, 1998; Kunze & Manger, 2010; Preedy, 2009; Saerens, Delvaux, Verstrepen, & Thevelein, 2010)	7.88	20.03-500.83	0.999961	4.38	14.58
18.69	Acetoin (mg/L)	1323	43, 45 , 88	1-86 (Baxter, Hughes, & Hornsey, 2001; Kunze & Manger, 2010)	32.35	1.01-75.60	0.999991	0.33	1.09
23.44	Ethyl caprylate ($\mu\text{g/L}$)	1429	88 , 101, 127	40-4000 (Preedy, 2009; Saerens, Delvaux, Verstrepen, & Thevelein, 2010)	13.60	199.99-4999.68	0.999994	18.72	62.41
26.73	Trans-2-nonenal ($\mu\text{g/L}$)	1477	55 , 70, 83	0.28-20.28 (Moreira, Meireles, Brandao, & de Pinho, 2013; Svoboda, Mikulíková, Běláková, Benešová, Marova, & Nesvadba, 2010)	20.75	0.50-62.60	0.999999	0.11	0.37
29.74	Ethyl decanoate ($\mu\text{g/L}$)	1552	88 , 101, 155	20-2550 (Hiralal, Pillay, & Olaniran, 2013; Horak, et al., 2010)	2.12	99.99-749.94	0.999646	20.42	68.05
31.38	Isovaleric acid ($\mu\text{g/L}$)	1612	41, 60 , 87	100-3400 (Cheong, Wackerbauer, & Kang, 2007; Jackson & Linskens, 2002; Kunze & Manger, 2010)	24.20	199.98-2499.74	0.999893	32.90	109.68
35.00	Phenylethyl acetate ($\mu\text{g/L}$)	1731	43, 91, 104	100-6790 (Hiralal, Olaniran, & Pillay, 2014; Saerens, Delvaux, Verstrepen, & Thevelein, 2010)	43.54	96.18-4809.12	0.999993	18.57	61.91
37.60	Phenylethyl alcohol (mg/L)	1828	91 , 92, 122	2.76-138.11 (Kunze & Manger, 2010; Preedy, 2009; Riu-Aumatell, Miró, Serra-Cayuela, Buxaderas, & López-Tamames, 2014)	16.21	1.25-149.94	0.999993	0.56	1.87
41.55	Octanoic acid ($\mu\text{g/L}$)	1983	60 , 73, 101	2000-12000 (Cheong, Wackerbauer, & Kang, 2007; Kunze & Manger, 2010)	9.22	350.35-10010.00	0.999979	65.59	218.65
44.56	Vinyl guaiacol ($\mu\text{g/L}$)	2109	107, 135, 150	7-1112 (Pizarro, Pérez-del-Notario, & González-Sáiz, 2010; Vanbeneden, Delvaux, & Delvaux, 2006; Vanbeneden, Saison, Delvaux, & Delvaux, 2008)	16.44	11.12-1112.00	0.999999	1.97	6.57

