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Artemisia umbelliformis Lam. and Génépi Liqueur: Volatile Profile as Diagnostic Marker for **Geographic Origin and to Predict Liqueur Safety**

This is the author's manuscript

Original Citation:

Availability:

This version is available http://hdl.handle.net/2318/1633099 since 2017-05-16T15:44:37Z

Published version:

DOI:10.1021/acs.jafc.6b03394

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This is the author's final version of the contribution published as:

Boggia, Lorenzo; Pignata, Giuseppe; Sgorbini, Barbara; Colombo, Maria Laura; Marengo, Arianna; Casale, Manuela; Nicola, Silvana; Bicchi, Carlo; Rubiolo, Patrizia. Artemisia umbelliformis Lam. and Génépi Liqueur: Volatile Profile as Diagnostic Marker for Geographic Origin and to Predict Liqueur Safety. JOURNAL OF AGRICULTURAL AND FOOD CHEMISTRY. 65 (13) pp: 2849-2856. DOI: 10.1021/acs.jafc.6b03394

The publisher's version is available at: http://pubs.acs.org/doi/pdf/10.1021/acs.jafc.6b03394

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1	Artemisia umbelliformis Lam. and génépi liqueur: volatile profile as diagnostic marker for				
2	geographic origin and to predict liqueur safety				
3	Lorenzo Boggia [†] , Giuseppe Pignata [§] , Barbara Sgorbini [†] , Maria Laura Colombo [†] , Arianna				
4	Marengo [‡] , Manuela Casale [§] , Silvana Nicola [§] , Carlo Bicchi [†] , Patrizia Rubiolo [†] *				
5	[†] Department of Drug Science and Technology (DSTF), University of Turin, Via Pietro Giuria 9				
6	10125 Torino, Italy				
7	§ Vegetable Crops & Medicinal and Aromatic Plants, VEGMAP; Department of Agricultural,				
8	Forest and Food Sciences, DISAFA; University of Turin. Via Leonardo da Vinci, 44 - Largo Paolo				
9	Braccini, 2, 10095 Grugliasco (TO), Italy				
10	[‡] Department of Life and Environmental Sciences, University of Cagliari, Viale S. Ignazio da				
11	Laconi 13, 09124 Cagliari, Italy				
12					
13	* Phone: +39 011 6707662. Fax: +39 011 2367661. E-mail: patrizia.rubiolo@unito.it.				

15 ABSTRACT

A. umbelliformis, commonly known as "white génépi", is characterized by a volatile fraction rich in α - and β - thujone, two monoterpenoids; under EU regulations these are limited to 35 mg/L in *Artemisia*-based beverages , because of their recognized activity on human central nervous system. 19 This study reports the results of an investigation to define the geographical origin and thujone

content of individual plants of *A. umbelliformis* from different geographical sites, cultivated experimentally at a single site, and to predict the thujone content in the resulting liqueurs, through their volatile fraction. Headspace Solid Phase Microextraction (HS-SPME) combined with Gas Chromatography-Mass Spectrometry (GC-MS) and non-separative HS-SPME-MS were used as analytical platforms to create a database suitable for chemometric description and prediction through Linear Discriminant Analysis (LDA). HS-SPME-MS was applied to shorten analysis time. With both approaches, a diagnostic prediction of: i) plant geographical origin, and ii) thujone

27 content of plant-related liqueurs could be made.

28

29 Keywords: Artemisia umbelliformis Lam., génépi liqueur, food safety

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33 INTRODUCTION

34 *Artemisia* L., with its 500 specific and sub-specific *taxa*, is the largest genus in the tribe *Anthemidae* 35 of the Asteraceae family^{1,2}. Its *taxa* grow worldwide at different latitudes, altitudes, and 36 environments, and have traditionally been used for their biological activity¹, even before recent 37 pharmacological developments^{3–5}.

Artemisia species were used historically to prepare infusions, wines, and liqueurs⁶. The best-known 38 39 species is A. absinthium, traditionally known as wormwood and in widespread use since Roman times as the base for aromatic wines and liqueurs^{6,7}. Wormwood has been studied in depth because 40 of its diffusion and ready accessibility⁶, but several other Artemisia species are also used in 41 beverages. A group of herbs used for liqueur production, traditionally known as "génépi", consists 42 43 of five rare species growing at high altitudes in the alpine area: A. eriantha Ten., A. genipi Weber, A. glacialis L., A. nivalis Br.-Bl. and A. umbelliformis Lam.. Génépi species traditionally have less 44 45 economic importance than wormwood, being almost exclusively used within their production and harvesting areas; however, over the last 40 years, liqueur production have markedly increased, not 46 least thanks to a number of small and medium-sized producers⁶. Génépi species have historically 47 been used in folk and traditional medicine because of their bioactivity⁶. They were known and used 48 49 as thermogenic agents against the common cold, in infusions against fever, and in aromatic wines and liqueurs, to stimulate the appetite and the digestion,⁸⁻¹⁰ respectively. The extensive ethno-50 51 pharmacological knowledge, documented from the second half of the eighteenth century, contrasts with the paucity of scientific studies of génépi species⁶. 52

Liqueur production on a larger scale is not possible if the raw material is limited to the wild plant harvest. Génépi cultivation, which is restricted by the plants' pedoclimatic needs, facilitates harvesting operations and protects wild génépi species ⁶. The five génépi species have a common environmental distribution, growing at around 1500-2000 m a.s.l.¹¹. The altitude requirement is due to specific soils in which these species face neither competition from weeds nor fungal pathogens, which become predominant at lower altitudes^{6,12}. 59 A. umbelliformis, commonly known as "white génépi", is the génépi species that has been cultivated more successfully, above 1600 m a.s.l.^{6,11,12}; this species has therefore been selected for wider-scale 60 61 cultivation. Independently of its domestication and production, the supply chain of A. umbelliformis 62 faces two issues linked to the food industry: product adulteration and product safety. The need for 63 an authentication process emerges, since it is one of the few Italian infusion liqueurs whose geographic limits are defined in the EU by Reg. EC 110/2008¹³. This regulation introduced three 64 categories for génépi, distinguishing the most important production areas situated in North-West 65 Italy and France: "Genepì della Valle d'Aosta (Italy)", "Genepì del Piemonte (Italy)", and "Génépi 66 67 des Alpes/Genepì delle Alpi (France, Italy)".

68 Thujones are natural monoterpenoids widely present in *Artemisia* species. α - and β - thujone are the most abundant volatiles in génépi species¹⁴⁻¹⁶, and are characterized by recognized activity on the 69 human central nervous system^{11,17,18}. For this reason, the maximum content of thujone in Artemisia-70 based beverages is limited to 35 mg/L in the EU (Reg. EC 110/2008^{6,13,18}). A. umbelliformis is the 71 génépi species with the highest content of thujones¹⁹. The variability of the plant's chemical 72 73 composition (i.e. the variability of thujone content) is a crucial point for génépi liqueur producers. 74 To overcome this problem, some agronomical selection has been done to create thujone-free chemotypes^{11,18}. However, chemical comparisons between thujone-free and thujone-containing 75 76 chemotypes have found a marked difference in the content of both aromatic and bitter compounds, responsible for the typical sensory characteristics of the liqueur¹⁸. 77

Chemical analysis combined with chemometrics can be useful to deal with the problems emerging from large-scale production^{20–25}. In the present study, Linear Discriminant Analysis (LDA)²⁶ was adopted for its ability to perform class discrimination, instead of Principal Component Analysis (PCA), which is widely used to describe large datasets. LDA is known to elucidate the underlying data structure, a key point when creating a prediction model ^{24,27}. Some PCA elaborations were, however, carried out and a comparison of the results to those of LDA confirm that the latter technique has a better prediction capacity (data not shown). This research studied *A. umbelliformis* of different geographical origins, cultivated experimentally at a single site. Headspace sampling with Solid Phase Microextraction, combined with gas chromatography - mass spectrometry (HS-SPME-GC-MS) or directly with mass spectrometry (HS-SPME-MS), was used as an analytical platform to create the first database suitable for chemometric description and prediction. Linear Discriminant Analysis (LDA) was here adopted to define and predict geographic origin and thujone content of plants and liqueurs

Multiple Headspace- Solid Phase Microextraction (MHS-SPME)^{26,28,29} was applied for thujone
 content determination.

93

94 MATERIAL AND METHODS

95 Chemicals. Thujones standard mixture (mixture of a-thujone and β-thujone, purity = 99.9%) and all
96 other reference compounds were from Sigma-Aldrich (St. Louis, MO). Sabinol was kindly supplied
97 by Robertet SA (Grasse, France), and sabinyl ester homologous series were synthesized in the
98 authors' laboratory¹⁸. HPLC and analytical grade solvents were from Carlo Erba Reagenti, Rodano,
99 Italy.

Plant material and experimental site preparation. Growing fields of *A. umbelliformis* were monitored in spring and summer at various alpine sites in Piedmont in North-West Italy (**Table 1**). The sites are located in the South-West Alps within the area indicated for the liqueur "Genepi del Piemonte"¹³. At the end of the blossoming stage, four *A. umbelliformis* geographic origins were selected (**Table 1**). Plant growing on the verge of the raised beds were excluded, to avoid the edge effect. Mother plants were chosen based on the phenological profile (data not shown) and then carefully harvested to preserve inflorescences and seeds.

Plants were dried at 30°C in a dry and aerated room at the Experimental Centre of DISAFA (44°53'11.67"N; 7°41'7.00"E - 231 m a.s.l.) in Tetti Frati, Carmagnola (TO), Italy. The seeds were cleaned to remove extraneous parts, under fan ventilation using a series of sieves of different meshes, by Semina S.r.l. (La Loggia (TO), Italy), and then stored at 6°C in a seed room. The seeds

111 of the four different geographic groups were sown the following summer in trays of a specific peatbased horticultural medium (Neuhaus Huminsubstrat N17; Klasmann-Deilmann[®] GmbH, Geeste, 112 113 Niedersachsen, Germany). The trays were placed in an air-conditioned greenhouse providing the necessary supply of water until the seeds germinated. After cotyledon expansion, A. umbelliformis 114 115 was transplanted into 104-cell plastic trays $(0.522m \times 0.322m \times 0.042m; 0.034m \text{ upper diameter})$ 116 cells), filled with the above peat medium. Travs were placed in an air-conditioned plastic 117 greenhouse and seedlings were irrigated overhead as required. After secondary leaf development, 118 trays were placed under a plastic film tunnel covered with black shading systems (50%-shade 119 cloth). Before transplanting the A. umbelliformis seedlings into the soil, the shade cloth was 120 removed to harden the plants and acclimatize them to outdoor environmental conditions. The count 121 of dead plants was performed weekly (data not shown). Plants were transplanted in the autumn at the experimental site of Pragelato (TO), Italy (45°00'36.41"N; 6°56'14.56"E - 1,500 m a.s.l). The 122 123 site is characterized by a soil with a moderate presence of skeleton, absence of slope, and optimal 124 sun exposition. After minimum tillage, a white/black (upper/lower side) plastic mulching film 125 provided with holes arranged in a quincunx formation (0.05 m hole diameter, 0.20 m distance 126 between holes) was used to set 6 mulched raised beds subdivided into 12 plots (4 geographic groups 127 (Gran Paradiso, GP; Val Chisone; Elva; Valle Gesso, VG) ×3 blocks) for a total of ca. 1,350 A. 128 *umbelliformis* plants.

Production, phenological and morphological parameters. The site was inspected to check *A*. *umbelliformis* stand establishment with a \approx 30 days frequency. The number of dead plants was monitored and recorded 30 days after transplanting, the following summer and autumn (1st year), and again the subsequent summer (2nd year). The percentage of dead plants was calculated and used as an indicator of stand establishment and the plants' adaptation to the site.

Phenological and morphological parameters were measured in the 1st summer and autumn and in the 2nd summer after transplanting, on 12 plants per geographic origin and per block, randomly chosen at the beginning of the experiment. Inflorescence parameters were only measured during the 137 2nd summer of experimentation. The phenological parameters measured on the canopy were: canopy 138 diameter, canopy height (**Figure 1**A); on the inflorescences: inflorescence height, inflorescence 139 number (**Figure 1**B). The morphological parameters measured on the canopy leaves were: number 140 of segments, leaf length, leaf segment length and width (**Figure 1**C); on the inflorescence leaves: 141 cauline leaf length, cauline leaf petiole length (**Figure 1**D); on the inflorescences: footstalk length, 142 number of glomerules (at first, second and third internode), number of plant glomerules, number of 143 head glomerules, total number of glomerules (**Figure 1**E).

Harvesting took place in the 2nd summer after transplanting, during the blossoming stage. Fresh
weight production per plant was measured, and weight after drying at 30°C until constant weight,
dry matter percentage was calculated as in the European Pharmacopoeia³⁰.

Headspace Solid Phase Microextraction (HS-SPME) sampling and analysis conditions: qualitative profiling of plants.

149 **Separative analysis.** The 104 selected plants surviving at the end of the 2nd summer were used to 150 analyze the volatile fraction. Portions of 25 mg were sampled, as the minimum plant weight 151 required to have both an adequate informative profile and good repeatability with headspace solid-152 phase microextraction (HS-SPME).

Each plant was sampled in a 20 mL headspace vial (2 replicates of 25 mg) using a 2 cm DVB/CAR/PDMS fiber (Supelco, Bellefonte, PA, USA). Sampling was carried out with a MPS2 automatic sampling system (Gerstel, Mülheim a/d Ruhr, Germany) at T=50 °C (15 min of preequilibrium, 15 min of sampling). The fiber was automatically transferred to the GC injector of a 7890 GC coupled to a 5975C MS (Agilent, Little Falls, DE, USA) from where the sampled analytes were recovered directly by thermal desorption and transferred to the GC column for analysis.

- 159 GC-MS conditions. Inlet T=250°C, split injection (5 min, 1/10 split ratio); Helium was used as
- 160 carrier gas at a flow rate of 1 mL/min. Column: MEGA5MS (30 m \times 0.25 mm i.d. \times 0.25 μ m;
- 161 MEGA, Legnano, MI, Italy). Temperature program: 50 °C (1 min)-3 °C/min-250 °C (5 min). MS
- 162 operated in EI mode at 70 eV with a mass range from 35 to 350 amu in full scan mode.

163 Data were processed with Agilent MSD ChemStation D.03.00.611 (Agilent Technologies). 164 Components were identified by comparing their linear retention indices ($I^{T}s$) (calculated *versus* a 165 C₉-C₂₅ hydrocarbon mixture) and their mass spectra to those of authentic standards.

The HS-SPME-GC-MS method repeatability was tested on ten samples in five replicates for the 30 selected compounds, reaching an overall value of relative standard deviation (RSD%) of 9.3%. As reported in **Supplementary Table 1**, the minimum RSD% was detected with sabinyl valerianate (2.5%) and the maximum RSD% with sabinene (29.3%).

Non-separative analysis. The same plant samples analyzed by HS-SPME-GC-MS were also analyzed without performing GC separation. Sampling was carried out on 2 replicates of 150 mg of each plant weighted in a 20 mL headspace vial using a 2 cm DVB/CAR/PDMS fiber (Supelco, Bellefonte, PA, USA) at 100 °C for 5 min. Sampling and injection were managed by a Gerstel MPS2 automatic system (Gerstel, Mülheim a/d Ruhr, Germany). The fiber was automatically transferred to the GC injector of a 7890 GC coupled to a 5975C MS (Agilent, Little Falls, DE, USA) and the sampled analytes thermally desorbed directly into the empty tubing for MS analysis.

177 MS analysis conditions. Inlet T:250°C, split injection (5 min, 1/20 split ratio); carrier gas: helium,

178 flow rate: 1 mL/min. MS transfer column: deactivated fused silica empty tubing (8 m \times 0.25 mm

i.d.; MEGA, Legnano, MI, Italy). Oven temperature: 250 °C (5 min). MS operated in EI mode at 70

180 eV with a mass range from 35 to 250 amu in full scan mode.

MS data were processed with Agilent MSD ChemStation D.03.00.611 (Agilent Technologies) and
Pirouette 4.0.2 (Infometrix Inc., Bothell, WA, USA).

The HS-SPME-MS method repeatability was tested on ten samples in five replicates for the 30 most abundant fragment ions, reaching an overall relative standard deviation (RSD%) of 1.8%. As reported in **Supplementary Table 2**, the minimum RSD% was found for m/z=43 (0.6%) and the maximum for m/z=107 (2.5%). 187 Multiple headspace solid phase microextraction (MHS-SPME-GC-MS): quantitation of 188 thujones in plants and liqueurs. Thirty six samples of *A. umbelliformis* among the 104 samples 189 employed in the previous steps were selected to produce experimental liqueurs. For each plant, one 190 individual liqueur was prepared, following the guidelines for liqueur preparation given in Annex II 191 (32) of EC Regulation No $110/2008^{13}$. MHS-SPME was then applied to the liqueurs (three 192 replicates of 10 µL for each sample introduced in a 20 mL headspace vial, three HS-SPME 193 samplings for each replicate).

Beside liqueur analysis, each plant sample was also analyzed by MHS-SPME-GC-MS (three replicates of 3 mg for each sample weighted in a 20 mL headspace vial, three HS-SPME samplings for each replicate).

197 Quantitative calibration was carried out by MHS-SPME using seven concentration levels of α - and 198 β -thujone standard mixture (92.6% α -thujone, 7.4% β -thujone) between 0.1 g/L and 10 g/L in 199 cyclohexane (two replicates of 10 μ L for each level, three HS-SPME samplings for each 200 replicate)³¹.

201 Sampling conditions. MHS-SPME analyses carried out using 2 were а cm 202 divinylbenzene/carboxen/polydimethylsiloxane (DVB/CAR/PDMS) fiber (Supelco, Bellefonte, PA, 203 USA) at 60 °C for 30 min. Sampling and injection were managed by a Gerstel Multipurpose 204 Sampler 2 (MPS2) (Gerstel, Mülheim a/d Ruhr, Germany). The fiber was automatically transferred 205 to a 7890 GC unit coupled to a 5975C MS to analyze the sampled analytes (Agilent, Little Falls, 206 DE, USA). During MHS-SPME sampling, the sample was not heated between sampling, but placed 207 back into the sample tray after each sampling.

 $GC-MS \ conditions. \ Inlet T:250^{\circ}C, \ split injection (5 min, 1/20 \ split ratio); \ carrier \ gas: \ helium, \ flow$ $rate: 1 \ mL/min. \ Column: \ MEGA-SE52 \ (30 \ m \times 0.25 \ mm \ i.d. \times 0.25 \ \mum; \ Mega, \ Legnano, \ MI,$ $Italy). \ Temperature \ program: \ 50^{\circ}C-3^{\circ}C/min-105^{\circ}C-20^{\circ}C/min-250^{\circ}C \ (1 \ min). \ MS \ operated \ in \ EI$ $mode \ at \ 70 \ eV \ with \ a \ mass \ range \ from \ 35 \ to \ 350 \ amu \ in \ full \ scan \ mode. \ Data \ were \ processed \ with$ $Agilent \ MSD \ ChemStation \ D.03.00.611 \ (Agilent \ Technologies).$

Statistical data treatment. Preliminary descriptive, significance analyses of variance (ANOVA)
and linear regression attempts were run on MHS-SPME-GC-MS data. Linear Discriminant Analysis
(LDA)^{27,32} was applied to correlate MHS-SPME-GC-MS data with HS-SPME-GC-MS data and
HS-SPME-MS data. Statistical analyses were performed by using SPSS 15.0 (IBM Corporation) or
R³³; RStudio³⁴ was used to write a LDA cycling script. Other R packages used were: caret³⁵,
MASS³⁶ and e1071³⁷.

219 LDA describes and predicts multivariate data matrices in which samples are described by one 220 categorical variable and *n* continuous variables. The LDA model calculates Discriminant Functions 221 (DFs) which are used first to describe the group distribution in the analyzed dataset and then to 222 predict the groups of samples. DFs are calculated on the basis of the relationships between the 223 continuous variables describing the dataset and the categorical variables defining the group of each 224 sample. DFs can be considered similar to the Principal Components in PCA, but they differ for 225 some characteristics, such as their meaning (they maximize distances between the groups of the dataset instead of maximizing the dataset explained variance), and for their number (PC number = 226 variable number; DF number = group number minus one)²⁷. 227

LDA was carried out on four different matrices based on different data: i) plant geographic origin as categorical variable and plant morphological and phenological data as continuous variables; ii) plant geographic origin as categorical variable and plant HS-SPME-GC-MS compound abundances as continuous variables; iii) liqueur MHS-SPME-GC-MS thujone level as categorical variable and plant HS-SPME-GC-MS compound abundances as continuous variables; iv) liqueur MHS-SPME-GC-MS thujones level as categorical variable and plant HS-SPME-MS fragment ions abundances as continuous variables.

235

236 **RESULTS AND DISCUSSION**

The study comprised three main steps: i) to describe statistically-significant differences relating to the geographical distribution of cultivated *A. umbelliformis* plants; ii) to develop and apply analytical methods to rapidly quantify thujone content of A. umbelliformis-based liqueur (génépi);

240 iii) to create multivariate statistical models to predict the above two points.

241 The morphological/phenological dataset and the chemical dataset were processed with LDA.

Origin discrimination through phenological and morphological data. At the end of the 2nd year of field tests, 61.8% of the *A. umbelliformis* plants survived without showing significant differences between plants from different geographical origins (data not shown). Morphological data collected in the 2nd year included both vegetative and reproductive parts of the 104 plant samples (**Figure 1**,

246 **Supplementary Table 3**).

ANOVA was carried out, when possible (p < 0.05), together with the Tukey Honestly Significant Difference (HSD) test. The measurements related to the length of footstalks and to the number of glomerules were found to be the most significant parameters in differentiating plant individuals originating from different geographical areas (**Supplementary Table 4**).

LDA was used to describe the morphological differentiation of the plant groups and was carried out in two steps: cross validation (CV) and prediction. The dataset was divided into two subsets. Approximately 60% of the samples were used to create the prediction model based on three Discriminant Functions (DF). Subsequently, the residual 40% of the samples were predicted by the three Discriminant Functions, in order to calculate the predictive power of the LDA model.

LDA was carried out considering phenological and morphological parameters, and the three DFs explain the total variance (DF1 = 51.0%; DF2 = 38.6%; DF3 = 10.4%). The scatter plot reported in

Figure 2 does not show a clear distribution of the samples in function of their geographical origin. .

However Elva and Valle Gesso are separated along DF2 DF1 and DF2 showed higher correlations

260 with the variables related to the length of the footstalk and the number of glomerules. The variables

related to the canopy and the leaves showed a higher correlation with DF3, which, however, did not

262 provide significant results (**Supplementary Table 5**).

Discriminant Functions can be used not only to describe the phenological and morphologicaldifferences, but also to predict the origin of the plant samples.

LDA usually reports the success of a prediction model with a confusion matrix that describes the recognition success of the group prediction performed on the basis of DFs. Cross validation and prediction results report the percentage of the positive prediction divided by group.

Table 2 shows that the LDA model recognizes samples' geographical origin in the range of 66.777.8% in cross validation and 29.6%-51.9% in prediction. This result shows that the use of LDA
with morphological and phenological data does not give satisfactory results.

Differentiation of plant origin by HS-SPME-GC-MS. Figure 3 reports the HS-SPME-GC-MS profile of an *A. umbelliformis* sample with thirty identified compounds used to create the multivariate data matrix. Similarly to the morphological data, LDA was applied to find differences between plants with different geographic origins.

Each of the three Discriminant Functions calculated was significant, unlike the morphological dataset. The scatter plot of the plant samples presents a distribution similar to the one obtained with the morphological data along DF1 (**Figure 4**). Groups are better separated, especially Elva (EL) and Valle Gesso (VG), while Gran Paradiso (GP) and Val Chisone still significantly overlap. DF3 helps to differentiate GP from VC (**Supplementary Table 5**).

The improvement with HS-SPME-GC-MS data was considerable, especially in terms of prediction rates obtained with the LDA model. As reported in the related confusion matrix (**Table 3**), classification rates were calculated to be above 90% (mean value) both in validation (min=88.9%; max=100.0%; mean=93.7%) and in prediction (min=82.6%; max=96.3%; mean=90.3%).

Quantitation of thujones in plants and liqueurs by MHS-SPME-GC-MS. MHS-SPME-GC-MS analysis was used to quantify α - and β -thujone in plants and liqueurs and, then, to find a correlation between the total thujone amount in plants and liqueurs. Thirty-six plant and liqueur samples were analyzed in three replicates, by performing three subsequent HS-SPME-GC-MS analyses on each replicate. Total α - and β -thujone GC peak area estimation of each replicate was quantified by MHS-SPME-GC-MS via external calibration, analyzing α - and β -thujone standard mixture at seven 290 concentration levels. (α-thujone: y = 2205.2x - 88987.3; $R^2 = 0.9996$; β-thujone: y = 1977.6x - 250263; $R^2 = 0.9995$).

The thujone amount in liqueur and plant material was used to calculate a linear regression, which could then be used to predict the thujone content in liqueurs on the basis of the related plant material (i.e. before the liqueur preparation, in agreement with the EC protocol¹³).

The linear regression model between thujone quantities in plants and liqueurs was not satisfactory for prediction (**Supplementary Figure 1**). As a consequence, the possibility of predicting the thujone level in liqueurs had to be investigated by a different approach. LDA was then adopted to predict the probability of a liqueur to enter (or not) within the thujone EC limit on the basis of the thujone content in the plant used to produce it.

300 **Thujone level prediction in liqueurs.** A new matrix was created using the total thujone amount as 301 categorical variable and HS-SPME-GC-MS compound abundances as continuous describing 302 variables. The samples were divided into two groups based on the total thujone concentration, using 303 the 35 mg/L EC limit to discriminate the high-thujone group from the low-thujone group.

With LDA, only one DF was calculated to discriminate the two groups, because the categorical response variable was represented by only two possibilities (above and below the EC limit).

As reported in **Figure 5**, a clear discrimination between the two groups was possible using HS-SPME-GC-MS data. This result shows that the samples above the limit have the lowest values in DF scores, while the samples below the limit have the highest DF scores. The result is clear considering that the two group distributions did not overlap, giving the model a strong possibility of recognizing the response group. The model efficiency is supported by the confusion matrix (**Table** 4) that reports high validation (<35 ppm=97.6%; >35 ppm=100.0%) and prediction (<35 ppm=95.2%; >35 ppm=92.9%) rates.

The combination of HS-SPME-GC-MS data with LDA modeling achieved satisfactory prediction rates. However, the analytical process (HS-SPME-GC-MS) is still too time-consuming (70 min/sample). The method proposed is reliable in qualitative and quantitative analyses when it is 316 necessary to consider each compound, but it should be made easier and quicker to fit better with the 317 aim of this study. The analysis can significantly be speeded up by eliminating the GC separation 318 step and reducing the sampling time to five minutes, thus reducing the total analysis time to eight 319 minutes.

The scan range of the non-separative HS-SPME-MS method has been fixed between 35 and 250 m/z. As in the previous LDA model, a matrix was created based on the total thujone amount as categorical variable, but using the 215 fragment ions as continuous describing variables of each sample.

As reported in **Figure 6** and in **Table 5**, the LDA results on non-separative MS data were slightly less predictive than those obtained with GC-MS data, but they were still in a significant range of significance and prediction. The model efficiency for non-separative approach is supported by the confusion matrix (**Table 5**) that reports good validation (<35 ppm=97.7%; >35 ppm=96.4%; mean=97.1%) and prediction (<35 ppm=88.6%; >35 ppm=89.3%; mean=89.0%) rates.

329 In conclusion, the combination of chromatographic techniques (HS-SPME-GC-MS, HS-SPME-MS 330 and MHS-SPME-GC-MS) with a chemometric tool (LDA) has successfully been used to create 331 prediction models that can help to solve two important issues concerning A. umbelliformis and its 332 related génépi liqueur. HS-SPME-GC-MS data recovered from a plant sample were found to be 333 good predictive markers for: i) plant geographical origin, and ii) thujone level in the plant-related 334 liqueur. Furthermore, a non-separative methods (HS-SPME-MS) was also developed to markedly 335 shorten the analysis time, while at the same time maintaining its capacity as prediction method for 336 thujone level in the related liqueur.

337 ACKNOWLEDGMENT

This study was carried out within the project "Studio di composti biologicamente attivi del metabolismo secondario di matrici di origine vegetale" financially supported by the Ricerca Locale (Ex 60% 2015) of the University of Turin, Turin (Italy).

341

342 ASSOCIATED CONTENT

- 343 Supporting Information
- 344 Supplementary Figure 1.
- 345 Linear regression model between total thujone content of plant and liqueur. A, Correlation between
- 346 plant and liqueur MHS-SPME-GC-MS data. B, Correlation between MHS-SPME-GC-MS liqueur
- 347 data and plant-based predicted liqueur values
- 348 Supplementary Table 1.
- 349 Repeatability of HS-SPME-GC-MS analysis of all compounds detected.
- 350 Supplementary Table 2
- 351 Repeatability of HS-SPME -MS analysis of the 30 most abundant MS fragments.
- 352 Supplementary Table 3
- 353 ANOVA and Tukey Honestly Significant Difference (HSD) test on morphological parameters.

354 Supplementary Table 4

- 355 Results of LDA performed with morphological parameters. Discriminant Functions are analyzed
- 356 testing their significance in discriminating the different geographical origin in the Wilks' Lambda
- 357 Table. The second table reports the correlation between the three Discriminant Functions and the
- 358 morphological parameters.

359 Supplementary Table 5

Results of LDA performed with compounds detected by HS-SPME-GC-MS. Discriminant
Functions are analyzed testing their significance in discriminating the different geographical origin
in the Wilks' Lambda Table. The second table reports the DF value for each geographical origin.
The third table reports the correlation between the three Discriminant Functions and the HS-SPMEGC-MS compounds.

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- 465

467 **Figure captions**

468 **Figure 1**

469 Génépi illustrated to highlight phenological (A, B) and morphological (C, D, E) parameters
470 measured (Drawing: Karla Alejandra Palacios Antenucci).

471 **Figure 2**

472 Scatter plot of the first two Discriminant Functions of morphological data LDA. The geographic
473 origin is here reported by numbers: 1, Gran Paradiso; 2, Val Chisone; 3 Elva; 4, Valle Gesso.

474 **Figure 3**

475 A. umbelliformis Lam. HS-SPME-GC-MS chromatographic profile. Legend: 1) α-thujene, 2) α-

476 pinene, 3) camphene, 4) sabinene, 5) β -pinene, 6) α -phellandrene, 7) α -terpinene, 8) *p*-cymene, 9)

477 1.8-cineole, 10) γ-terpinene, 11) cis-sabinene hydrate, 12) α-terpinolene, 13) α-thujone, 14) β-

478 thujone, 15) *trans*-pinocarveol, 16) sabinol, 17) pinocarvone, 18) borneol, 19) terpinen-4-ol, 20) α-

479 terpineol, 21) cuminaldehyde, 22) α-copaene, 23) sabinyl propionate, 24) sabinyl isobutyrate, 25)

480 *trans*-β-farnesene, 26) sabinyl isovalerianate, 27) sabinyl valerianate, 28) δ-cadinene, 29)
481 caryophyllene oxide, 30) neryl isovalerianate

482 **Figure 4**

483 Scatter plot of the first two Discriminant Functions of HS-SPME-GC-MS data LDA. The
484 geographic origin is here reported by numbers: 1, Gran Paradiso; 2, Val Chisone; 3 Elva; 4, Valle
485 Gesso.

486 **Figure 5**

487 Histograms representing the frequencies of discriminant function scores of the analyzed samples by
488 HS-SPME-GC-MS. The upper panel shows samples below the limit, while the lower panel shows
489 samples above the limit

490 **Figure 6**

- 491 Histograms representing the frequencies of discriminant function scores of the analyzed samples by
- 492 HS-SPME-MS. The upper panel shows samples below the limit, while the lower panel shows
- 493 samples above the limit.

Tables

Table 1 - Geographic origin of A. umbelliformis used in the experiment as mother plants.

Geographic origin	n Code	Locations	Altitude	Geographic
Ocographic origin			Annua	coordinates
Gran Paradiso	GP	Campiglia (Torino)	1,400 m a.s.l.	45°32'27.00"N,
				7°32'17.78"E
V-1 Chiese	VC	Pragelato (Torino)	1,500 m a.s.l.	45°00'36.41"N,
Val Chisone				6°56'14.56"E
	EL	Elva (Cuneo)	1,600 m a.s.l.	44°32'16.31"N,
Elva				7°05'33.87"E
	VG	Castelmagno (Cuneo)	1,900 m a.s.l.	44°24'13.10"N,
Valle Gesso				7°09'53.98"E

		Predicted Group Membership (%)			
		GP	VC	EL	VG
	GP	66.7	25.0	0.0	8.3
X 7-1:1-4:	VC	18.5	66.7	3.7	11.1
Validation	EL	3.8	19.2	76.9	0.0
	VG	7.4	7.4	7.4	77.8
	GP	50.0	33.3	4.2	12.5
Duadiation	VC	51.9	29.6	3.7	14.8
Prediction	EL	7.7	42.3	42.3	7.7
	VG	11.1	14.8	22.2	51.9

Table 2 - Validation and prediction confusion matrices in Gran Paradiso (GP); Val Chisone (VC);Elva (EL); Valle Gesso (VG). The recognition rates of the model are not successful.

		Predicted Group Membership (%)			
		GP	VC	EL	VG
	GP	89.1	10.9	0.0	0.0
X7 . 1 · 1 . 4 ·	VC	11.1	88.9	0.0	0.0
Validation	EL	1.9	1.9	96.2	0.0
	VG	0.0	0.0	0.0	100.0
	GP	82.6	17.4	0.0	0.0
D	VC	13.0	87.0	0.0	0.0
Prediction	EL	1.9	1.9	94.3	1.9
	VG	0.0	0.0	3.7	96.3

Table 3 - Confusion matrices in validation and prediction in Gran Paradiso (GP); Val Chisone(VC); Elva (EL); Valle Gesso (VG) using LDA model created with HS-SPME-GC-MS data.

		Predicted Group (%)		
		< 35 ppm	> 35 ppm	
Validation	< 35 ppm	97.6	2.4	
	> 35 ppm	0.0	100.0	
Prediction	< 35 ppm	95.2	4.8	
	> 35 ppm	7.1	92.9	

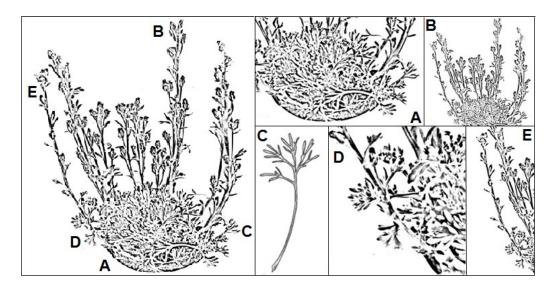
Table 4 - Confusion matrix for liqueur thujone level prediction with HS-SPME-GC-MS data.

		Predicted Group (%)	
		< 35 ppm	> 35 ppm
Validation	< 35 ppm	97.7	2.3
	> 35 ppm	3.6	96.4
Prediction	< 35 ppm	88.6	11.4
	> 35 ppm	10.7	89.3

Table 5 - Confusion matrix for liqueur thujone level prediction with HS-SPME-MS data.

Figures

Figure 1





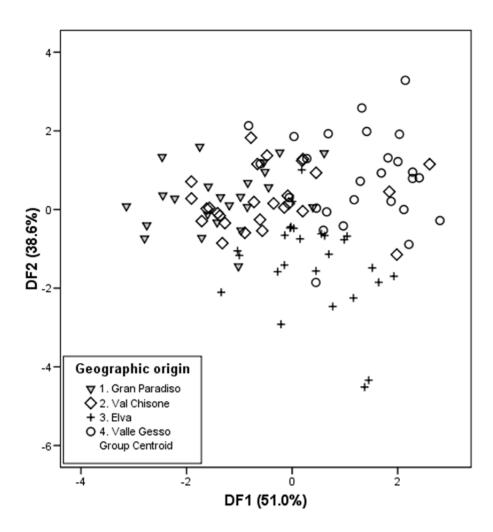
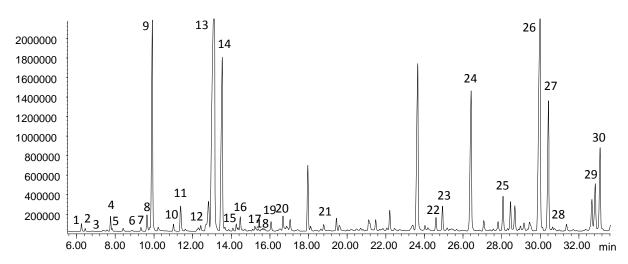
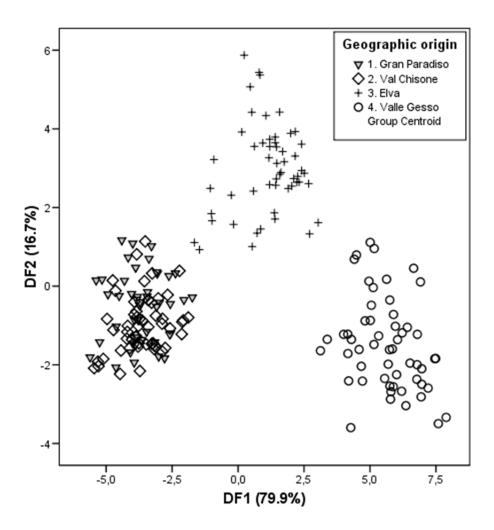


Figure 3

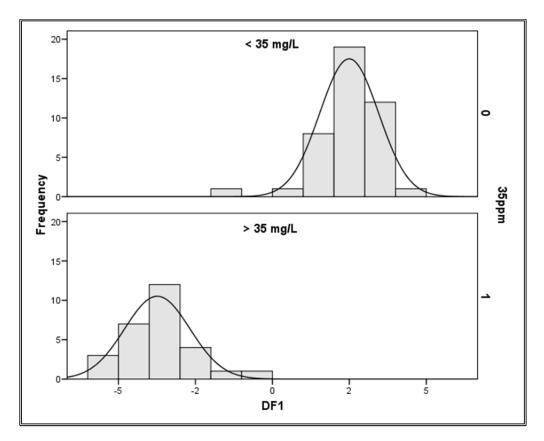
Abundance



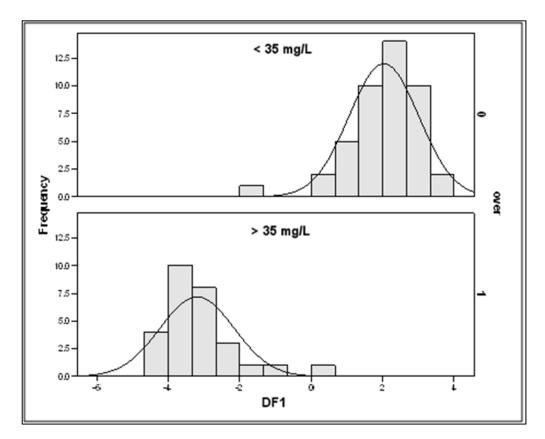












GRAPHIC FOR TABLE OF CONTENTS

