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## Metabotyping of Andean pseudocereals and characterization of emerging mycotoxins

Pedro G Vásquez-Ocmín, Guillaume Marti, Alice Gadea, Guillaume Cabanac, Juan Vásquez-Briones, Sandro Casavilca-Zambrano, Nadia Ponts, Patricia Jargeat, Mohamed Haddad, Stéphane Bertani

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1 grains and try to parallel this information with the presence of the molecules produced, especially  
2 mycotoxins. Metabotyping permitted to add new insights into the chemotaxonomy of pseudocereals,  
3 confirming the previously established phylotranscriptomic clades: Chenopods I (clusters quinoa and  
4 canihua), and Amaranthaceae s.s. (cluster kiwicha). Moreover, we report for the first time the presence  
5 of mycotoxins in pseudocereals. Sixteen samples of Peru (out of 27) and one sample from France (out  
6 of one) were contaminated with Beauvericin, an emerging mycotoxin. There were several  
7 mycotoxigenic fungi detected, including *Aspergillus sp.*, *Penicillium sp.*, and *Alternaria sp.*, but not  
8 *Fusaria*.

9 Keywords: Andean pseudocereals, metabotyping, mycotoxins, fungal communities.

10

## 11 **1. Introduction**

12 Pseudocereals are best known for three crops derived from the Andes: quinoa (*Chenopodium quinoa*),  
13 canihua (*C. pallidicaule*), and kiwicha (*Amaranthus caudatus*). The use of these pseudocereals as  
14 "superfood" for healthier diets has been growing exponentially over the past decade because of their  
15 high content of essential amino acids, antioxidants, fibers, and vitamins while being gluten-free  
16 (Pereira et al., 2019). Due to its high nutritional value, the Food and Agriculture Organization (FAO) has  
17 deemed quinoa a "golden seed" because of its potential to ensure food safety and nutrition for the  
18 world's population amid climate change. Thus, pseudocereals, are not only recognized for their  
19 nutritional benefits but also for their ability to adapt to a variety of agro-environmental conditions.  
20 Indeed, Peru, Bolivia, and Ecuador are the main producers of quinoa, canihua, and kiwicha, where  
21 these pseudocereals are especially well adapted to the abiotic stresses found in Andean highlands  
22 (Bonifacio, 2019).

23 There is a higher level of polyphenism among the pseudocereals cultivated in the Andean  
24 countries. For instance, three types of quinoa have been categorized according to the color of the  
25 seeds: red quinoa, white quinoa, and black quinoa. However, the chemical foundation that would rely  
26 upon such polyphenism has not been thoroughly investigated. Meanwhile, the chemical food safety of  
27 pseudocereals remains poorly documented as well. Despite their appeal to an organic diet,  
28 pseudocereals may still have the potential for adverse effects from secondary metabolites produced  
29 by infesting microorganisms, such as mycotoxin-producing fungi. Mycotoxins are naturally occurring  
30 secondary metabolites with adverse effects on human and animal health upon ingestion. They are  
31 mainly produced by strains of the three fungal genera *Aspergillus*, *Fusarium*, and *Penicillium* (Bräse et  
32 al., 2009). More than 25% of the crop production worldwide bears contamination with mycotoxins,  
33 making them ubiquitous in cereals and cereal-derived foods (Eskola et al., 2020; Lee & Ryu, 2017).  
34 Therefore, it is essential to better understand local diversity and mycotoxigenic risk for pseudocereals

1 in the context of a global demand that encourages market diversification and sustainable and equitable  
2 development.

3 Metabolomics based on technology like hyphenated liquid chromatography-mass spectrometry  
4 (LC-MS) assisted by bioinformatics and statistical analysis is a powerful approach for both the detection  
5 and the annotation of key compounds, as well as for deciphering their biological functions. Our team  
6 has previously implemented a workflow based on untargeted LC-MS-based metabolomics able to  
7 achieve the metabotyping of plants by identifying their main biomarkers (Vásquez-Ocmín et al., 2021).  
8 In the present study, we applied this approach to both achieve a comprehensive chemical mapping of  
9 pseudocereal samples collected in the Andes, and to classify them according to their chemotype.  
10 Additionally, targeted LC-MS metabolomics was applied on the same samples to quantify their  
11 contents in mycotoxins Aflatoxin B<sub>1</sub>, Beauvericin, Fumonisin B<sub>1</sub>, Ochratoxin A, and Patulin. Finally, an  
12 inventory of the fungal community infesting pseudocereal samples from the Andes was documented,  
13 thereby correlating the presence of toxicogenic fungi and their associated secondary metabolites.

14

## 15 **2. Material and methods**

16 All solvents used for chromatography were LC-MS grade (HiPerSolv Chromanorm, VWR, Fontenay-  
17 Sous-Bois, France). Milli-Q RG system (Millipore, France) was used to produce high purity water of 18.2  
18 MΩ.cm resistivity. Formic acid was >98% for LC-MS (Fluka, Buchs, Switzerland). Mycotoxins standards  
19 [*i.e.*, Aflatoxin B<sub>1</sub> (AflaB<sub>1</sub>), Beauvericin (BEA), Fumonisin B<sub>1</sub> (FumB<sub>1</sub>), Ochratoxin A (OTA), and Patulin  
20 (PAT)] were purchased from Sigma-Aldrich (St. Louis, MO, USA) and immediately stored at -30 °C.

21

### 22 *2.1. Plant material*

23 Twenty-seven grain samples were collected between September 2019 and December 2020 in  
24 seven regions of Peru (elevation range: 1,410-3,855 meters height above mean sea level) and one grain  
25 sample was bought in a local market in Toulouse, France. For quinoa samples from Peru, eight grains  
26 were collected directly (unwashed) and the 13 others having been washed (see the list and description  
27 of varieties in **supp info 1**). After the collection, all grain samples were immediately stored in airtight  
28 bags at -20°C. Authorization to access genetic resources was granted by the competent authority (INIA)  
29 under Biodiversity Collection Permit Number 021-2018-MINAGRI-INIA-DGIA/SDRIA.

30

### 31 *2.2 Metabolome analyses*

32

#### 33 *2.2.1. Untargeted metabolomics*

34 For each grain sample (100 mg), a biphasic microextraction was performed in triplicate  
35 following: i) first, 1.5 mL of Methyl-tert butyl ether (MtBE)/MeOH (75:25, v/v), then vortex and

1 sonication for 3 min, ii) second, 1.5 mL of water/MeOH (75:25, v/v), then vortex and sonication for 3  
2 min. The suspension was centrifuged at 14,000 rpm for 3 min at 5°C, then, 500 µL of medium polar  
3 extract was taken and dried at 30°C using a vacuum concentrator. The residue was then resuspended  
4 using methanol and then stored at -30 °C prior analysis. A quality control (QC) sample was made by  
5 pooling aliquots from each extract. Final concentration for untargeted metabolomic analyses was  
6 performed at 50-fold dilutions.

7 On washed samples, an ultra-high-performance liquid chromatography high-resolution mass  
8 spectrometry (UHPLC–HRMS) analyses were performed on a Q Exactive™ Plus Hybrid Quadrupole-  
9 Orbitrap™ Mass Spectrometer equipped with a Heated Electrospray Ionization (HESI-II) Probe coupled  
10 to a UltiMate™ 3000 Rapid Separation LC (RSLC) system (Thermo Fisher Scientific, Villebon-Sur-Yvette,  
11 France). Separation was performed on a Luna Omega Polar C18 UHPLC column (150 mm × 2.1 mm i.d.,  
12 1.6 µm) with a guard column (Phenomenex, Sartrouville, France). The mobile phase gradient used  
13 water containing 0.05 % formic acid (A) and acetonitrile containing 0.05% of formic acid (B). A gradient  
14 method at a constant flow rate of 0.3 mL/min was applied under the following conditions: 0 % B for 1  
15 min to 100 % B in 20 min, held at 100 % B over 4 min, then equilibration with 100 % over 4 min. The  
16 column oven temperature was set to 40°C, the autosampler temperature was set to 5°C, and the  
17 injection volume was fixed to 1 µL. Mass detection was performed in positive ionization (PI) and  
18 negative ionization (NI) modes at resolution 30,000 [fullwidth at half-maximum (fwhm) at 400m/z] for  
19 MS1 and resolution 17,500 with automatic gain control (AGC) target of  $1 \times 10^6$  for full scan MS1 and  
20  $1 \times 10^5$  for MS2. Ionization spray voltage was set to 3.5 kV (for PI) and 2.5 kV (for NI), and the capillary  
21 temperature was set to 256 °C for both modes. The mass scanning range was  $m/z$  100–1500. Each full  
22 MS scan was followed by a data-dependent acquisition of MS/MS spectra for the six most intense ions  
23 using stepped normalized collision energy of 20, 40, and 60 eV.

24

### 25 2.2.2. Targeted metabolomics

26 For each grain sample (5 g), extraction was performed in triplicate using 20 mL of ethyl acetate  
27 and sonication for 1 h. The suspension was centrifuged at 14,000 rpm for 10 min, and the supernatant  
28 dried at 30°C using a vacuum concentrator. The residue was then resuspended using methanol and  
29 then stored at -30 °C prior analysis. Final concentration for targeted metabolomic analyses was  
30 performed at 12-fold dilutions.

31 On all set samples in **supp info 1**, a Triple Quad 5500+ LC-MS/MS System – QTRAP Ready (AB  
32 Sciex, Villebon sur Yvette, France) equipped with a Turbo Spray source. Separation was performed on  
33 an ACQUITY UPLC® BEH C18 column (100 mm × 2.1 mm i.d., 1.7 µm, Waters, MA, USA) equipped with  
34 a guard column. The mobile phase gradient used water containing 0.1 % formic acid (A) and acetonitrile  
35 containing 0.1% of formic acid (B). A flow rate of 0.3 mL/min was applied following: isocratic

1 conditions, 10% of B and 90% of A for 3 min, then, a gradient of B until 70% in 7 min, and went up to  
2 90% in 1.1 min (curve 6). After 1.9 min, it went back to the initial condition in 0.1 min, and then a re-  
3 equilibration for 2.9 min. The column oven temperature was set to 40°C, the autosampler temperature  
4 was set to 5°C, and the injection volume was fixed to 5 µL. Mass detection was performed in PI and NI  
5 modes at nominal resolution. Ionization spray voltage was set to 4.5 kV (for PI) and 4.5 kV (for NI), and  
6 the capillary temperature was set to 500 °C, curtain gas set at 20, ion source gas at 1 at 60 and ion  
7 source gas 2 at 70 for both modes. Two main transitions for each standard were performed in PI and  
8 NI (details in **supp info 2**). A curve of calibration was carried out for targeted metabolomics from a mix  
9 of standards to final concentrations of: 100 ng/mL, 10 ng/mL, 1 ng/mL, 0.1 ng/mL, 0.01 ng/mL, and  
10 0.001 ng/mL in acetonitrile.

11

## 12 *2.3. Data mining process*

13

### 14 *2.3.1. Untargeted metabolomics*

15 LC-MS data were processed according to the MSCleanR workflow (Fraisier-Vannier et al., 2020;  
16 Vázquez-Ocmín et al., 2021). Briefly, batches in PI and NI were processed separately with MS-DIAL  
17 version 4.80 (Tugawa, Cajka, et al., 2015). MS1 and MS2 tolerances were set to 0.01 and 0.05 Da,  
18 respectively, in centroid mode for each dataset. Peaks were aligned on QC reference with an RT  
19 tolerance of 0.2 min, a mass tolerance of 0.015 Da, and a minimum peak height detection at  $1 \times 10^6$   
20 for both ionization modes. MS-DIAL data (both PI and NI) were deconvoluted together with MS-CleanR  
21 by selecting all filters with a minimum blank ratio set to 0.8 and a maximum relative standard deviation  
22 (RSD) set to 40. The maximum mass difference for feature relationships detection was set to 0.005 Da,  
23 and the maximum RT difference was set to 0.025 min. The Pearson correlation links were applied for  
24 a correlation  $\geq 0.8$  and a p-value significance threshold = 0.05. Two peaks were kept in each cluster for  
25 further database requests and the kept features were annotated with MS-FINDER version 3.52  
26 (Tugawa et al., 2016). The MS1 and MS2 tolerances were set to 5 and 15 ppm, respectively. The  
27 formula finder was exclusively processed with C, H, O, and N atoms. Two pathways from several sets  
28 of data were carried out for the compound's annotation. For metabolite annotation at level 1:  
29 experimental LC-MS/MS data for 500 compounds (including five mycotoxins standards), analyzed on  
30 the same instrument and with the same method, were used as references using MS-DIAL based on  
31 accurate and exact mass, fragmentation, and retention time. For metabolite annotation level 2: the  
32 MS-DIAL, MONA (MassBank of North America), and GNPS (Global Natural Product Social Molecular  
33 Networking) mass spectral records were used for spectral match using a dot product score cut-off of  
34 800. The *in silico* match (annotation level 3) were prioritized according to: 1) compounds identified in  
35 the literature for both *Chenopodium* and *Amaranthus* genera, and both *Amaranthaceae* and

1 *Chenopodiaceae* botanic family (Dictionary of Natural Products version 28.2, CRC press) mined by MS  
2 Finder based on exact mass and in silico fragmentation; 2) generic databases included in MS-FINDER  
3 (*i.e.*, FooDB, PlantCyc, ChEBI, T3DB, Natural Products Atlas, NANPDB, COCONUT, and KNApSack);  
4 3) fungal compounds (including mycotoxins) were performed using four “in-house” databases from  
5 literature to compile: i) only mycotoxins (n=300) and ii) compounds produced from major  
6 mycotoxigenic fungi *Fusarium*, *Aspergillus*, and *Penicillium*.

7 Mass spectra similarity networking was carried out from PI mode using MetGem (Olivon et al.,  
8 2018) on the final annotated .msp file and metadata files compiled for both batches obtained with  
9 MSCleanR. Values used were MS2  $m/z$  tolerance = 0.02 Da, minimum matched peaks = 4 and minimal  
10 cosine score value = 0.7. Visualization of the molecular network (MN) was performed on Cytoscape  
11 version 3.9.1 (Shannon et al., 2003). Original LCMS raw data were uploaded in zenodo (DOI:xxxx)

12

### 13 2.3.2. Targeted metabolomics and mycotoxins quantification

14 Dataset acquired in format .wiff was converted in .abf format using Analysis Base File  
15 Converter (Reifycs, Tokyo, Japan). Analyses were performed with MRMPROBS program version 2.60  
16 (Tsugawa et al., 2013; Tsugawa, Ohta, et al., 2015), using as parameters of peak detection: five scans  
17 for minimum peak width; 2,000 as minimum peak height; 0.2 min as retention time tolerance for peak  
18 identification; 15% as amplitude tolerance; and 70% as minimum posterior. A list of transitions in .txt  
19 format was used as a compound library file. A final list containing the information about the Area above  
20 zero for each transition selected in the samples was exported in .txt format. From the final list of  
21 information about the mycotoxin standards, a calibration curve was created for each concentration  
22 mentioned above. Validation parameters were determined following the International Conference on  
23 Harmonization (ICH) Guidelines (ICH, 2005), and the following characteristics were evaluated: 1)  
24 linearity, 2) limit of detection (LOD), and 3) limit of quantification (LOQ). Wheat contaminated with  
25 enniatins, and beauvericin was used as positive control. Original LC-MS raw data were uploaded in  
26 zenodo (DOI:10.5281/zenodo.6700223).

27

### 28 2.4. Statistical analyses

29 The final annotated metabolome dataset generated by the MS-CleanR workflow was uploaded  
30 on the MetaboAnalyst 5.0 online platform (Chong et al., 2019). The data were normalized by sum and  
31 scaled by Pareto before statistical analysis. First, principal component analysis (PCA) was applied as an  
32 exploratory data analysis to provide an overview of LC-MS fingerprints. Then, a hierarchical cluster  
33 analysis (HCA) was performed to obtain a dendrogram of varieties using the untargeted metabolome  
34 dataset. For each variable the average value across replicates was calculated; then distance matrices  
35 were obtained using either 1-r (Pearson correlation coefficient), followed by HCA on each matrix using

1 Ward's minimum variance criterion using Orange 3.32.0 (Demšar et al., 2013). On the clusters  
2 obtained, a partial least squares discriminant analysis (PLS-DA) was conducted using clusters as Y-  
3 value, and the top 50 features were plotted on a heatmap using ANOVA. GraphPad Prism 5.0 was used  
4 to perform the bar-plot for data quantification.

5

## 6 *2.5. Fungi isolation and molecular identification*

7 For each sample (**supp info 1**), dried seeds were deposited on Malt Extract Agar (MEA) plates  
8 (diameter 15 cm) containing chloramphenicol (100 µg/mL) under the laminar airflow (LAF). Plates were  
9 incubated at 27°C daylight. After at least 48h of incubation, growing fungi were replicated on MEA  
10 plates without chloramphenicol and incubated at 27°C, with daylight. All the isolated mycelia are  
11 stored in UMR 152 laboratory collection (cryopreservation at -80°C in 20% of glycerol).

12 Total DNA was extracted from mycelium directly picked from the plate, using the Wizard®  
13 Genomic DNA Purification kit (Promega, Charbonnières-Les-Bains, France). The final pellet was  
14 resuspended in 40 µL of sterile ultra-high quality (UHQ) water and diluted 10 times in UHQ water. The  
15 ITS rDNA region was amplified by polymerase chain reaction (PCR) using the primer set ITS5/ITS4  
16 (White et al., 1990). Reactions were carried out in a final volume of 25 µL, containing 2 µL of diluted  
17 DNA, 5 µL GoTaq polymerase buffer 5x (Promega), 0.5 µL of each primer, 0.5 µL of dNTP, 0.5 units of  
18 GoTaq® Hot Start polymerase (Promega), and 16.5 µL of UHQ sterile water. PCR cycling conditions of  
19 the Mastercycler® Nexus Thermal Cycler (Eppendorf AG, Hamburg, Germany) were as follows: initial  
20 denaturation at 95°C for 3 min; 35 cycles at 95°C for 45 s, 55°C for 45 s, 72°C for 1 min; and a final  
21 extension at 72°C for 10 min. For identification at the species level of the *Aspergillus* and *Penicillium*  
22 isolates, beta-tubuline and calmoduline regions were PCR-amplified using the primers Bt2a/Bt2b  
23 (Glass & Donaldson, 1995) and CMD5/CMD6 (Hong et al., 2005), in the same conditions as above,  
24 except for annealing temperature (58°C instead of 55°C).

25 PCR amplicons were sequenced by Eurofins Genomics (Ebersberg, Germany) according to a  
26 Sanger method using ITS5, Bt2a, and CMD5 respectively. Sequences were submitted to the BLASTn  
27 program (<http://blast.ncbi.nlm.nih.gov/Blast.cgi>) to determine the close matches in the GenBank  
28 database. All ITS sequences were aligned with MAFFT v6.814b (Kato et al 2002) using Geneious®6.1.8.  
29 The PhyML method (Guindon & Gascuel, 2003) was used *via* the Geneious® platform to generate a  
30 maximum-likelihood phylogenetic tree with the following setting: GTR substitution model, 100  
31 bootstraps, estimated transition/transversion ratio, the estimated proportion of invariable sites,  
32 estimated gamma distribution, branch length, and optimized substitution rate. Phylogenetic trees  
33 were visualized and edited with MEGAX (Kumar et al., 2018).

34

35

### 3. Results and discussion

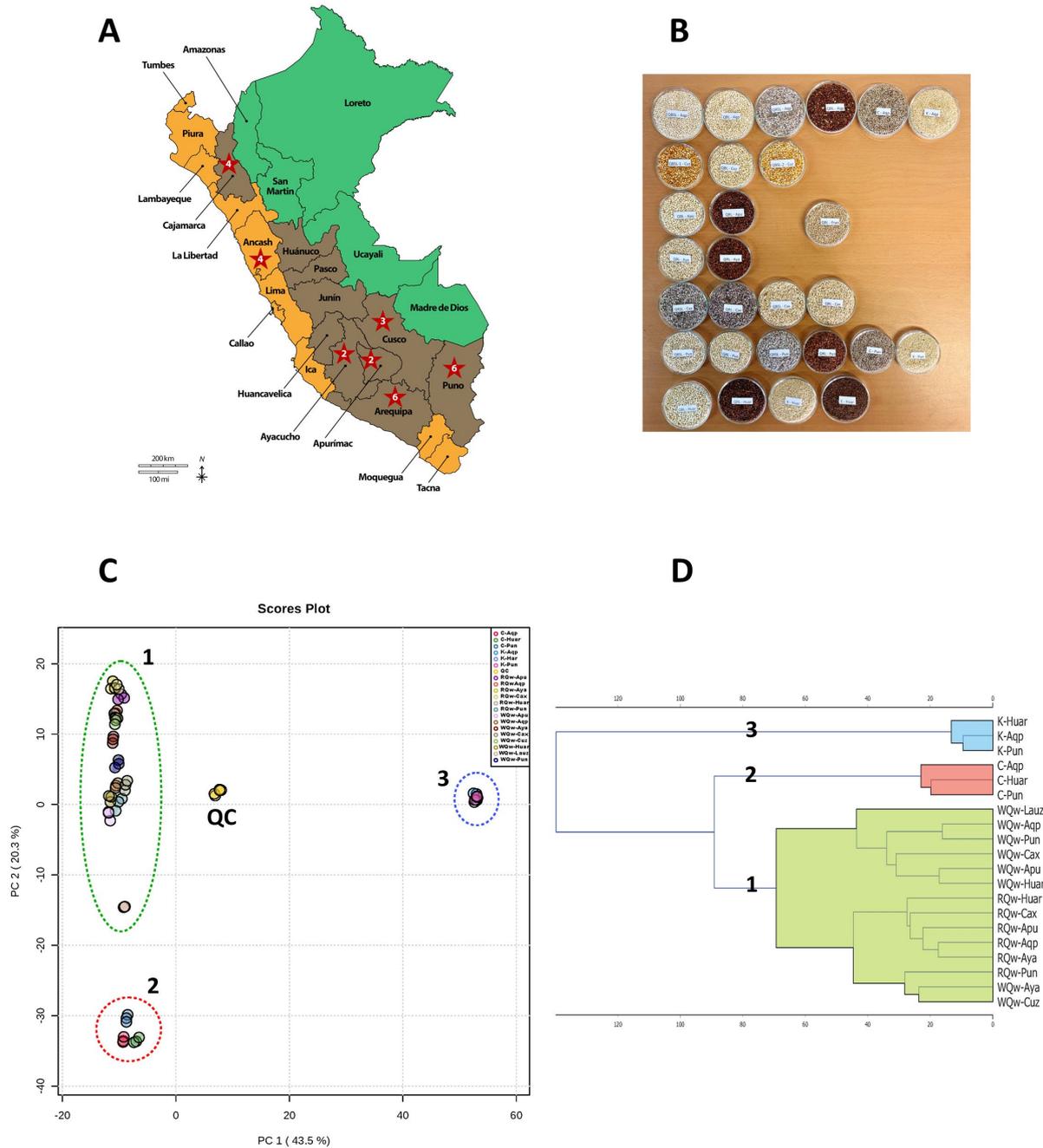
#### 3.1. Untargeted metabolomics for the metabotyping of Andean pseudocereals

Nowadays, chemotype discrimination between plants of the same botanical genus or family represents a challenge in the functional food sector. Chemotypes are characterized by the presence of certain metabolites, as major or minor compounds, in a plant. Despite their potential to enhance the food basket, the chemodiversity of pseudocereals has received little attention so far. To explore this chemodiversity, we used an untargeted metabolomics approach referred to as "metabotyping" as we did in a previous work investigating Cannabis cultivars (Vásquez-Ocmín et al., 2021). We thus analyzed the whole metabolomes of several pseudocereal grains collected in the Andes of Peru, as well as one quinoa sample from France (**Fig. 1-A**). To ensure a comprehensive analysis of medium polar extract, positive ionization mode was investigated using the MS-CleanR workflow, which resulted in a list of 219 features ( $m/z \times RT$ ). Depending on the level of annotation mentioned above, 19 compounds were yielded for the genus (8.7%), 43 compounds for the family (19.6%), 128 compounds for the generic level (58.4 %), and 29 unknowns (13.2 %). The fitness of compound annotation generally depends on the databases (DB) used. Despite the few studies on pseudocereals, we confirmed an intern DB from family and genera gathered from the literature, responsible for 28.3 % of annotations (annotation level 3-1). This was complemented with our experimental DB (annotation level 1) and the spectral match with other DB in line (annotation level 2). Using NPClassifier and ClassyFire (Djoumbou Feunang et al., 2016; Kim et al., 2021) the major classes and subclasses of compounds following their biosynthetic pathway were identified (**supp info 3**).

Phylogenetic taxonomy has grouped the previously separated subfamily *Chenopodiaceae* within the family *Amaranthaceae* s.l., which include now the five phylotranscriptomic clades *Amaranthaceae* s.s., *Betoideae*, "Chenopods I", "Chenopods II", and *Polycnemoideae* (Morales-Briones et al., 2020). In our study, we have profiled by LC-MS 20 pseudocereal grain samples that belonged to: Chenopods I [*Chenopodium quinoa* (14 samples) and *C. pallidicaule* (3 samples)] and *Amaranthaceae* s.s. [*Amaranthus caudatus* (syn: *A. cruentus*) (3 samples)]. To obtain an unsupervised overview of chemical fingerprint distribution, we firstly performed a principal component analysis (PCA). The PCA score plot from the two first principal components accounted for 63.8% of the total variance (**Fig. 1-B**). As expected, repeated injections of QC were grouped near the center of the PCA score plot demonstrating a good reproducibility and stability of the analytical workflow. Then, a hierarchical clustering analysis (HCA) was performed with 20 samples (three analytical repeats for each sample), obtaining three very well differentiated clusters (**Fig. 1-C**). So, our metabolomic signature showed that *Chenopodium* species are grouped in two distinct clusters: **cluster 1** (14 samples of **quinoa**) and **cluster 2** (3 samples of **canihua**). **Cluster 3** brought together the three *Amaranthus* grains (**kiwicha**). On the

1 other hand, the provenance (in all clusters) and color of grains (in the **cluster quinoa**) did not play a  
2 significant role in the metabolomic classification intraspecies. However, inside the **cluster quinoa**, two  
3 other subgroups for red and white quinoa can be observed. An exception is white quinoa samples from  
4 Ayacucho and Cuzco that were compiled with red quinoa samples, which could infer a putative  
5 hybridization intraspecies. Additionally, to validate the HCA model and highlight characteristic  
6 biomarkers of each cluster, we performed a supervised Partial Least Squares-Discriminant  
7 Analysis (PLS-DA) based on the variable importance in projection (VIP) and a heat map was  
8 generated from the top 50 metabolites selected by Anova (**supp info 4, 6-8**).

9



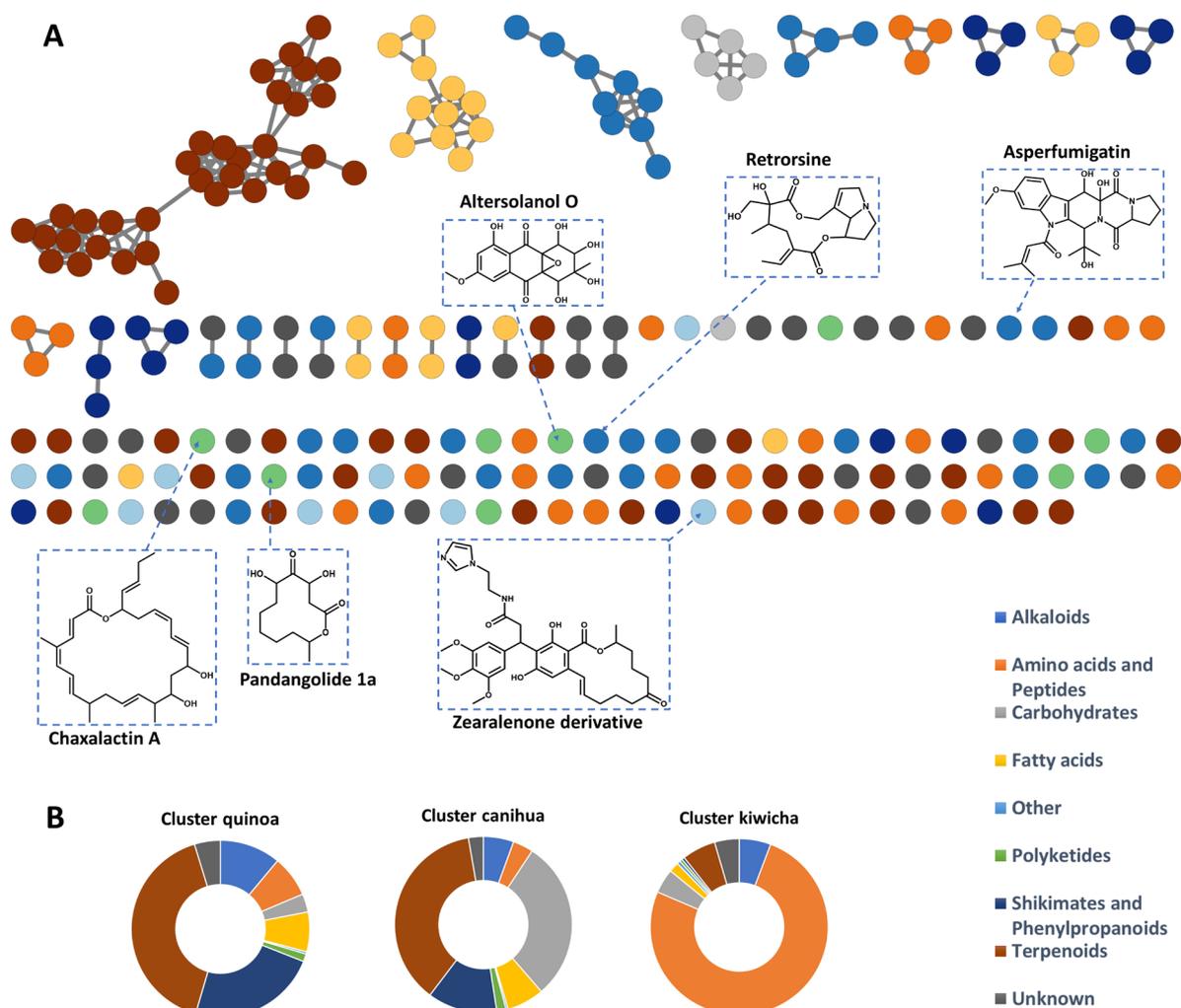
1  
 2 **Figure 1. (A)** Geographic origins of the grain samples. The names of the 24 administrative departments  
 3 of Peru are indicated in black. The departments of collection are stamped with a red star marked with  
 4 the number of samples collected therein. Green = Peruvian Amazon (including highland and lowland  
 5 jungles); brown = Andean highlands; orange = arid and semi-arid Pacific coastal lands. **(B)** Set of  
 6 samples used for the metabolomics analyses. **(C)** Global PCA of 20 pseudocereal grains from Peru  
 7 (except for WQ-Lauz, France). **(D)** Hierarchical clustering analysis (HCA) following the whole  
 8 metabolome to group all grain samples. Pseudocereal varieties: WQw-Cuz = white quinoa washed from  
 9 Cuzco, WQw-Aya = white quinoa washed from Ayacucho, RQw-Aya = red quinoa washed from

1 Ayacucho, WQw-Apu = white quinoa washed from Apurimac, RQw-Apu = red quinoa washed from  
2 Apurimac, WQw-Cax = white quinoa washed from Cajamarca, RQw-Cax = red quinoa washed from  
3 Cajamarca, WQw-Pun = white quinoa washed from Puno, RQw-Pun = red quinoa washed from Puno,  
4 K-Pun = kiwicha from Puno, C-Pun = canihua from Puno, WQw-Aqp = white quinoa washed from  
5 Arequipa, RQw-Aqp = red quinoa washed from Arequipa, K-Aqp = kiwicha from Arequipa, C-Aqp =  
6 canihua from Arequipa, WQw-Huar = white quinoa washed from Huaraz, RQw-Huar = red quinoa  
7 washed from Huaraz, K-Huar = kiwicha from Huaraz, C-Huar = canihua from Huaraz, WQw-Lauz = white  
8 quinoa washed from Lauzerte, France.

9  
10 After our workflow analysis, unanticipated metabolites produced by filamentous fungi were  
11 identified in the final metadata. These metabolites were placed in a molecular network (**Fig. 2-A**). The  
12 molecular network of the whole dataset analyzed in PI mode was constructed to identify possible  
13 similarities between molecules based on spectral mass similarity (MS-MS fragmentation patterns) (**Fig.**  
14 **2**). By grouping the major classes of compounds in clusters according to their biosynthetic pathway,  
15 there were a number of characteristic groups, such as amino acids and peptides, carbohydrates,  
16 alkaloids, fatty acids, polyketides, shikimates and phenylpropanoids, terpenoids, and others. Here, we  
17 can also observe the macrolide derivatives pandangolide 1a isolated from *Cladosporium* sp. (Gesner et  
18 al., 2005), and chaxlactin A from *Streptomyces* strain (Rateb et al., 2011), the indole diketopiperazine  
19 alkaloid derivative asperfumifatin previously isolated from the endophytic *Aspergillus fumigatus* (Xie  
20 et al., 2015), and the polyketide derivative altersolanol O from *Alternaria* sp. (Chen et al., 2014) (**Fig.**  
21 **2-A**). Another compound identified was a zearalenone derivative (3-(14,16-dihydroxy-3-methyl-1,7-  
22 dioxo-3,4,5,6,7,8,9,10-octahydro-1H-2-benzoxacyclotetradecin-15-yl)-N-[2-(1H-imidazol-1-yl)ethyl]-3-  
23 (3,4,5-trimethoxyphenyl) propenamide). This compound was also identified as a marker for the **cluster**  
24 **kiwicha (supp info 4-C)**. Zearalenone (ZEA) type of compounds are produced by several mycotoxigenic  
25 *Fusarium* (*F. graminearum*, *F. cerealis*, *F. culmorum*, *F. equiseti*, etc.), and are considered as  
26 mycotoxins. ZEA is distributed worldwide and naturally occurs in agricultural crops, particularly in  
27 maize. This mycotoxin could contaminate products made of barley, wheat, oats, rice, sorghum, and  
28 others (G.-L. Zhang et al., 2018). The structure is a macrolide type, and the name “zearalenone” is  
29 derived from the combination of the terms maize (*Zea mays*)—“zea”, resorcylic acid lactone—“ral”, —  
30 “en” for the presence of a double-bond, and “one” for the ketone group (Ropejko & Twarużek, 2021).  
31 The flexibility of the molecule is enough to adopt a conformation able to bind to the mammalian  
32 estrogen receptor, although with an affinity lower than the natural estrogen 17- $\beta$ -estradiol resulting  
33 in severe effects on the reproductive system in several animal species, particularly pigs (Galaverna &  
34 Dall’Asta, 2012; G.-L. Zhang et al., 2018). Even if the molecule identified has the same macrolide  
35 structure of ZEA, this latter also presents a tyrosine alkaloid moiety. So far, to our knowledge, no

1 bibliography about their toxicity was found. In this sense, close attention was taken to the compounds  
2 having structural similarities. In addition, another macrolide derivative (retrorsine) previously reported  
3 from *Senecio gilliesiano* or *S. gilliesianus* (Asteraceae) (Guidugli et al., 1986) was annotated. All these  
4 metabolites described above are placed as self-loop in the molecular network. This suggests that there  
5 are no other molecule derivatives in the whole metabolome of the grains.

6 Moreover, we displayed the information correlated between the major class of compounds  
7 and the cluster producing (**Fig. 2-B**). As described in the literature, here we can observe that terpenoids  
8 (41%) and molecules belonging at shikimates and phenylpropanoids (24%) (*e.g.* flavonols,  
9 anthocyanidins, cinnamic acid derivatives, etc) are the most representative for the **cluster quinoa** (Lin  
10 et al., 2019; Tabatabaei et al., 2021). Among the terpenoids, saponins were not identified, it is due to  
11 the postharvest process, where the grains are washed on various days with plenty of water to remove  
12 saponins localized in the pericarp (Jarvis et al., 2017). Triterpene saponins are known for their  
13 ecological defense role in the plant but are also known by the local people for providing a bitterness  
14 “*amargor*” or bitter-tasting “*sabor amargo*” to the unwashed grains (McCartney et al., 2019). For the  
15 **cluster canihua**, terpenoids (37%) and carbohydrates (29%) are the most representative class of  
16 molecules. At the opposite of the *Chenopodium* grains of the **clusters quinoa** and **canihua**, the  
17 majoritarian compounds identified for the **cluster kiwicha** were the amino acids (76%), mainly small  
18 peptides (**Fig. 2-B**). Due to the nutritional richness, in the last years the pseudocereals studied here are  
19 appreciated worldwide (Fletcher, 2016; Perez-Rea & Antezana-Gomez, 2018).



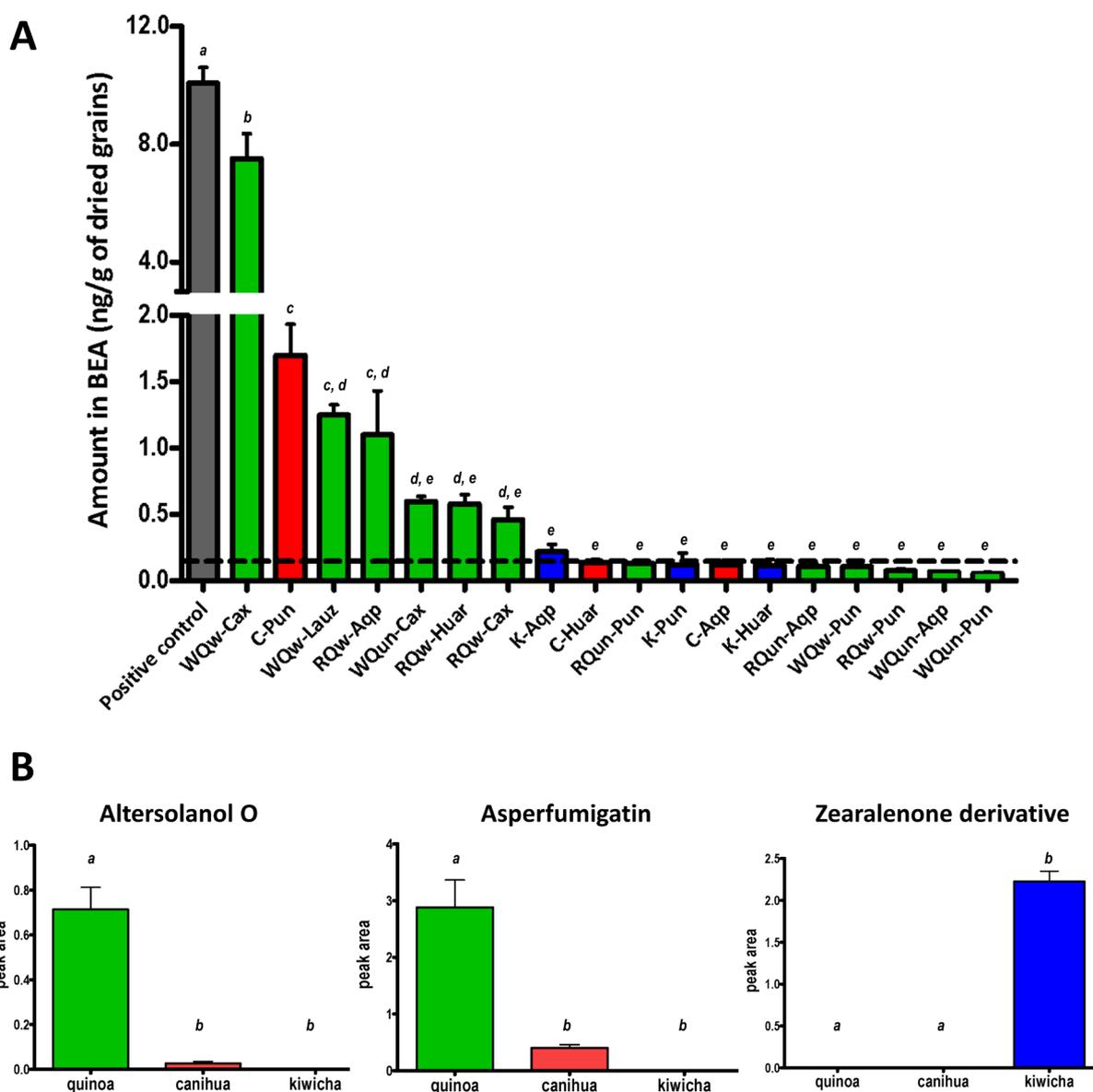
1  
 2 **Figure 2.** Molecular network (PI mode) from compounds identified in pseudocereals, based on  
 3 MetGem and visualized by Cytoscape. Colors in **A** and **B** are correlated with the major classes of  
 4 compounds annotated. Structures and names of some metabolites annotated from fungal biosources  
 5 are represented in **A**. In **B**, the clusters are correlated with **Fig 1-C**. In these clusters, we represented  
 6 the percentage of the major classes of compounds obtained by cluster.

7  
 8 **3.2. Targeted metabolomics reveals the presence of an emerging mycotoxin in Andean**  
 9 **pseudocereals**

10 We complemented our approach with a targeted metabolomic analysis using LC-MS Qtrap  
 11 Multiple Reaction Monitoring (Qtrap-MRM). Specifically, we focused on Aflatoxin B1 (AflaB1),  
 12 Beauvericin (BEA), Fumonisin B1 (FumB1), Ochratoxin (OTA), and Patulin (Pat), which are mycotoxins  
 13 produced by several species of *Fusarium*, *Aspergillus*, and *Penicillium*. Before implementing QqQ-  
 14 MRM, we designed the main transitions (fragments) in both PI and NI modes of mycotoxins standards,  
 15 as described previously (**sup. info 2**) (Stead et al., 2014). We then assessed whether post-harvest  
 16 washing could eliminate mycotoxins by comparing targeted metabolomes of washed and unwashed

1 quinoa grains. On the five mycotoxins targeted, only FumB<sub>1</sub> and BEA were detected in PI mode (**supp**  
2 **info 9**). Among all samples, only seven of the quinoa samples trace amounts of FumB<sub>1</sub> (not quantifiable,  
3 <LOQ) (**supp info 10**). This result supports the interest in the consumption of quinoa as safe food,  
4 taking into account that our team has previously reported that other Andean grains may be  
5 contaminated with high levels of FumB<sub>1</sub> (notably maizes), exceeding the maximum food safety level of  
6 1,000 ng/g (Ducos et al., 2021). Fumonisin, which have a long carbon chain aminoalcohol structure as  
7 their basic skeleton, are structurally similar to sphingosines (sphingoids). These mycotoxins are  
8 primarily produced by *F. verticillioides* and *F. proliferatum* (Smith, 2018). It has been suggested that  
9 FumB<sub>1</sub> may increase the risk of cancer, notably in kidney and liver. Thus, FumB<sub>1</sub> was classified by the  
10 International Agency for Research on Cancer (IARC) as a potential human carcinogen (Group 2B)  
11 (Claeys et al., 2020; Ostry et al., 2017). Concerning BEA, it was detected in 17 pseudocereal grain  
12 samples from Peru and in the only sample from France, with levels equal to or greater than the limit  
13 reported by the European Food Safety Authority (EFSA) for dietary acute exposure (*i.e.*, 0.05 ng/mL or  
14 0.14 ng/g of dried seeds) (**Fig. 3-A**) (EFSA, 2014). BEA and enniatins are emerging mycotoxins since  
15 they are not routinely determined nor regulated like other *Fusarium* mycotoxins, but evidence of their  
16 occurrence is rapidly increasing as an important natural contaminant in cereals and cereal-derived  
17 foods (Malachova et al., 2011; Vaclavikova et al., 2013). BEA is a cyclic hexadepsipeptide synthesized  
18 by mycotoxigenic fungi, such as *Beauveria bassiana* and *Fusarium* spp. (Wu et al., 2018). This  
19 mycotoxin is harmful to human cells, displaying toxicity at lower concentrations than AflaB<sub>1</sub> (Svingen  
20 et al., 2017). Disturbingly, our findings indicate very elevated levels of BEA (7.9 ± 0.85 ng/g) in washed  
21 quinoa samples from Cajabamba (region of Cajamarca), 150-fold higher than the limit set by EFSA.  
22 Cajabamba is located on the eastern slopes of the Andes (2,648 meters height above mean sea level,  
23 toward the highland jungle); hence, its ecoregion might have favorable conditions for the formation of  
24 mycotoxins by fungi (temperature, pH, moisture, etc.). The contamination of quinoa grains with BEA  
25 originating from Cajamarca represents a point of concern regarding food safety, as local authorities  
26 have recently promoted the cultivation of quinoa in this region (Gobierno Regional de Cajamarca,  
27 2020). Significant amounts of BEA were also detected in canihua samples from Puno (C-Pun), with  
28 quinoa samples from France (WQw-Lauz) and red quinoa samples from Arequipa (RQw-Aqp). We also  
29 compared the relative mean peak area of detected mycotoxin in the untargeted metabolomics  
30 approach. In this sense, asperfumigatin, and altersolanol O are more present in samples for the quinoa  
31 and zearalenone derivative for all kiwicha samples (**Fig 3-B**). In **cluster quinoa**, the samples with more  
32 presence of asperfumigatin were red quinoa from Huaraz, Arequipa and Cajamarca, and white quinoa  
33 from Cajamarca (RQw-Huar, RQw-Aqp, RQw-Cax, WQw-Cax); and for altersolanol O, mainly white  
34 quinoa from Arequipa and Ayacucho, and red quinoa from Puno (WQw-Aqp, and WQw-Aya, RQw-Pun)  
35 (data not showed).

1



2  
3 **Figure 3.** Quantification of BEA mycotoxin and others metabolites by LCMS. **(A)** Bar-plot of BEA amount  
4 (mean + sd, n=3) in Andean pseudocereal samples expressed in ng/g of dried grains. The dotted line  
5 represents the theoretical level of beauvericin among the legislation (0.14 ng/g in dried seeds, EFSA).  
6 **(B)** Bar-plot displaying the relative quantification by the cluster of three molecules produced by fungus  
7 microorganisms in AP. The lowercase letters indicate significant differences between pseudocereal  
8 samples (one-way ANOVA).  
9

### 10 3.3. Fungal inventory infesting the Andean pseudocereals

11 The fungal inventory was realized with the aims to better know the filamentous fungi presents  
12 in the Andean pseudocereal samples and try to correlate this information with the presence of  
13 molecules produced themselves, especially the mycotoxins. Infestation by at least one fungal

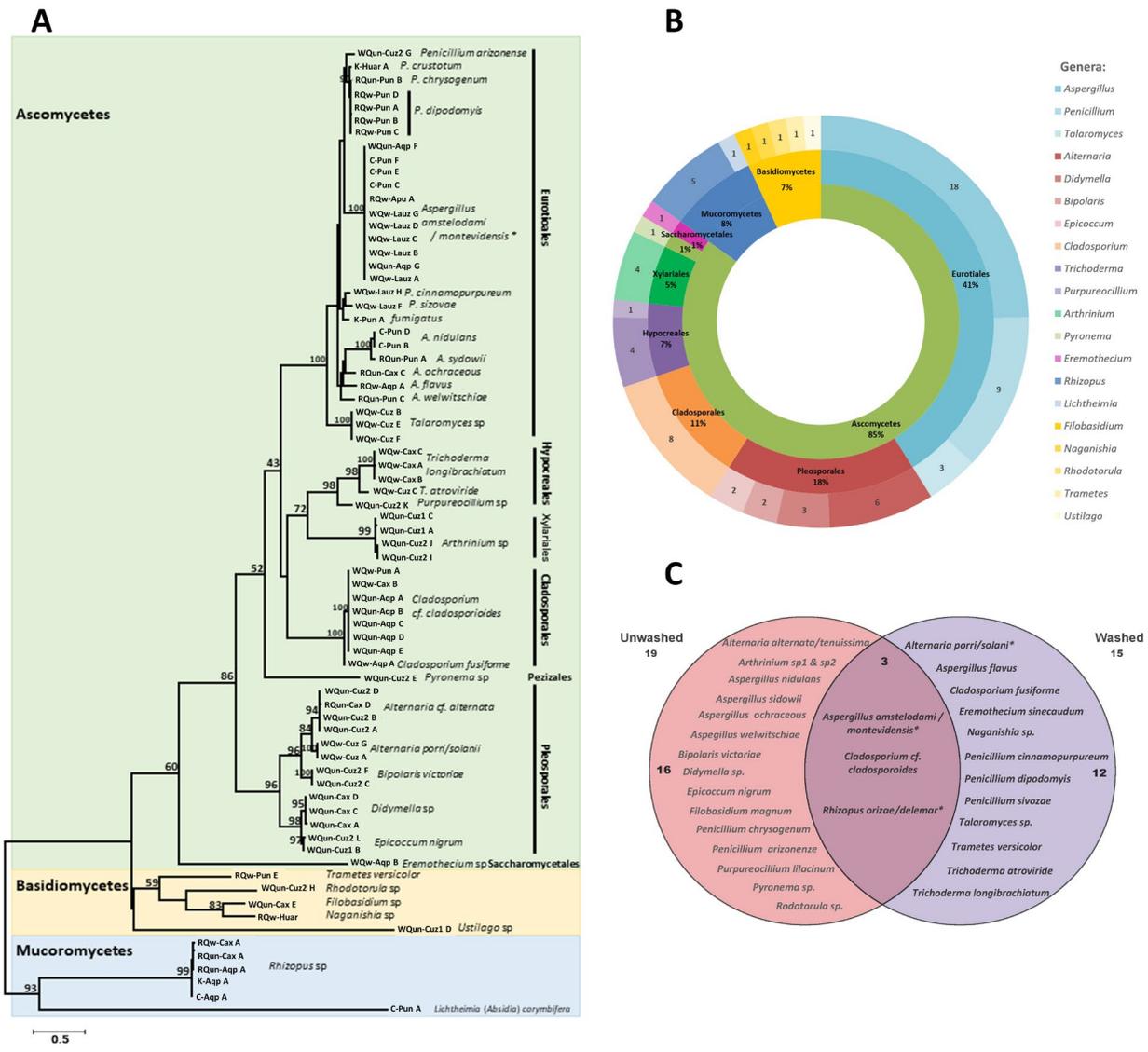
1 endobiont was spotted in 22 up 28 grain samples (78%). Sequence analysis was able to identify 74  
2 fungal isolates (out of 85 total) at the genus level. Overall, we pinpointed 15 fungal families, composed  
3 by 20 genera and 14 species (**supp info 11**). Ascomycetes were the most common phylum of fungi  
4 (84%) (**Fig. 4 A-B**), represented by five orders: i) Eurotioales (*Aspergillus* sp., *Penicillium* sp., and  
5 *Talaromyces* sp.), represents nearly two-third of the strains isolated (58%); ii) Pleosporales (19%),  
6 comprising: *Alternaria* sp., *Bipolaris* sp., *Didymella* sp., and *Epicoccum* sp.; iii) Cladosporales (11%) (only  
7 *Cladosporium* sp.); iv) Hypocreales (6%), and v) Xylariales (6%). In the same way, five Basidiomycetes  
8 and two Mucoromycetes species were spotted. Based on the ITS phylogeny and beta-tubulin and  
9 calmoduline genes analysis, a large diversity of *Aspergillus* sp. and *Penicillium* species could be  
10 observed. However, the precise identification of some others species was not achieved using ITS5  
11 primers and the utilization of genus specific primers (*i.e.* gpd 1 for *Alternaria*) would allow to go  
12 further in identifications (Berbee et al., 1999).

13 Mycotoxigenic fungi have been traditionally classified into two main groups: i) field fungi that  
14 contaminate crops before and during harvest (*e.g.*, *Fusarium*, *Alternaria*, and *Cladosporium* spp.) and,  
15 ii) storage fungi (*e.g.*, *Penicillium* and *Aspergillus* spp.) that tend to contaminate crops in a post-harvest  
16 manner. Overall, the mycotoxigenic fungi we found attached to Andean pseudocereals belonged to  
17 several genera, including *Aspergillus*, *Penicillium*, and *Alternaria*. For instance, *Alternaria* sp. is an  
18 ubiquitous plant pathogen that can infest a broad spectrum of plants, causing damage to crops and  
19 representing a hazard to human health (Troncoso-Rojas & Tiznado-Hernández, 2014). Moreover, some  
20 genera have also the ability to protect their host against phytopathogens, such as the heterogeneous  
21 genus *Cladosporium* able to synthesize metabolites such as cladosporine with antifungal activity on  
22 the spore germination of *Penicillium* and *Aspergillus* species (Bensch et al., 2015; Scott et al., 1971).

23 In our study even if BEA was detected in several samples, our fungal inventory about  
24 pseudocereals did not show the presence of *Fusaria* or *Beauveria*. It could be explained by fact that  
25 mycotoxin may persist while the fungus has disappeared. In addition, the natural production of these  
26 fungal groups is very sensitive to many environmental factors, *i.e.*, altitude, drought, etc. Experimental  
27 limits, mainly the use of a limited number of seeds for each sample leading to non-exhaustive fungal  
28 isolation could also explain the absence of a *Fusarium* community in our inventory. Indeed, in our  
29 experimental condition, others species, better adapted to our experimental design might inhibit the  
30 growth of other strains. Likewise, the presence of the fungus does not prove the presence of  
31 mycotoxins. Some species are known to inhibit the mycotoxins production of other species such as the  
32 post-harvest pathogen *Rhizopus oryzae*, able to degrade mycotoxins like aflatoxins, OTA, or patulin, by  
33 the presence of a specific enzyme (Varga et al., 2005). The strain *Rhizopus* sp. was present  
34 concomitantly in our samples (RQw-Aqp and RQun-Cax) infested by *A. flavus* and *A. cf. ochraceus*,  
35 producers of aflatoxins like AflaB<sub>1</sub> or OTA. Our chemical analysis did not detect AflaB<sub>1</sub> and OTA in these

1 samples, so we hypothesized that *Rhizopus sp.* could have detoxified the sample of the mycotoxins.  
 2 On the other hand, our inventory highlighted the presence of *Alternaria*, *Aspergillus*, and *Cladosporium*  
 3 which could validate the metabolites production of alternanolol O (*Alternaria sp.*), asperfumigatin  
 4 (*Aspergillus fumigatus*), and Pandangolide 1a (*Cladosporium sp.*), respectively.

5 A comparison between fungal species encountered on washed and unwashed quinoa was also  
 6 performed. Nineteen species were isolated on unwashed seeds and 11 on washed samples (**Fig 4-C**).  
 7 Among them, only three species grew in both conditions: *Aspergillus amstelodami/montevideensis*,  
 8 *Cladosporium cf. cladosporoides*, and *Rhizopus sp.* These washing steps have double consequences: on  
 9 the one hand, it can cause the elimination of the fungal strains present on seeds at the time of the  
 10 harvest, and, on the other hand, this phenomenon involving water can allow the emergence of other  
 11 strains being able to develop under favorable conditions.  
 12



13

1 **Figure 4.** Maximum-likelihood phylogenetic tree based on samples ITS sequences and closest ITS  
2 reference sequences from GenBank. The tree was obtained by applying the PhyML method in the  
3 Geneious® platform. Bootstrap values > 50% are indicated above branches. (A) Codes of samples are  
4 described in **supp info 1** and letters were assigned for each strain isolated. (B) Main genus of fungal  
5 community identified from Andean pseudocereals, genera are expressed in number of strains and  
6 percentage. (C) Diagram of Venn for washed and unwashed conditions from quinoa related to their  
7 fungal community. \* : not possible to choose between the 2 species

8

#### 9 **4. Conclusion**

10 Global interest in pseudocereals as "superfood" has increased over the last decade. However, their  
11 chemodiversity has not been well studied to date. In the present work, metabotyping was able to add  
12 new insights into the chemotaxonomy of pseudocereals, confirming the previously established  
13 phylotranscriptomic clades. Metabolomics identified three clusters among *Chenopodium* (**clusters 1**  
14 **and 2**) and *Amaranthus* grain samples (**cluster 3**). In addition, we highlighted their main markers and  
15 major class of compounds by cluster. Moreover, we report for the first time the presence of  
16 mycotoxins in the pseudocereals. Sixteen samples of Peru (out of 27) and one sample from France (out  
17 of one) were contaminated with Beauvericin, an emerging mycotoxin. One sample of white quinoa  
18 from Cajabamba (Cajamarca) showed an alarming high level of Beauvericin. To link the occurrence of  
19 mycotoxins with the presence of toxin-producing fungi in pseudocereals, an inventory of filamentous  
20 fungi was taken. Several mycotoxigenic fungi were detected, including *Aspergillus sp.*, *Penicillium sp.*,  
21 and *Alternaria sp.*, but not *Fusaria*. The present study underscores the need to better understand the  
22 chemistry of pseudocereals, as well as further research into the risk associated with emerging  
23 mycotoxins.

24

#### 25 **Authors contribution**

26 Conceptualization: SB, PGV-O, and GM; methodology: PGV-O, GM, and SB; chemical data  
27 acquisition/curation: PGVO and GM; biological data curation: AG, PJ, and SB; investigation: PGV-O, GM,  
28 AG, SB, GC, JAV-B, SC-Z, PJ, NP, and MH. PGV-O wrote the original draft and all the authors contributed  
29 and validated the final manuscript.

30

#### 31 **Declaration of Competing Interest**

32 The authors declare that they have no known competing financial interests or personal relationships  
33 that could have appeared to influence the work reported in this paper.

34

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10

## 11 References

- 12 Bensch, K., Groenewald, J. Z., Braun, U., Dijksterhuis, J., de Jesús Yáñez-Morales, M., & Crous, P. W.  
13 (2015). Common but different: The expanding realm of *Cladosporium*. *Studies in Mycology*, *82*,  
14 23–74. <https://doi.org/10.1016/j.simyco.2015.10.001>
- 15 Berbee, M. L., Pirseyedi, M., & Hubbard, S. (1999). Cochliobolus Phylogenetics and the Origin of Known,  
16 Highly Virulent Pathogens, Inferred from ITS and Glyceraldehyde-3-Phosphate Dehydrogenase  
17 Gene Sequences. *Mycologia*, *91*(6), 964–977. <https://doi.org/10.2307/3761627>
- 18 Bonifacio, A. (2019). Mejoramiento de la quinua (*Chenopodium quinoa* Wild.) y qañawa (*Chenopodium*  
19 *pallidicaule* Aellen) en un contexto de cambio climático en los Andes altos. *Ciencia e*  
20 *Investigación Agraria*, *46*(2), 113–124. <https://doi.org/10.7764/rcia.v46i2.2146>
- 21 Bräse, S., Encinas, A., Keck, J., & Nising, C. F. (2009). Chemistry and Biology of Mycotoxins and Related  
22 Fungal Metabolites. *Chemical Reviews*, *109*(9), 3903–3990.  
23 <https://doi.org/10.1021/cr050001f>
- 24 Chen, B., Shen, Q., Zhu, X., & Lin, Y. (2014). The Anthraquinone Derivatives from the Fungus *Alternaria*  
25 *sp.* XZSBG-1 from the Saline Lake in Bange, Tibet, China. *Molecules*, *19*(10), 16529–16542.  
26 <https://doi.org/10.3390/molecules191016529>
- 27 Chong, J., Wishart, D. S., & Xia, J. (2019). Using MetaboAnalyst 4.0 for Comprehensive and Integrative  
28 Metabolomics Data Analysis. *Current Protocols in Bioinformatics*, *68*(1), e86.  
29 <https://doi.org/10.1002/cpbi.86>
- 30 Claeys, L., Romano, C., Ruyck, K. D., Wilson, H., Fervers, B., Korenjak, M., Zavadil, J., Gunter, M. J.,  
31 Saeger, S. D., Boevre, M. D., & Huybrechts, I. (2020). Mycotoxin exposure and human cancer  
32 risk: A systematic review of epidemiological studies. *Comprehensive Reviews in Food Science*  
33 *and Food Safety*, *19*(4), 1449–1464. <https://doi.org/10.1111/1541-4337.12567>
- 34 Demšar, J., Curk, T., Erjavec, A., Gorup, Č., Hočevar, T., Milutinovič, M., Možina, M., Polajnar, M.,  
35 Toplak, M., Starič, A., Štajdohar, M., Umek, L., Žagar, L., Žbontar, J., Žitnik, M., & Zupan, B.  
36 (2013). Orange: Data Mining Toolbox in Python. *Journal of Machine Learning Research*, *14*(35),  
37 2349–2353.
- 38 Djoumbou Feunang, Y., Eisner, R., Knox, C., Chepelev, L., Hastings, J., Owen, G., Fahy, E., Steinbeck, C.,  
39 Subramanian, S., Bolton, E., Greiner, R., & Wishart, D. S. (2016). ClassyFire: Automated  
40 chemical classification with a comprehensive, computable taxonomy. *Journal of*  
41 *Cheminformatics*, *8*(1), 61. <https://doi.org/10.1186/s13321-016-0174-y>
- 42 Ducos, C., Pinson-Gadais, L., Chereau, S., Richard-Forget, F., Vásquez-Ocmín, P., Cerapio, J. P.,  
43 Casavilca-Zambrano, S., Ruiz, E., Pineau, P., Bertani, S., & Ponts, N. (2021). Natural Occurrence  
44 of Mycotoxin-Producing *Fusaria* in Market-Bought Peruvian Cereals: A Food Safety Threat for  
45 Andean Populations. *Toxins*, *13*(2), 172. <https://doi.org/10.3390/toxins13020172>

- 1 EFSA. (2014). *Scientific Opinion on the risks to human and animal health related to the presence of*  
2 *beauvericin and enniatins in food and feed | Autorité européenne de sécurité des aliments.*  
3 <https://www.efsa.europa.eu/en/efsajournal/pub/3802>
- 4 Eshelli, M., Qader, M. M., Jambi, J. E., Hursthouse, S. A., & Rateb, E. M. (2018). Current Status and  
5 Future Opportunities of Omics Tools in Mycotoxin Research. *Toxins*, *10*(11).  
6 <https://doi.org/10.3390/toxins10110433>
- 7 Eskola, M., Kos, G., Elliott, C. T., Hajšlová, J., Mayar, S., & Krska, R. (2020). Worldwide contamination  
8 of food-crops with mycotoxins: Validity of the widely cited 'FAO estimate' of 25%. *Critical*  
9 *Reviews in Food Science and Nutrition*, *60*(16), 2773–2789.  
10 <https://doi.org/10.1080/10408398.2019.1658570>
- 11 Fletcher, R. J. (2016). Pseudocereals: Overview. In C. Wrigley, H. Corke, K. Seetharaman, & J. Faubion  
12 (Eds.), *Encyclopedia of Food Grains (Second Edition)* (pp. 274–279). Academic Press.  
13 <https://doi.org/10.1016/B978-0-12-394437-5.00039-5>
- 14 Fraiser-Vannier, O., Chervin, J., Cabanac, G., Puech, V., Fournier, S., Durand, V., Amiel, A., André, O.,  
15 Benamar, O. A., Dumas, B., Tsugawa, H., & Marti, G. (2020). MS-CleanR: A Feature-Filtering  
16 Workflow for Untargeted LC–MS Based Metabolomics. *Analytical Chemistry*, *92*(14), 9971–  
17 9981. <https://doi.org/10.1021/acs.analchem.0c01594>
- 18 Gesner, S., Cohen, N., Ilan, M., Yarden, O., & Carmeli, S. (2005). Pandangolide 1a, a Metabolite of the  
19 Sponge-Associated Fungus *Cladosporium sp.*, and the Absolute Stereochemistry of  
20 Pandangolide 1 and iso-Cladospolide B. *Journal of Natural Products*, *68*(9), 1350–1353.  
21 <https://doi.org/10.1021/np0501583>
- 22 Glass, N. L., & Donaldson, G. C. (1995). Development of primer sets designed for use with the PCR to  
23 amplify conserved genes from filamentous ascomycetes. *Applied and Environmental*  
24 *Microbiology*, *61*(4), 1323–1330.
- 25 Gobierno Regional de Cajamarca. (2020). *GRC impulsa siembra de quinua en Cajabamba.*  
26 [https://www.gob.pe/institucion/regioncajamarca/noticias/187174-grc-impulsa-siembra-de-](https://www.gob.pe/institucion/regioncajamarca/noticias/187174-grc-impulsa-siembra-de-quinua-en-cajabamba)  
27 [quinua-en-cajabamba](https://www.gob.pe/institucion/regioncajamarca/noticias/187174-grc-impulsa-siembra-de-quinua-en-cajabamba)
- 28 Guidugli, F. H., Pestchanker, M. J., De Salmeron, M. S. A., & Giordano, O. S. (1986). 1-  
29 Hydroxyplatyphyllide, a norsesquiterpene lactone from *Senecio gilliesiano*. *Phytochemistry*,  
30 *25*(8), 1923–1926. [https://doi.org/10.1016/S0031-9422\(00\)81175-9](https://doi.org/10.1016/S0031-9422(00)81175-9)
- 31 Guindon, S., & Gascuel, O. (2003). A Simple, Fast, and Accurate Algorithm to Estimate Large  
32 Phylogenies by Maximum Likelihood. *Systematic Biology*, *52*(5), 696–704.  
33 <https://doi.org/10.1080/10635150390235520>
- 34 Hong, S.-B., Go, S.-J., Shin, H.-D., Frisvad, J. C., & Samson, R. A. (2005). Polyphasic taxonomy of  
35 *Aspergillus fumigatus* and related species. *Mycologia*, *97*(6), 1316–1329.  
36 <https://doi.org/10.1080/15572536.2006.11832738>
- 37 ICH. (2005). *Validation of Analytical Procedures: Text and Methodology. International Conference on*  
38 *Harmonization*. IFPMA, Geneva, Switzerland (2005).
- 39 Jarvis, D. E., Ho, Y. S., Lightfoot, D. J., Schmöckel, S. M., Li, B., Borm, T. J. A., Ohyanagi, H., Mineta, K.,  
40 Michell, C. T., Saber, N., Kharbatia, N. M., Rupper, R. R., Sharp, A. R., Dally, N., Boughton, B. A.,  
41 Woo, Y. H., Gao, G., Schijlen, E. G. W. M., Guo, X., ... Tester, M. (2017). The genome of  
42 *Chenopodium quinoa*. *Nature*, *542*(7641), 307–312. <https://doi.org/10.1038/nature21370>
- 43 Kim, H. W., Wang, M., Leber, C. A., Nothias, L.-F., Reher, R., Kang, K. B., van der Hooft, J. J. J., Dorrestein,  
44 P. C., Gerwick, W. H., & Cottrell, G. W. (2021). NPClassifier: A Deep Neural Network-Based  
45 Structural Classification Tool for Natural Products. *Journal of Natural Products*, *84*(11), 2795–  
46 2807. <https://doi.org/10.1021/acs.jnatprod.1c00399>
- 47 Kumar, S., Stecher, G., Li, M., Knyaz, C., & Tamura, K. (2018). MEGA X: Molecular Evolutionary Genetics  
48 Analysis across Computing Platforms. *Molecular Biology and Evolution*, *35*(6), 1547–1549.  
49 <https://doi.org/10.1093/molbev/msy096>
- 50 Lee, H. J., & Ryu, D. (2017). Worldwide Occurrence of Mycotoxins in Cereals and Cereal-Derived Food  
51 Products: Public Health Perspectives of Their Co-occurrence. *Journal of Agricultural and Food*  
52 *Chemistry*, *65*(33), 7034–7051. <https://doi.org/10.1021/acs.jafc.6b04847>

- 1 Lin, M., Han, P., Li, Y., Wang, W., Lai, D., & Zhou, L. (2019). Quinoa Secondary Metabolites and Their  
2 Biological Activities or Functions. *Molecules*, 24(13), 2512.  
3 <https://doi.org/10.3390/molecules24132512>
- 4 Malachova, A., Dzman, Z., Veprikova, Z., Vaclavikova, M., Zachariasova, M., & Hajslova, J. (2011).  
5 Deoxynivalenol, Deoxynivalenol-3-glucoside, and Enniatins: The Major Mycotoxins Found in  
6 Cereal-Based Products on the Czech Market. *Journal of Agricultural and Food Chemistry*,  
7 59(24), 12990–12997. <https://doi.org/10.1021/jf203391x>
- 8 McCartney, N. B., Ahumada, M. I., Muñoz, M., Rosales, M., Fierro, A. M., & Chorbadjian, R. A. (2019).  
9 Effects of saponin-rich quinoa (*Chenopodium quinoa* Willd.) bran and bran extract in diets of  
10 adapted and non-adapted quinoa pests in laboratory bioassays. *International Journal of  
11 Agriculture and Natural Resources*, 46(2), 125–136. <https://doi.org/10.7764/rcia.v46i2.2159>
- 12 Morales-Briones, D. F., Kadereit, G., Tefarikis, D. T., Moore, M. J., Smith, S. A., Brockington, S. F.,  
13 Timoneda, A., Yim, W. C., Cushman, J. C., & Yang, Y. (2020). Disentangling Sources of Gene Tree  
14 Discordance in Phylogenomic Data Sets: Testing Ancient Hybridizations in Amaranthaceae s.l.  
15 *Systematic Biology*, 70(2), 219–235. <https://doi.org/10.1093/sysbio/syaa066>
- 16 Olivon, F., Elie, N., Grelier, G., Roussi, F., Litaudon, M., & Touboul, D. (2018). MetGem Software for the  
17 generation of molecular networks based on the t-SNE algorithm. *Analytical Chemistry*, 90(23),  
18 13900–13908. <https://doi.org/10.1021/acs.analchem.8b03099>
- 19 Ostry, V., Malir, F., Toman, J., & Grosse, Y. (2017). Mycotoxins as human carcinogens—The IARC  
20 Monographs classification. *Mycotoxin Research*, 33(1), 65–73.  
21 <https://doi.org/10.1007/s12550-016-0265-7>
- 22 Pereira, E., Encina-Zelada, C., Barros, L., Gonzales-Barron, U., Cadavez, V., & C.F.R. Ferreira, I. (2019).  
23 Chemical and nutritional characterization of *Chenopodium quinoa* Willd (quinoa) grains: A  
24 good alternative to nutritious food. *Food Chemistry*, 280, 110–114.  
25 <https://doi.org/10.1016/j.foodchem.2018.12.068>
- 26 Perez-Rea, D., & Antezana-Gomez, R. (2018). Chapter 12—The Functionality of Pseudocereal Starches.  
27 In M. Sjöö & L. Nilsson (Eds.), *Starch in Food (Second Edition)* (pp. 509–542). Woodhead  
28 Publishing. <https://doi.org/10.1016/B978-0-08-100868-3.00012-3>
- 29 Rateb, M. E., Houssen, W. E., Harrison, W. T. A., Deng, H., Okoro, C. K., Asenjo, J. A., Andrews, B. A.,  
30 Bull, A. T., Goodfellow, M., Ebel, R., & Jaspars, M. (2011). Diverse Metabolic Profiles of a  
31 *Streptomyces* Strain Isolated from a Hyper-arid Environment. *Journal of Natural Products*,  
32 74(9), 1965–1971. <https://doi.org/10.1021/np200470u>
- 33 Scott, P. M., Walbeek, W. V., & Maclean, W. M. (1971). Cladosporin, a new antifungal metabolite from  
34 *Cladosporium cladosporioides*. *The Journal of Antibiotics*, 24(11), 747–755.  
35 <https://doi.org/10.7164/antibiotics.24.747>
- 36 Shannon, P., Markiel, A., Ozier, O., Baliga, N. S., Wang, J. T., Ramage, D., Amin, N., Schwikowski, B., &  
37 Ideker, T. (2003). Cytoscape: A Software Environment for Integrated Models of Biomolecular  
38 Interaction Networks. *Genome Research*, 13(11), 2498–2504.  
39 <https://doi.org/10.1101/gr.1239303>
- 40 Smith, G. W. (2018). Chapter 71—Fumonisin. In R. C. Gupta (Ed.), *Veterinary Toxicology (Third Edition)*  
41 (pp. 1003–1018). Academic Press. <https://doi.org/10.1016/B978-0-12-811410-0.00071-4>
- 42 Stead, S., Roberts, D., Gledhill, A., de Rijk, T., Van Poucke, C., & De Saeger, S. (2014). *The Development  
43 of a Sensitive Multi-Residue LC-MS/MS Method for the Quantitative Determination of  
44 Mycotoxins in Animal Feedstuffs and Silage Using Xevo TQ-S*. Waters.  
45 <https://www.waters.com/webassets/cms/library/docs/720004961en.pdf>
- 46 Svingen, T., Lund Hansen, N., Taxvig, C., Vinggaard, A. M., Jensen, U., & Have Rasmussen, P. (2017).  
47 Enniatin B and beauvericin are common in Danish cereals and show high hepatotoxicity on a  
48 high-content imaging platform. *Environmental Toxicology*, 32(5), 1658–1664.  
49 <https://doi.org/10.1002/tox.22367>
- 50 Tabatabaei, I., Alosekh, S., Shahid, M., Leniak, E., Wagner, M., Mahmoudi, H., Thushar, S., Fernie, A. R.,  
51 Murphy, K. M., Schmöckel, S. M., Tester, M., Mueller-Roeber, B., Skirycz, A., & Balazadeh, S.

- 1 (2021). *The diversity of quinoa morphological traits and seed metabolic composition* (p.  
2 2021.07.17.452781). bioRxiv. <https://doi.org/10.1101/2021.07.17.452781>
- 3 Troncoso-Rojas, R., & Tiznado-Hernández, M. E. (2014). Chapter 5—*Alternaria alternata* (Black Rot,  
4 Black Spot). In S. Bautista-Baños (Ed.), *Postharvest Decay* (pp. 147–187). Academic Press.  
5 <https://doi.org/10.1016/B978-0-12-411552-1.00005-3>
- 6 Tsugawa, H., Arita, M., Kanazawa, M., Ogiwara, A., Bamba, T., & Fukusaki, E. (2013). MRMPROBS: A  
7 Data Assessment and Metabolite Identification Tool for Large-Scale Multiple Reaction  
8 Monitoring Based Widely Targeted Metabolomics. *Analytical Chemistry*, *85*(10), 5191–5199.  
9 <https://doi.org/10.1021/ac400515s>
- 10 Tsugawa, H., Cajka, T., Kind, T., Ma, Y., Higgins, B., Ikeda, K., Kanazawa, M., VanderGheynst, J., Fiehn,  
11 O., & Arita, M. (2015). MS-DIAL: Data-independent MS/MS deconvolution for comprehensive  
12 metabolome analysis. *Nature Methods*, *12*(6), 523–526. <https://doi.org/10.1038/nmeth.3393>
- 13 Tsugawa, H., Kind, T., Nakabayashi, R., Yukihiro, D., Tanaka, W., Cajka, T., Saito, K., Fiehn, O., & Arita,  
14 M. (2016). Hydrogen Rearrangement Rules: Computational MS/MS Fragmentation and  
15 Structure Elucidation Using MS-FINDER Software. *Analytical Chemistry*, *88*(16), 7946–7958.  
16 <https://doi.org/10.1021/acs.analchem.6b00770>
- 17 Tsugawa, H., Ohta, E., Izumi, Y., Ogiwara, A., Yukihiro, D., Bamba, T., Fukusaki, E., & Arita, M. (2015).  
18 MRM-DIFF: Data processing strategy for differential analysis in large scale MRM-based  
19 lipidomics studies. *Frontiers in Genetics*, *5*. <https://doi.org/10.3389/fgene.2014.00471>
- 20 Vaclavikova, M., Malachova, A., Veprikova, Z., Dzuman, Z., Zachariasova, M., & Hajslova, J. (2013).  
21 ‘Emerging’ mycotoxins in cereals processing chains: Changes of enniatins during beer and  
22 bread making. *Food Chemistry*, *136*(2), 750–757.  
23 <https://doi.org/10.1016/j.foodchem.2012.08.031>
- 24 Varga, J., Péteri, Z., Tábori, K., Téren, J., & Vágvolgyi, C. (2005). Degradation of ochratoxin A and other  
25 mycotoxins by *Rhizopus* isolates. *International Journal of Food Microbiology*, *99*(3), 321–328.  
26 <https://doi.org/10.1016/j.ijfoodmicro.2004.10.034>
- 27 Vásquez-Ocmín, P. G., Marti, G., Bonhomme, M., Mathis, F., Fournier, S., Bertani, S., & Maciuk, A.  
28 (2021). Cannabinoids vs. whole metabolome: Relevance of cannabinomics in analyzing  
29 *Cannabis* varieties. *Analytica Chimica Acta*, *1184*, 339020.  
30 <https://doi.org/10.1016/j.aca.2021.339020>
- 31 White, T., Bruns, T., Lee, S., & Taylor, J. (1990). Amplification and direct sequencing of fungal ribosomal  
32 RNA genes for phylogenetics. In *PCR Protocols: A Guide to Methods and Applications*.  
33 Academic Press, In. <http://pdf.xuebalib.com:1262/3x0d5gC6z4eF.pdf>
- 34 Wu, Q., Patocka, J., Nepovimova, E., & Kuca, K. (2018). A Review on the Synthesis and Bioactivity  
35 Aspects of Beauvericin, a *Fusarium* Mycotoxin. *Frontiers in Pharmacology*, *9*.  
36 <https://doi.org/10.3389/fphar.2018.01338>
- 37 Xie, F., Li, X.-B., Zhou, J.-C., Xu, Q.-Q., Wang, X.-N., Yuan, H.-Q., & Lou, H.-X. (2015). Secondary  
38 Metabolites from *Aspergillus fumigatus*, an Endophytic Fungus from the Liverwort  
39 *Heteroscyphus tener* (Steph.) Schiffn. *Chemistry & Biodiversity*, *12*(9), 1313–1321.  
40 <https://doi.org/10.1002/cbdv.201400317>
- 41