AGRICULTURAL AND FOOD CHEMISTRY

Metabolomic Analysis of *Ranunculus* spp. as Potential Agents Involved in the Etiology of Equine Grass Sickness

Johanna Michl,⁺ Maryam Modarai,⁺ Sarah Edwards,⁺ and Michael Heinrich^{*,+,§}

⁺Centre for Pharmacognosy and Phytotherapy, The School of Pharmacy, University of London, 29-39 Brunswick Square, London WC1N 1AX, United Kingdom

⁹Southern Cross Plant Science, Southern Cross University, P.O. Box 157, Lismore, NSW 2480, Australia

Supporting Information

ABSTRACT: Identification of toxic or harmful agents continues to be a key goal in agricultural chemistry. This paper reports a metabolomic analysis of *Ranunculus repens* and related species, which were recently postulated to be cocausative agents in the etiology of equine grass sickness (EGS). Specifically, samples collected at EGS sites were compared with those from non-EGS sites. Furthermore, interspecific and seasonal variations and the species' response to edaphic and climatic factors were investigated. ¹H NMR spectroscopy in combination with multivariate data analysis was applied to the crude methanol extracts of the *Ranunculus* samples, as well as their chloroform fractions. Samples from EGS sites were significantly different from control samples. The metabolite composition varied greatly between different *Ranunculus* species. No significant changes could be observed between samples collected in different seasons. This work provides strong evidence that *Ranunculus* is involved in the etiology of EGS and has implications for agricultural management of pastures.

KEYWORDS: *Ranunculus* (buttercups), equine grass sickness, metabolomics, principal component analysis, partial least-squares, ¹H NMR

INTRODUCTION

Equine grass sickness (EGS) or equine dysautonomia is a degenerative polyneuropathy affecting horses mainly in Great Britain.¹ It is very likely that EGS has a multifactorial etiology, including intoxication with *Clostridium botulinum* type C.² As the name suggests, EGS is associated with grazing.¹ Recent research in our group has shown that a range of edaphic and botanical factors are strongly associated with EGS outbreaks including the regular occurrence of *Ranunculus* spp. on EGS sites.³ Fresh *Ranunculus* contains ranunculin, a glycoside that is enzymatically hydrolyzed to produce protoanemonin,⁴ a toxic metabolite that can cause blistering⁵ and gastrointestinal irritation⁶ (Supporting Information, Figure 1S).

The initial hypothesis was that ranunculin levels of *Ranunculus* samples collected from EGS sites are higher compared to those from control sites. HPLC quantification was therefore carried out and showed that ranunculin levels in samples from EGS sites were indeed significantly higher.³

However, confining the analysis to just a few metabolites to confirm or deny if *Ranunculus* species are involved in EGS would have been misleading, because the role of *Ranunculus* as a causal agent of this disease is likely to be more complex.

NMR metabolomics offers a more suitable approach, because it can be used to compare the whole plant metabolome of samples from EGS and non-EGS sites. It also allows taking into consideration interspecific variation and variation caused by environmental changes.

Metabolic profiling has been proposed as a method for the analysis of phytomedicines,⁷⁻¹⁰ and food products ^{11,12} and it also has been used widely for understanding toxin- and stress-induced metabolomic changes¹³⁻¹⁵ in mammals. A number of

studies show how plants respond to their environment, for example, neighboring species,¹⁶ temperature,¹⁷ salt,¹⁸ phosphorus,¹⁹ and cadmium.²⁰ However, these studies are usually carried out under extreme conditions created artificially in the laboratory, whereas the influence of naturally occurring changes such as the changing of seasons or physiological ion levels in the soil has not been studied.

The aim of this work is to investigate (a) the differences between EGS and non-EGS samples using multivariate statistical methods such as principal component analysis (PCA) or partial least-squares—discriminant analysis (PLS-DA) and (b) the differences between *Ranunculus* species and seasonal changes in the metabolite composition by using PCA as well as (c) metabolic changes triggered by soil minerals (PLS).

MATERIALS AND METHODS

Chemicals. All chemical reagents used were of analytical grade. Deuterated chloroform (CDCl₃, 99.8%) and deuterated methanol (MeOD, 99.8%) were obtained from Cambridge Isotope Laboratories, Inc. (Andover, MA). 3-(Trimethylsilyl)propionic-2,2,3,3- d_4 acid, sodium salt (TSP, 98%), and tetramethylsilane (TMS, 99.9%) were obtained from Sigma-Aldrich (St. Louis, MO).

Plant Material. Between April 2007 and March 2009 *Ranunculus* spp. samples were collected from 9 control sites and 12 EGA sites and identified by Prof. Michael Heinrich and Dr. Sarah Edwards. Furthermore, over the course of a year 10 samples were collected from one site

Received:	April 11, 2011
Revised:	August 15, 2011
Accepted:	August 15, 2011
Published:	August 15, 2011



Table 1. Information on the Samples Collected on EGS (E) and Control Sites (C): Species, Number of EGS Cases Reported, Date of Collection, and County (see also Edwards et al.³).

	species	EGS cases	collection	county
E1	R. acris	1	May 15, 2007	Stirlingshire
E2	R. acris	1	May 15, 2007	Stirlingshire
E3	R. repens	2	May 17, 2007	Moray
E4	R. repens	5	May 25, 2007	Northamptonshire
E5	R. repens	5	May 25, 2007	Northamptonshire
E6	R. acris	4	June 5, 2007	Lincolnshire
E7	R. repens	1	June 12, 2007	Hampshire
E8	R. acris	4	July 30, 2007	Hertfordshire
E9	R. repens	4	Aug 20, 2007	Buckinghamshire
E10	R. bulbosus	4	Aug 20, 2007	Buckinghamshire
E11	R. repens	2	Dec 13, 2007	Lancashire
E12	R. repens	2	Jan 29, 2008	Powys
E13	R. bulbosus	2	May 15, 2008	Hertfordshire
E14	R. bulbosus	2	May 15, 2008	Hertfordshire
E15	R. repens	2	May 15, 2008	Hertfordshire
E16	R. repens	1	May 19, 2008	Berkshire
E17	R. repens	1	May 19, 2008	Berkshire
E18	R. repens	1	July 28, 2008	Suffolk
C1	R. acris	0	Nov 28, 2007	London
C2	R. repens	0	June 11, 2007	Oxfordshire
C3	R. acris	0	June 11, 2007	Oxfordshire
C4	R. repens	0	July 8, 2007	Northamptonshire
C5	R. repens	0	July 15, 2007	Angus
C6	R. repens	0	July 23, 2008	Berkshire
C7	R. acris	0	July 23, 2008	Berkshire
C8	R. acris	0	July 23, 2008	Oxfordshire
C9	R. repens	0	July 28, 2008	Buckinghamshire
C10	R. repens	0	July 28, 2008	Hertfordshire
C11	R. acris	0	July 28, 2008	Buckinghamshire
C12	R. repens	0	Aug 29, 2008	Buckinghamshire

in Oxfordshire. In total, 30 samples were collected. Eighteen were identified as *Ranunculus repens* L., 9 as *Ranunculus acris* L., and 3 as *Ranunculus bulbosus* L. (Ranunculaceae; Table 1). All sites are located in Great Britain (Figure 1).

Extraction. Ground plant material was extracted in a glass column with methanol for 6 h. The extracts were evaporated to dryness below 40 $^{\circ}$ C. Ten percent of the crude methanol extract was dried under nitrogen and used for ¹H NMR measurements.

Ninety percent of the crude methanol extract was partitioned between 50 mL of 90% methanol and 50 mL of hexane three times. The hexane fractions were discarded. Water was added to the methanol fraction to obtain 70% methanol. The 70% methanol fraction was partitioned with 50 mL of chloroform three times. The chloroform fractions were evaporated to dryness below 40 $^{\circ}$ C, dried under nitrogen, and used for ¹H NMR measurements. The extraction was made in triplicates for each sample.

NMR Experiments. Samples for NMR analysis were prepared by dissolving the methanol extracts in MeOD containing 0.01% TSP and the chloroform extracts in CDCl₃ containing 0.01% TMS to a concentration of 10 mg/mL. The solutions were transferred into 5 mm NMR tubes. ¹H NMR spectra were recorded on a Bruker Avance 400 spectrometer operating at 400.13 MHz. For each sample 128 transients were recorded as 65K data points with a spectral width of 8278 Hz using



Figure 1. Locations of EGS (\blacksquare) and control sites (\bigcirc).

3.96 s acquisition time, 1.0 s relaxation delay, and a 30° pulse angle. FIDs were Fourier transformed with LB = 0.3 Hz. The spectra were referenced to the internal TMS or TSP. ¹H NMR spectra were manually corrected for phase and baseline distortions using Topspin (v 1.3., Bruker Biospin).

Data Reduction and Statistical Analysis. The spectra in the range of 0.2-10.5 ppm were divided into 258 regions ("buckets") of 0.04 ppm using AMIX software (v. 3.5.5., Bruker Biospin), and the signal intensity in each region was integrated. This bucket size was chosen to allow the small signal shift compensation that occurred. Water signals (4.6-5.0 ppm in MeOD spectra) and residual proton signals corresponding to MeOD (3.24-3.4 ppm) and CDCl₃ (7.2-7.4 ppm) were excluded. The data could then be imported into Microsoft Excel for the addition of labels and normalization. The spectral areas were normalized to the total sum of the spectra integral to avoid dilution effects. PCA, PLS-DA, and PLS were carried out on the normalized and centered NMR data set using the software SIMCA P+ (v. 12, Umetrics, Umea, Sweden). PCA is an unsupervised pattern recognition technique,²¹ which can be used to reduce the complexity of a data set and visualize the data in two- or three-dimensional plots. PLS is a regression technique for relating the measured data (X) and the response variable (Y) by a linear multivariate model.²² PLS with discriminant analysis (PLS-DA) is



Figure 2. Typical ¹H NMR spectra of methanol and chloroform extracts. Highlighted peaks correspond to resonance signals for ranunculin aglycon.⁶

an analysis useful in classification, where the *Y* variable is chosen to represent the class membership.²³ Validation of the models was performed using full cross-validation and permutation tests (200 repeats). The quality of all PCA, PLS-DA, and PLS models was measured using the correlation coefficient R^2 and a cross-validated correlation coefficient Q^2 (goodness of prediction).²² Hotelling's T2 region is shown as an ellipse in the scores plots. It defines the 95% confidence interval of the modeled variation.²⁴

RESULTS AND DISCUSSION

Identification of Metabolites in the Extracts. ¹H NMR spectra of methanol and chloroform extracts were inspected visually. Well-defined signals were present in the aromatic region of the spectra. The following peaks corresponding to ranunculin aglycon could be found in the ¹H NMR spectra of methanol extracts: 7.67 (H-3, dd, J = 1.51, 5.77 Hz), 6.19 (H-2, dd, J = 2.01, 5.75 Hz), and 5.16 (H-4, m) ppm. With regard to the chloroform extracts, the signals shifted from 7.67 to 7.49 ppm (H-3, dd, J = 1.51, 5.77 Hz) and from 6.19 to 6.22 ppm (H-2, dd, *J* = 2.01, 5.75 Hz), respectively. Additional peaks could be identified at 4.00 (H-5A, dd) and 3.80 (H-5B, dd) ppm (Figure 2). The signals were assigned to the resonance peaks of ranunculin aglycon by comparison with data of the literature.⁶ The ranunculin aglycon was found to be the major ranunculin metabolite in all extracts. Freeze-drying of the extracts led to inactivation of the enzymes responsible for the biotransformation of ranunculin into its toxic metabolite protoanemonin (Supporting Information, Figure 1S).



Figure 3. Scores plot of a two-component PCA model of the ¹H NMR spectral data of all methanol extracts (δ 4.6–5.0 and 3.24–3.5 regions excluded, 0.04 ppm buckets, data mean-centered): \Box , *R. bulbosus*; \blacktriangle , *R. acris*; +, *R. repens*.

Therefore, it is likely that the main metabolite to be found in the plant extracts is either ranunculin or its aglycon. Ranunculin was present in only trace amounts in some of the samples and had spontaneously degraded into its aglycon due to long storage times.⁶

Furthermore, the ¹H NMR spectra of the methanol extracts contained a variety of primary metabolites such as α -glucose (5.10 ppm, d, J = 3.66 Hz), β -glucose (4.46 ppm d, J = 9.82 Hz), and protons from fatty acids: 5.34 ppm (-CH=CH-groups), 2.31 ppm (=CHCH2-CH=), 2.07 ppm (-H2C-CH=CH-), 1.59 ppm (-CH2-CH2-COOH), 1.31 ppm (CH₂ protons), and 0.9 ppm (CH₃ protons). The chloroform extracts, on the other hand, contained mainly ranunculin aglycon and fatty acids.

Metabolomic Classification of Ranunculus Species Using PCA. For the data set consisting of all samples (EGS, control, and 10 samples from the same Oxfordshire control site) a twocomponent model explained 67% of the variance. As observed in Figure 3, there is a discrimination possible between R. repens and R. acris and R. bulbosus by PC2. The extracts of R. acris and R. bulbosus form a cluster in the lower half of the scores plot, whereas most R. repens samples are located in the upper half of the scores plot. The loadings plot (data not shown) shows that the most influential chemical shifts are in the lipid and sugar regions. Differentiation between R. repens and the other two species was caused by higher levels of β -glucose, fatty acids, and an unknown aromatic compound (7.12 and 6.08 ppm in R. repens samples and higher amounts of α -glucose (5.09 ppm) and other unidentified compounds in the sugar region (4.32, 3.64, and 2.52 ppm) for R. acris and R. bulbosus samples. However, ranunculin aglycon chemical shifts are not among the important influences that can be identified in the loadings plot. An ANOVA analysis confirmed that differences in the relative signal intensities for ranunculin aglycon between the three Ranunculus species were not statistically significant.

Differentiation between EGS and Control Samples Using PCA and PLS-DA. Due to the observed interspecies variation, only *R. repens* samples were used for determining if there is a significant difference between EGS and control site samples. Those collected from the same Oxfordshire control site over the year were also excluded to balance the number of EGS and control samples and to avoid bias caused by the fact that they are likely to have similar properties. A two-component PCA model based on all *R. repens* EGS and control samples explained 78% of the variance. As observed in Figure 4a, all control samples with

а

ည် 0.000

0.002

0.00

-0.001

-0.002

-0.006





Figure 4. Scores plots of PCA (a) and PLS-DA (b) models of the ¹H NMR spectral data of *R. repens* methanol extracts (δ 4.6–5.0 and 3.24–3.5 regions excluded, 0.04 ppm buckets, data mean-centered): \bigcirc , control sites; \blacksquare , EGS sites.

the exception of the three replicates of sample C2 form one group, whereas there appears to be much larger variation within the EGS samples. With regard to the loadings plot, the differentiation between the two groups is caused by fatty acids, glucose, and other carbohydrates, whereas the ranunculin aglycon does not contribute to the variability. PCA of the chloroform extracts showed a similar result (data not shown). However, ranunculin aglycon could be identified as one of the important influences in the PC1 loadings plot of the chloroform extracts. Positive loadings for ranunculin aglycon corresponded to EGS samples, indicating that the ranunculin aglycon content in EGS samples is higher. The negative loadings corresponding to control samples could be assigned as protons from fatty acids.

In addition to PCA, PLS-DA was carried out. The *Y* variable was defined as 1 for EGS samples and 0 for control samples. The PLS-DA scores plot for the methanol extracts show a clear discrimination between EGS and control samples (with the exception of the three replicates of sample C2) (Figure 4b). The correlation coefficient (R^2Y) for the three-component PLS-DA model was 0.65, with a Q^2 of 0.52. Furthermore, permutation tests were performed to validate the PLS-DA model. All R^2 and Q^2 values were lower than for the real model, indicating good predictability (Supporting Information, Figure 2S). The PLS-DA coefficients plot showed positive coefficients for ranunculin aglycon, as well as α -glucose and protons corresponding to CH₂ groups of fatty acids (Figure 5).

To confirm whether EGS samples contained higher contents of ranunculin aglycon or not, we carried out a relative quantification by comparing the signal intensities for spectral region of the H-3 proton of ranunculin aglycon in the methanol extracts of all EGS and control samples. Ranunculin aglycon levels were



Figure 5. Coefficients plot of a three-component PLS-DA model of the ¹H NMR spectral data of *R. repens* methanol extracts (δ 4.6–5.0 and 3.24–3.5 regions excluded, 0.04 ppm buckets, data mean-centered). Highlighted peaks correspond to resonance signals for ranunculin aglycon.⁶

 0.33 ± 0.19 for EGS samples, which is significantly higher (p < 0.01) than $0.18 \pm 0.11\%$ of the total integral for the control samples (Supporting Information, Figure 3S).

Impact of Environmental Factors on the Metabolite Profile. The chemical composition of plants can be influenced by many environmental factors, such as climatic conditions and the properties of the soil they grow on.

To investigate if seasonal changes have an impact on the chemical composition of the plant material, 10 samples were collected on one site at different times of the year. PCA was applied to their spectral data to see if groupings of samples collected in the same season could be observed in the scores plot. The results showed that there were no significant differences between samples collected in different seasons. The seasonal changes in the plants' metabolome were much lower than expected.

Previous work in our group³ suggests that soil nitrate levels at EGS sites are significantly higher than at control sites. Therefore, we wanted to investigate whether the levels of nitrate and other minerals present in the soil might have an impact on the metabolite composition of *Ranunculus*.

PLS was used to determine if there is a correlation between the content of soil minerals and the metabolite composition of the Ranunculus samples. The Y variable was defined as the level of minerals (nitrate nitrogen, ammonium nitrogen, phosphorus, potassium, magnesium, copper, zinc, total iron, available iron, lead, arsenic, cadmium, mercury, nickel, selenium, and fluoride) in soil samples collected at the same sites.³ Low R^2 and Q^2 values of the PLS models indicate that the soil mineral content does not influence the metabolite composition of the Ranunculus samples. Whereas the correlation with nitrate levels in the soil is low, higher percentages of the variation in the methanol extracts can be explained by the variation in levels of available iron, arsenic, selenium, fluoride, and molybdenum (Figure 6). Interestingly, a positive correlation between ranunculin aglycon signals and available iron levels could be found, indicating that high available iron levels in the soil might trigger the biosynthesis of ranunculin (Supporting Information, Figure 4S).

We had previously postulated that the well-known irritant effect from eating *Ranunculus* combined with the consumption of



Figure 6. R^2 and Q^2 values of five-component PLS models showing the correlation between ¹H NMR spectral data of *R. repens* methanol extracts (δ 4.6–5.0 and 3.24–3.5 regions excluded, 0.04 ppm buckets, data mean-centered) and levels of minerals in soil samples collected at the same sites.

herbage containing high amounts of heavy metals such as iron³ or herbage with reduced antioxidant activity¹ might cause severe imbalances in a horse's gastrointestinal tract and therefore are likely to be responsible for the occurrence of EGS.

The results of this study support the above hypothesis that *Ranunculus* is involved in the etiology of EGS, but they also indicate that the content of toxic lactones is not the only difference between EGS and non-EGS samples. ¹H NMR spectroscopy in combination with multivariate data analysis was able to show a clear discrimination between EGS and non-EGS samples. This work also highlights the importance of PCA and PLS for discriminating between different plant species and for understanding the implication of environmental factors on the plants' metabolome. It is evident that EGS has a very complex and multifactorial etiology. Therefore, it is unlikely that ingestion of *Ranunculus* alone can directly cause EGS. However, our data suggest that *Ranunculus* is an important factor in EGS development and is likely to be a cocausative environmental agent of this poorly understood disease.

ASSOCIATED CONTENT

Supporting Information. Biotransformation of ranunculin, PLS-DA permutation tests, ranunculin aglycon levels, and PLS coefficients plot for available iron. This material is available free of charge via the Internet at http://pubs.acs.org.

AUTHOR INFORMATION

Corresponding Author

*E-mail: michael.heinrich@pharmacy.ac.uk or michael.heinrich@ scu.edu.au. Phone: +61-2-6626-3010. Fax: +61-2-66223459.

ACKNOWLEDGMENT

We are grateful to Patricia Brown for a generous donation to the Centre for Pharmacognosy and Phytotherapy, School of Pharmacy, University of London, which made this research possible. We are most grateful to all horse/pony owners and landowners who have kindly contributed to this study. We are also thankful to Prof. Brigitte Kopp, for the supervision of J.M. at the University of Vienna.

ABBREVIATIONS USED

C, control samples; CDCl₃, deuterated chloroform; E, equine grass sickness samples; EGS, equine grass sickness; FID, free induction decay; LB, line broadening; MeOD, deuterated methanol; PCA, principal component analysis; PC, principal component; PLS, partial least-squares; PLS-DA, partial least-squares—discriminant analysis; TMS, tetramethylsilane; TSP, 3-(trimethylsilyl)-propionic-2,2,3,3- d_4 acid, sodium salt.

REFERENCES

(1) McGorum, B. C.; Fry, S. C.; Wallace, G.; Coenen, K.; Robb, J.; Williamson, G.; Aruoma, O. I. Properties of herbage in relation to equine dysautonomia: biochemical composition and antioxidant and prooxidant actions. *J. Agric. Food Chem.* **2000**, *48*, 2346–2352.

(2) McCarthy, H. E.; French, N. P.; Edwards, G. B.; Poxton, I. R.; Kelly, D. F.; Payne-Johnson, C. E.; Miller, K.; Proudman, C. J. Equine grass sickness is associated with low antibody levels to *Clostridium botulinum*: a matched case-control study. *Equine Vet. J.* 2004, 36, 123–129.

(3) Edwards, S. E.; Martz, K. E.; Rogge, A.; Heinrich, M. Edaphic and phytochemical factors as predictors of equine grass sickness cases in the UK. *Front. Pharmacol.* **2010**, *2*, doi: 10.3389/fphar.2010.00122.

(4) Hill, R.; Van Heyningen, R. Ranunculin; the precursor of the vesicant substance of the buttercup. *Biochem. J.* **1951**, *49*, 332–335.

(5) Bonora, A.; Dall'olio, G.; Bruni, A. Separation and quantitation of protoanemonin in Ranunculaceae by normal- and reversed-phase HPLC. *Planta Med.* **1985**, *51*, 364–367.

(6) Bai, Y. L.; Benn, M. H.; Majak, W.; McDiarmid, R. Extraction and HPLC determination of ranunculin in species of the buttercup family. *J. Agric. Food Chem.* **1996**, *44*, 2235–2238.

(7) Agnolet, S.; Jaroszewski, J. W.; Verpoorte, R.; Staerk, D. ¹H NMR-based metabolomics combined with HPLC-PDA-MS-SPE-NMR for investigation of standardized *Ginkgo biloba* preparations. *Metabolomics* **2010**, *6*, 292–302.

(8) Kim, H. K.; Choi, Y. H.; Erkelens, C.; Lefeber, A. W.; Verpoorte, R. Metabolic fingerprinting of *Ephedra* species using ¹H-NMR spectroscopy and principal component analysis. *Chem. Pharm. Bull.* **2005**, *53*, 105–109.

(9) Frederich, M.; Choi, Y. H.; Angenot, L.; Harnischfeger, G.; Lefeber, A. W.; Verpoorte, R. Metabolomic analysis of *Strychnos nux-vomica, Strychnos icaja* and *Strychnos ignatii* extracts by ¹H nuclear magnetic resonance spectrometry and multivariate analysis techniques. *Phytochemistry* **2004**, *65*, 1993–2001.

(10) Modarai, M.; Yang, M.; Suter, A.; Kortenkamp, A.; Heinrich, M. Metabolomic profiling of liquid *Echinacea* medicinal products with *in vitro* inhibitory effects on cytochrome P450 3A4 (CYP3A4). Planta Med. **2010**, *76*, 378–385.

(11) Capanoglu, E.; Beekwilder, J.; Boyacioglu, D.; Hall, R.; de Vos, R. Changes in antioxidant and metabolite profiles during production of tomato paste. *J. Agric. Food Chem.* **2008**, *56*, 964–973.

(12) Cavaliere, B.; De Nino, A.; Hayet, F.; Lazez, A.; Macchione, B.; Moncef, C.; Perri, E.; Sindona, G.; Tagarelli, A. A metabolomic approach to the evaluation of the origin of extra virgin olive oil: a convenient statistical treatment of mass spectrometric analytical data. *J. Agric. Food Chem.* **2007**, *55*, 1454–1462.

(13) Galindo, F. G.; Dejmek, P.; Lundgren, K.; Rasmusson, A. G.; Vicente, A.; Moritz, T. Metabolomic evaluation of pulsed electric field-induced stress on potato tissue. *Planta* **2009**, *230*, 469–479.

(14) Yap, I. K.; Clayton, T. A.; Tang, H.; Everett, J. R.; Hanton, G.; Provost, J. P.; Le Net, J. L.; Charuel, C.; Lindon, J. C.; Nicholson, J. K. An integrated metabonomic approach to describe temporal metabolic disregulation induced in the rat by the model hepatotoxin allyl formate. *J. Proteome Res.* **2006**, *5*, 2675–2684. (15) Wang, Y.; Holmes, E.; Tang, H.; Lindon, J. C.; Sprenger, N.; Turini, M. E.; Bergonzelli, G.; Fay, L. B.; Kochar, S.; Nocholson, J. K. Experimental metabonomic model of dietary variation and stress interactions. *J. Proteome Res.* **2006**, *5*, 1535–1542.

(16) Broz, A. K.; Broeckling, C. D.; De-la-Pena, C.; Lewis, M. R.; Greene, E.; Callaway, R. M.; Sumner, L. W.; Vivanco, J. M. Plant neighbor identity influences plant biochemistry and physiology related to defense. *BMC Plant Biol.* **2010**, *10*, 115.

(17) Kaplan, F.; Kopka, J.; Haskell, D. W.; Zhao, W.; Schiller, K. C.; Gatzke, N.; Sung, D. Y.; Guy, C. L. Exploring the temperature-stress metabolome of *Arabidopsis*. *Plant Physiol.* **2004**, *136*, 4159–4168.

(18) Johnson, H. E.; Broadhurst, D.; Goodacre, R.; Smith, A. R. Metabolic fingerprinting of salt-stressed tomatoes. *Phytochemistry* **2003**, *62*, 919–928.

(19) Hernandez, G.; Ramirez, M.; Valdes-Lopez, O.; Tesfaye, M.; Graham, M. A.; Czechowski, T. Phosphorus stress in common bean: root transcript and metabolic responses. *Plant Physiol.* **2007**, 144, 752–767.

(20) Bailey, N. J.; Oven, M.; Holmes, E.; Nicholson, J. K.; Zenk, M. H. Metabolomic analysis of the consequences of cadmium exposure in *Silene cucubalus* cell cultures via ¹H NMR spectroscopy and chemometrics. *Phytochemistry.* **2003**, *62*, 851–858.

(21) Trygg, J.; Lundstedt, T. Chemometrics techniques for metabonomics. *The Handbook of Metabonomics and Metabolomics*, 1st ed.; Lindon, J. C., Nicholson, J. K., Holmes, E., Eds.; Elsevier Science: Amsterdam, The Netherlands, 2007; pp 171–199.

(22) Wold, S.; Sjöström, M.; Eriksson, L. PLS regression: a basic tool of chemometrics. *Chemom. Intell. Lab.* **2001**, *58*, 109–130.

(23) Westerhuis, J. A.; Hoefsloot, H. C.; Smit, S.; Vis, D. J.; Smilde, A. K.; van Velzen, E. J.; van Duijnhoven, J. P.; van Dorseten, F. A. Assessment of PLSDA cross validation. *Metabolomics* **2008**, *4*, 81–89.

(24) Lundstedt, T.; Trygg, J.; Holmes, E. Chemometrics in metabonomics. J. Proteome Res. 2007, 6, 469–479.