A Simplified, Integrated Solution for Untargeted Metabolomics

Routine Discovery Workflows with the X500R QTOF System

Baljit K. Ubhi¹ and Timothy J. Garrett²
¹SCIEX, CA, USA, ²University of Florida, FL, USA

Metabolomics focuses on the chemical processes central to cellular metabolism, giving scientists a snapshot of the cellular metabolic state, or metabolome at a given time. It consists of intermediates, conjugates, hormones, and even exogenous metabolites from food or the environment that have an impact on health. Mass spectrometry is the tool of choice for the measurements of these metabolites. However, they can be increasingly challenging workflows to setup for the novice metabolomics researcher or the technician. Therefore, a robust solution for screening metabolomics samples is of increased interest and the need for a more integrated and routine mass spectrometry system is in demand. The new X500R QTOF System was developed for routine, robust workflows and requires minimal MS expertise. The system integrates all data acquisition; processing and review as well as reporting into a single piece of easy to use, easy to learn software.

Cancer research represents a key area where metabolomics can provide new biomarkers of disease from easy to obtain biofluids such as urine or plasma. An untargeted screening approach using mass spectrometry that can be adopted by a broader range of research labs (not only the expert labs) would enable this valuable research to be more broadly performed. Here a small pilot study using a urine sample set from a previous prostate cancer study was used to determine whether the untargeted metabolomics workflow using the X500R System could find key differences between the samples.

Key Benefits of the Untargeted Metabolomics Workflow on the X500R QTOF System

1. Simplified operation with new interface and workflow setup powered by SCIEX OS Software with minimal MS expertise.
2. Get to actionable results quickly with user-friendly processing software and libraries.
3. Maximize your lab’s efficiency using a robust and integrated system with automated calibration and a benchtop footprint.

Figure 1: Metabolomic Profiles can Clearly Differentiate Diseased from Healthy Samples. Using the untargeted metabolomics strategy described here, samples from a prostate cancer study were analyzed, as well as QC samples (spiked with known compounds) and a control matrix sample (urine). Shown in the above Scores plot, PCA analysis clearly differentiates healthy from disease and samples, confirming the original sample classification. Samples within the same group cluster together highlighting the reproducibility of the technique. The related loadings plot is used to determine which peaks are responsible for this differentiation.
Methods

Sample Preparation: Urine samples were obtained with disease classifications that were previously determined using accepted clinical techniques. The specific gravity of the samples was measured by testing 15 μL of urine samples with a refractometer prism. A 50 μL volume of the thawed urine samples was then transferred to a clean, labeled microcentrifuge-filter tube. An isotopically labeled internal standard mixture (20 μL) was added. The urine sample was then diluted with 400 μL 98:2 acetonitrile/water with 0.1% sodium azide, then vortexed. The sample was then centrifuged, and the supernatant was isolated. This was then dried down and reconstituted in 50 μL of 0.1% formic acid in water. The samples were transferred to glass vial and loaded into the autosampler.

Chromatography: The reverse phase HPLC separation was performed using a Shimadzu LC System, operating at a flow rate of 350 μL/min. The column used was an Ace Excel C18-PFP column (100 x 1mm, 2 µm) from ACE, maintained at 30 ºC. A standard reverse phase gradient was used employing mobile phase A as 0.1% formic acid in water and mobile phase B as acetonitrile. The injection volume was 3 μL in positive ion mode and 5 μL in negative ion mode.

Mass Spectrometry: The data was collected using information dependent acquisition (IDA) on the X500R QTOF System (SCIEX). Using optimized source conditions, the MS mass range analyzed was 50-600 m/z and the MS/MS was acquired with a mass range of 40-600 m/z with a 25 msec accumulation time. The collision energy was set to 35 V with a 15 V collision energy spread to ensure high quality MS/MS on most metabolites.

Data Processing: The data was processed in MarkerView™ Software 1.3 and PCA analysis was performed. Ions of interest were saved as an Interest List and copied into SCIEX OS Software where a formula was generated for each m/z-RT ion pairs. These formulae were scored using MS and MS/MS data, then searched using the ChemSpider database.
Finding the Differentiating Features

Using untargeted metabolomics one can differentiate sample groupings based on their metabolic profiles. MarkerView Software can view any trends or clustering within a dataset and tie them back to the sample groupings to find ions of interest responsible for the differences. In this pilot prostate cancer study, the samples with known disease classifications as well as QC samples (spiked with mix of known standard compounds) and the control urine (matrix QC) were analyzed by MarkerView Software. In Figure 1, the PCA scores plots shows that the samples from the same group cluster together well highlighting good reproducibility across the data collection. Removing the control urine and QC samples and repeating the PCA, a clear differentiation between the groups of samples is seen, highlighting that the metabolic profile is different between healthy and disease samples (Figure 3, left). The loadings plot highlights the m/z - RT ion pairs responsible for causing the differentiation between the groups (Figure 3, right), with the features higher in concentration in healthy vs disease on the right and those features lower in concentration in healthy vs disease on the left.

Identifying the Metabolites

At this stage these m/z - RT ion pairs are still features therefore need identification and confirmation. The m/z - RT ion pairs of interest are saved to an Interest List and copied directly into SCIEX OS Software for identification (Figure 4, Analytics pane). Analytics allows data processing such identification (library search against SCIEX libraries), peak integration, quantitation and any other statistics which may need to be reviewed for confidence in assignment. In Figure 4, selecting the m/z in the left panel of Components and Groups displays the extracted ion chromatogram (XIC) as well as the MS and MS/MS spectra for review. The Formula Finder algorithm then generates a set of possible formulas on an unknown ion based on the parent mass, mass error and isotope distribution pattern. These formulas can then be searched in the ChemSpider database for possible structures (Figure 5).

In the example shown (Figure 4 and 5), a particular ion of interest (166.0859 m/z) with a formula of C_9H_{11}O_2, matched many different structures in the ChemSpider database – specifically 1853 structures. To determine which structure is the...
correct match, the experimental MS/MS for that m/z is matched to the theoretical fragment pattern of each of the returned structures. In this case, the first match in the list is DL-Phenylalanine (Figure 5). Selecting that match, the structure is displayed, as well as the MS/MS matching to the theoretical fragments, and the resulting mass error. This rich information confirms the differential feature found to be unusually higher in a few samples is phenylalanine. Other small metabolites were found to be differential between the healthy and diseased samples which are known and reported elsewhere.

Conclusions

The X500R QTOF system is a robust easy to operate, benchtop system that requires minimal MS expertise to perform untargeted metabolomics analyses. As metabolomics continues to expand in disease research, robust easy-to-use solutions that provide quality answers will be increasingly important.

In this study, samples from a pilot prostate cancer study was analyzed and a clear difference between healthy and disease urine samples were detected using this untargeted metabolomics approach, confirming the original disease classifications. MarkerView Software was used to determine a list of the statistically significant analytes that distinguished the samples, and then the SCIEX OS compound searching provided formulae finding as well as structural matching through the ChemSpider database. This pilot study provided confidence in the approach and the next larger phase of the study analyzing a much larger set of samples is underway.

Figure 5: ChemSpider Database Matching. Formulae assigned in SCIEX OS Software (analytics) can be searched through the ChemSpider database for structural matching. The experimental data is matched to MS/MS spectra from theoretically fragmented potential candidate structures in order to confirm the identification. Here Phenylalanine was identified as a key metabolite that was up-regulated in healthy samples vs disease.