

HHS Public Access

Author manuscript Analyst. Author manuscript; available in PMC 2017 June 21.

Published in final edited form as: Analyst. 2016 June 21; 141(12): 3648–3656. doi:10.1039/c6an00200e.

Single-cell Mass Spectrometry with Multi-solvent Extraction Identifies Metabolic Differences between Left and Right Blastomeres in the 8-cell Frog (Xenopus) Embryo

Rosemary M. Onjiko1, **Sydney E. Morris**1, **Sally A. Moody**2, and **Peter Nemes**1,*

¹Department of Chemistry, The George Washington University, Washington DC, 20052

²Department of Anatomy & Regenerative Biology, The George Washington University, Washington DC, 20052

Abstract

Single-cell metabolic mass spectrometry enables the discovery (untargeted) analysis of small molecules in individual cells. Using single-cell capillary electrophoresis high-resolution mass spectrometry (CE-HRMS), we recently uncovered small-molecule differences between embryonic cells located along the animal–vegetal and dorsal–ventral axes of the 16-cell frog (Xenopus laevis) embryo, raising the question whether metabolic cell heterogeneity also exists along the left–right body axis. To address this question, we here advance single-cell CE-HRMS for identifying and quantifying metabolites in higher analytical sensitivity, and then use the methodology to compare metabolite production between left and right cells. Our strategy utilizes multiple solvents with complementary physicochemical properties to extract small molecules from single cells and improve electrophoretic separation, increasing metabolite ion signals for quantification and tandem HRMS. As a result, we were able to identify 55 different small molecules in D1 cells that were isolated from 8-cell embryos. To quantify metabolite production between left and right cells, we analyzed $n = 24$ different D1 cells in technical duplicate–triplicate measurements. Statistical and multivariate analysis based on 80 of the most repeatedly quantified compounds revealed 10 distinct metabolites that were significantly differentially accumulated in the left or right cells ($p <$ 0.05 and fold change 1.5). These metabolites were enriched in the arginine–proline metabolic pathway in the right, but not the left D1 cells. Besides providing analytical benefits for single-cell HRMS, this work provides new metabolic data on the establishment of normal body asymmetry in the early developing embryo.

Graphical abstract

^{*}**Correspondence to:** Peter Nemes, Department of Chemistry, The George Washington University, 800 22nd Street, NW, Suite 4000, Washington, DC 20052, USA, (Tel) (1) 202-994-5663, petern@gwu.edu.

Author Contributions: P.N., R.M.O., and S.A.M. designed the research; S.A.M. isolated the Xenopus blastomeres; R.M.O., S.E.M., and P.N. analyzed the cells; P.N. and R.M.O. interpreted the data and prepared the manuscript.

Single-cell Metabolic CE-ESI-HRMS

Keywords

single cell; mass spectrometry; metabolomics; left/right asymmetry; Xenopus

INTRODUCTION

Chemical analysis is central to understanding how molecular processes coordinate normal development such as breakage of the body symmetry in the early developing embryo.^{1, 2} Studies in many different animals, based on gene-by-gene manipulations and more recent RNA-sequencing approaches, have identified differentially inherited or distributed molecules that form signaling centers, which then specify the formation of the animal– vegetal and dorsal–ventral axes of the developing embryo.^{2–8} However, despite significant research efforts, the molecular determinants of the left–right body axis are yet to be fully understood. In many animals, left-right asymmetry is influenced by directed flow of the Nodal signaling factor across the midline of the gastrula, but there also is evidence for earlier asymmetry in cytoskeletal elements, ion pumps and channels, and signaling molecules^{9–13}. By extending high-resolution mass spectrometry (HRMS) to single blastomeres, we recently quantified asymmetrical distribution for proteins¹⁴ and smallmolecules (metabolites)¹⁵ along the animal–vegetal and dorsal–ventral body axes. Because metabolites are dynamically influenced by transcriptional and translational events,¹⁰ the discovery analysis of the metabolome raises a powerful tool to further investigate the initial molecular foundations of left–right patterning.

Metabolic analysis with single-cell resolution is enabled by several analytical techniques (see recent reviews in references $16-20$). Single-cell mass spectrometry (MS) provides labelfree detection, qualitative–quantitative information, and capability for targeted or untargeted (discovery) operation. Single-cell MS technologies include but are not limited to secondary ion MS (SIMS) for lipids in mating bacteria²¹ and single *Xenopus* embryos²², matrixassisted^{23_25} and matrix-free^{26, 27} laser desorption ionization for endogenous metabolites and drugs in animal, plant, or microbial cells and cell cultures. Additionally, single-cell MS can be performed under atmospheric-pressure conditions. For example, lipid and fatty acid changes were monitored in bovine²⁸, porcine²⁹, and mouse³⁰ oocytes and preimplantation embryos by desorption electrospray ionization as well as *Xenopus* eggs by laser ablation electrospray ionization³¹. Laser desorption ionization³² and ablation^{33, 34} found cellular heterogeneity and subcellular gradients, and direct-microsampling electrospray ionization $(ESI)³⁵$ ³⁷ helped to track drug metabolism in live cells. Other single-cell MS strategies incorporate a separation step to reduce chemical complexity and matrix effects prior to

ionization and detection of small molecules. For example, peptides were characterized in neurons using microsampling with liquid chromatography³⁸, metabolites from stimulated single heart cells were electrophoretically separated in lab-on-a-chip devices,³⁹ and capillary electrophoresis (CE) with ESI-HRMS was used to detect metabolites in isolated molluskan and mammalian single neurons^{40_42}. We recently adapted CE-ESI-HRMS to single *Xenopus* blastomeres and quantified dorsal–ventral and animal–vegetal metabolic cell heterogeneity in the 16-cell embryo.¹⁵

Here we further the sensitivity of the single-cell CE-ESI-HRMS workflow and explore cellto-cell differences along the left–right axis in the early developing Xenopus embryo. We use solvent systems with complementary physicochemical properties to enhance small-molecule extraction from blastomeres and to improve their separation by CE. The workflow allowed us to quantify ~80 and identify ~55 different metabolites in single D1 blastomeres that were isolated from the 8-cell *Xenopus* embryo. Next, multi-solvent extraction was used to quantify metabolic differences between left and right D1 cells in $n = 3-5$ biological replicates with each extract analyzed in technical duplicate–triplicate. The CE-ESI-HRMS metadata were evaluated using multivariate and statistical data analysis tools, revealing significant small-molecule differences between the left and right D1 cells. Metabolic pathway enrichment analysis identified the differentially accumulated metabolites to be represented in the arginine–proline pathway in the right, but not the left D1 blastomeres. Multi-solvent extraction presents analytical benefits for metabolomics by single-cell MS, which in turn fosters the understanding of basic biochemical processes underlying cell or organism development.

EXPERIMENTAL

Chemicals

Formic acid, acetic acid, methanol, acetonitrile, and water were LC-MS-grade, from Fisher Scientific (Fair Lawn, NJ). Ammonium hydroxide and acetylcholine were from Acros Organics (Fair Lawn, NJ). Eagle's minimum essential medium (Sigma Aldrich; St. Louis, MO) was used as a mixture of the standard L-amino acids: Ala, Arg, Asn, Asp, Glu, Gln, Gly, His, Ile, Leu, Lys, Phe, Pro, Ser, Tyr, and Val. All solvents were LC-MS grade, and all chemical standards were reagent grade or higher.

Solutions

Fresh Steinberg's solution (100%) was prepared as previously described¹⁵. Two-fold dilution of this solution yielded 50% Steinberg's solution. Metabolite extraction solutions spanned a range of organic content (polar to apolar) and pH to promote the extraction of complementary types of small molecules. The " $polar_{pH4}$ " solution was 50% (v/v) methanol prepared with 0.5% (v/v) acetic acid, yielding pH 3.89. The "apolar_{pH5}" solution contained 40% (v/v) acetonitrile and 40% (v/v) methanol, yielding pH 4.70. The "*apolar_{pH8}*" solution contained 40% (v/v) acetonitrile and 40% (v/v) methanol with pH titrated to 8.30 using ammonium hydroxide.

Animals and Cell Isolation

Adult male and female Xenopus laevis frogs were purchased from Nasco (Fort Atkinson, WI), and housed in a breeding colony at GWU (IACUC #A311). Fertilized 8-cell embryos were obtained by gonadotropin-induced egg laying and *in vitro* fertilization. The jelly coats of the embryos were removed as previously described.⁴³ Dejellied embryos were selected at the 2-cell stage if the lightly pigmented animal hemisphere was bisected by the first cleavage furrow in order to accurately predict the dorsal–ventral axis.⁴⁴ These embryos were transferred into a Petri dish containing 100% Steinberg's solution at room temperature and their cleavage patterns monitored with a stereomicroscope. Upon reaching the 8-cell stage, typically ~2.25 h post-fertilization, those in which left and right D1 blastomeres could be identified based on pigmentation and location in reference to established cell-fate maps⁴⁵ were transferred to 50% Steinberg's solution in an agarose-coated Petri dish. The left and right D1 blastomeres were dissected free using sharpened forceps following an earlier protocol.⁴⁶ Each cell type was collected at $n = 5$ biological replicates. Each isolated blastomere was labeled with a unique cell identifier to aid interpretation of results.

Single-Cell Extracts

Each isolated blastomere was directly transferred into a distinct 0.6-mL Eppendorf microvial (Fisher) containing 20 µL of methanol chilled to ~4 °C to quench^{15, 41, 47–49} enzymatic reactions. Subsequently, the methanol solution containing the blastomere was dried in a vacuum concentrator (Labconco; Kansas City, MO) at 4 °C. The sample was reconstituted in 5 µL of metabolite extraction solution, sonicated in an ice-cold water bath for 3 min, and vortexed for 1 min. Metabolite extracts were centrifuged at $8,000 \times g$ at 4 °C (Sorvall Legend X1R; Thermo Scientific, Waltham, MA) for 3 min and stored at −80 °C until measurement by CE-ESI-MS.

Single-cell CE-ESI-HRMS

Extracts were thawed, vortex-mixed, and centrifuged at $8,000 \times g$ at 4 °C to pellet cell debris before measurement using a CE-ESI-MS platform that we recently developed for small molecules in single *Xenopus* blastomeres.¹⁵ Briefly, the instrument consists of a custombuilt CE platform capable of injecting ~1–20 nL from ~1 µL extract and separating small molecules, a co-axial sheath flow CE-ESI interface without nebulizer gas to ionize molecules, and a quadrupole orthogonal acceleration time-of-flight (q-oaTOF) highresolution tandem mass spectrometer to mass-analyze ions (Impact HD, Bruker Daltonics, Billerca, MA). The mass spectrometer was tuned and mass-calibrated to $\lt 1$ ppm accuracy following vendor instructions and operated at 40,000 FWHM resolution. Experimental parameters for CE were as follows: injection volume, 10 nL; CE fused silica dimensions, 40/105 μ m internal/outer diameter \times 90 cm length; background electrolyte, 1% (v/v) formic acid; electrophoretic separation voltage, 17–23 kV (applied to the injection end). ESI parameters included: electrospray solvent, 50% methanol containing 0.1% (v/v) formic acid; spray flow rate, 1 µL/min; spray voltage, -1,700 V (applied to the orifice plate of the mass spectrometer); electrospray regime, cone jet (controlled as previously described 49). Mass spectrometer settings for single-stage and data-dependent acquisition (DDA) were: mass range for survey (MS¹) and MS², m/z 50–500; survey scan rate, 2 Hz; MS² scan rate, 2 Hz;

collision-induced dissociation energy/gas, 18 eV/nitrogen; fragmentation on 3 most-intense features; active exclusion, after 3 spectra for 60 s; smart exclusion, activated with 5 \times . Quality control and operation of the platform followed our recent protocol to enable 60 amol lower limit of detection and $3-5$ log-order dynamic range of quantification¹⁵. Instrumental repeatability was characterized daily using 50 nM acetylcholine. Over 10 days of continuous data collection, quantitative repeatability was 11% relative standard deviation (RSD) in migration time and 17% RSD in peak area for quantification. We required a minimum repeatability of <25% RSD in peak area and separation time before measuring cell extracts. The separation capillary was flushed with BGE for 5 min at the end of each separation experiment followed by a 2-min blank (BGE injected) analysis to test CE-ESI-MS signal stability before injection of the next sample.

Data Analysis and Software

Raw MS data files were processed using custom-written scripts in Compass Data Analysis ver. 4.0 (Bruker Daltonics) according to our previous protocol.⁴⁹ Briefly, each raw file was externally mass-calibrated to < 1 ppm accuracy (HPC calibration mode in Data Analysis) with sodium-formate clusters that formed in the CE-ESI ion source as sodium ions separated from the extracts, and molecular features (unique m/z vs. time domains) were manually searched between m/z 50–500 with a 500-mDa incremental step. To note the molecular feature, the accurate mass $(m/z$ value) and the migration time was noted for the apex of each feature within its elution window. The resulting metadata was analyzed in MetaboAnalyst 3.0^{50} , a public web-based metabolomic pipeline, with the following settings: type of normalization, sum (to signal abundance); type of scaling, auto. Fisher's least significant difference analysis (LSD) was adopted post hoc. Statistical analysis utilized Student's twotailed t-test (homoscedastic) with $p < 0.05$ chosen to mark significance for normally distributed data. A fold change of 1.5 was chosen to denote biological significance. Octanol-water distribution coefficients were calculated in MarvinSketch 16.1.11 (2016 ChemAxon, Budapest, Hungary). Ion fragmentation pathways during collision-induced dissociation were predicted in Mass Frontier 7.0 (Thermo Scientific) to aid molecular identifications.

Safety Considerations

Standard safety procedures apply for handling chemicals. Capillaries, which present wounding hazard, must be handled with care using safety glasses. As high voltage poses electrical shock hazard, electrically conductive parts of the CE-ESI interface must be grounded or shielded to prevent accidental exposure; in this work, the CE system was housed in a Plexiglass enclosure with an interlock-enabled door controlling the CE highvoltage power supply as an active safety mechanism.

RESULTS AND DISCUSSION

Three-solvent Small-molecule Extraction for Single Cells

Our first goal was to enhance the detectable coverage of the metabolome for single blastomeres in the Xenopus embryo. In traditional metabolomic experiments, detection and identification of metabolites are enhanced by modulating compound distribution between the

extraction solvent and the specimen.^{51, 52} For example, amino acids are efficiently extracted using methanol-based solvent systems, and extraction performance can be improved by multiple extraction steps in sequence or fine adjustments to the solvent composition.^{53_55} Similarly, our accumulated experience showed 50% (v/v) methanol containing 0.5% (v/v) acetic acid to be capable of extracting an appreciable number of polar small molecules from single neurons^{41, 49} and embryonic cells¹⁵. Here we proposed that the utilization of multiple extraction solvents deepens the metabolic coverage in single-cell measurements using CE-ESI-HRMS.

This strategy was tested by evaluating metabolite extraction using different solvent systems. In addition to extraction by methanol, the study included apolar conditions to help lyse the cell membrane and facilitate the detection of apolar compounds. To guide the selection of pH for the extraction solvents, we calculated the octanol/water distribution coefficient (D) between pH 2–10 for ~20 randomly selected small molecules that were detected in our previous single-cell studies.¹⁵ As shown in Figure S1, the coefficients were highly pHdependent for several metabolites. For example, D varied by 3-log-orders for arginine, lysine, methionine, aspartic acid, and glutamic acid between pH 4 and 9. This information supported the notion that extraction may be enhanced by solvents with complementary polar–apolar characteristics and pH. For this study, we selected 50% methanol containing 0.5% acetic acid with pH 3.80 (denoted as "polar_{pH4}") as it performed well for polar compounds in our earlier work.¹⁵ To help extract apolar metabolites, we increased the apolar characteristic of this solvent formulation to 40% acetonitrile and 40% methanol with pH titrated to 4.70 ("apolar_{pH5}") and pH 8.30 ("apolar_{pH8}"). These three solvents were anticipated to extract polar and apolar compounds under basic and acidic conditions with complementary performance.

Next, the solvent systems were applied to single blastomeres. Single midline animal-dorsal (D1) cells were identified in the right hemisphere of 8-cell Xenopus embryos (Fig. 1), and these cells (D1R) were microdissected following established cell-fate maps and isolation protocols.^{45, 46} The separated cells were immediately transferred into a microvial containing chilled methanol (4 °C) to rapidly quench enzyme activity. While cells appeared physically intact in the solvent, exposure to methanol likely caused extraction for some compounds (e.g., polar metabolites). To control extraction across different solvent polarity–pH domains, contents of the microvial (cell and methanol) were lyophilized at 4 °C, causing methanolextracted metabolites to precipitate on the surfaces of the cells and the vial. Ultimately, metabolites were extracted by adding polar_{pH4}, apolar_{pH5}, or apolar_{pH8} solvents to the contents of each microvial (cell and precipitated metabolites), followed by sonication in icecold water to lyse the cells and facilitate extraction. A total of $n = 3-4$ different D1R blastomeres (biological replicates) were measured for each solvent to account for technical and biological variability and empower statistical data analysis. The resulting extracts were centrifuged to pellet cell debris and precipitated proteins. A 10 nL portion of the aliquot was analyzed using our custom-built single-cell CE-ESI-HRMS platform (Fig. 1).

The metabolic compositions of the resulting extracts were qualitatively compared. The primary mass spectrometric data were manually surveyed to identify molecular features, defined here as distinct accurate mass $(m/z$ value) vs. separation time domains. The resulting

molecular features were manually revised to disregard background signals from the solvents (e.g., common contaminants in ESI-HRMS) as well as isotopes, adducts, or noncovalent clusters. As a result, 156–232 different molecular features were detected using the multiple solvents, 25–35% of which were exclusive to each solvent (Fig. 2A) and 92 were detected in at least 3 biological replicates (see list in Table S1). We identified 55 different molecular features by comparing their accurate mass, migration time, and fragmentation behavior via collision-induced dissociation using mass spectrometric databases (e.g., Metlin⁵⁶, Human Metabolome Database⁵⁷, and mzCloud), computational prediction (Mass Frontier), chemical standards, as well as our in house-built CE-ESI-MS/MS² metabolomic database for Xenopus laevis¹⁵. Identified metabolites are tabulated in Table S2 and include amino acids, energy carriers, bases, small organic acids, osmolites, and dipeptides. These metabolite identifications correspond to a $~40\%$ enhancement by multi-solvent extraction compared to our previous single-solvent approach¹⁵.

Performance was quantified on the basis of CE separation power and signal-to-noise (S/N) ratio. The single-cell analysis workflow benefited from high separation power to simplify the complex metabolome for detection, provide separation time as compound-dependent information to help metabolite identification, and minimize spectral interferences to aid quantification. Representative separation of compounds is shown in Figure 2B. Based on the separation of arginine, creatine, glutamic acid, and GSH, apolar $_{nH5}$ produced $~4$ -times higher S/N than the counterparts, and apolar solvents yielded 16–37% higher theoretical plate numbers. We attribute these enhancements in CE to on-column sample preconcentration by field-amplified sample stacking. This resulted from lower electrical conductivity in the acetonitrile-containing solvents and a dynamic pH -junction⁵⁸ at the interfacing of the acidic background electrolyte $(\sim pH 3)$ with the sample prepared in basic $apolar_{nH8}$.

The cell extracts were compared using multivariate and statistical tools. Selected-ion electropherograms were generated for the 80 most repeatedly quantified small molecules (Table S1) between the blastomeres and their under-the-curve peak areas were integrated to serve as a quantitative proxy for metabolite abundance. The median quantitative error for these 80 features based on technical duplicates was ~12% relative standard deviation (RSD) for the polar_{pH4}, ~18% RSD for the apolar_{pH5}, and 23% RSD for the apolar_{pH8} extracts, providing sufficient repeatability to interpret biological significance with a fold change 1.5. To account for naturally variable cell sizes between embryos, we normalized the metadata to total signal abundance (see also Experimental); indeed, the normalized areas followed a normal distribution and were centered at zero (0) counts (data not shown). These normalized metadata were evaluated using analysis of variance (ANOVA) and hierarchical cluster analysis (HCA). Figure 3 presents the HCA-heatmap calculated for the 70 most statistically significant molecular features of the metadata. Differential clustering between the molecular features and extraction conditions (see dendrogram branches) highlight groups of small molecules that were differentially extracted by the respective solvents (see grey squares). Fisher's least significant difference (LSD) analysis found about a dozen of these differences to be statistically significant (see Table S3). Complementary metabolite extraction raised the potential to enhance the quantification of the single-cell metabolome.

Assessing Metabolic Differences between Left-Right D1 Blastomeres

Last, we asked whether blastomeres harbor different metabolomes on the left versus right side of the Xenopus embryo (see Fig. 1). This question is of importance because there are conflicting reports regarding left-right asymmetries in ion and neurotransmitter flow at cleavage stages.⁹, ^{12, 14} We selected the D1 blastomere in the 8-cell embryo for study because: i) it already exhibits left-right asymmetry in H⁺/K⁺-ATPase alpha subunit mRNA and protein⁵⁹, ⁶⁰; ii) blastomeres in the 8-cell embryo are considerably large, \sim 180 nL in volume, facilitating manual cell dissection; iii) D1 blastomeres are precursors to organs that show laterality in the tadpole (e.g., heart and hindgut⁶¹). To enhance the success of finding metabolic differences between the blastomeres, we minimized biological variability by using only embryos derived from a single set of parents, thus ensuring a common genetic background. We manually dissected left D1 (D1L) and right D1 (D1R) blastomeres and extracted each with the polar_{pH4}, apolar_{pH5}, or apolar_{pH8} solvents. To enhance the power of statistical data analysis, $n = 3-5$ different blastomeres (biological replicates) were processed using each solvent for the left or the right D1 cells, amounting to a total of 24 different blastomeres collected. With each blastomere measured in 1–3 technical replicates, a total of 36 different single-cell CE-ESI-HRMS measurements were performed in this portion of the study.

Small-molecule composition was compared between D1L and D1R blastomeres. The peak areas were determined for the 80 most repeatedly extracted compounds (Table S1). The data was normalized (see Experimental) to account for potential differences in cell sizes, and these metadata on left vs. right cell content were analyzed for statistical significance (Student's t-test) and biological importance (fold change between D1L/D1R). This correlation is presented as a volcano plot in Figure 4. Most small molecules were comparably produced in the D1L and D1R blastomeres, indicating that the main metabolic activities of the left and right blastomeres are similar in the 8-cell Xenopus embryo. Other metabolites had significantly different abundances between the left and right blastomeres. With complementary performance in separation, sensitivity, and quantitative repeatability, the different extraction solvents were able to uncover metabolic differences between cells to varying extent of statistical and biological significance (p value/FC). For example, significantly different amounts of GABA were extracted from D1L and D1R using the apolar ($\langle 0.05/ 1.5 \rangle$ but not the polar_{pH4} (0.585/1.4) solvent. Trolamine enrichment was revealed in D1L based on polar_{pH4} extracts, which was also captured to a biological, but not statistical significance during apolar extraction (0.07/2.41 for apolar_{pH5} and 0.257/1.72 for apolar_{pH8}). Likewise, leucine accumulation in D1L was evidenced based on apolar_{pH5} to a biological significance in apolar_{pH8} (0.055/1.59), but not polar_{pH4} (0.626/1.271). Combined, metabolites that were differentially enriched included leucine, isoleucine, ethanolamine, GABA, and trolamine with higher abundance in D1L, and creatine, acetylcarnitine, spermidine, S-adenosylmethionine, and putrescine with higher abundance in D1R (Table 1). Combined, these findings suggest slight but detectable asymmetry in metabolic activity along the left–right axis in the 8-cell Xenopus embryo.

These differentially enriched metabolites were mapped against known metabolic pathways. Using MetaboAnalyst as the search engine and *Danio rerio* (zebrafish) as the model, the 10

CONCLUSIONS

In the presented work, we addressed the analytical sensitivity of discovery single-cell HRMS measurements to ask a cell biological question, whether blastomeres have different metabolomes along the left–right body axis in the early developing embryo. We enhanced the detectable coverage of the single-cell metabolome by designing solvent systems with complementary physicochemical properties (polarity and pH). Multi-solvent extraction also facilitated metabolite detection during single-cell CE-ESI-HRMS. On-column enrichment led to higher signal-to-noise ratios, which in turn improved small-molecule identifications and quantification. Complementary analytical performance by multiple extraction solvents enabled the detection of statistically and biologically significant enrichment differences for 10 different metabolites between D1L and D1R blastomeres, which would have been limited to fewer metabolites during a classical approach based on single-solvent extraction. Comparison with KEGG metabolic database found these metabolites to be enriched in the arginine–proline pathway in D1R, but not in D1L blastomeres, suggesting altered metabolism between the cells. Overall, these data captured asymmetry in the metabolic activity of D1 blastomeres on the left–right sides of the 8-cell Xenopus embryo.

Although the biological significance of the observed metabolic activity differences between left–right D1 blastomeres is unknown to us at present, these results demonstrate the potential of HRMS to aid cell and developmental biology studies. Continuing advances in cell sampling and treatment (e.g., sampling by microcapillaries $35-37, 42, 62$) enable the analysis of progressively smaller cells. To offset lower signal-to-noise ratios resulting from smaller materials measured in these studies, new technologies in sample handling, ionization, and HRMS detection sensitivity are required. Advancing tandem and multistage (MSⁿ) highresolution mass spectrometric databases and related search engines, such as Metlin⁵⁶, Human Metabolome Database⁵⁷, and mzCloud (based on precursor ion fingerprinting⁶³ fragmentation trees⁶⁴), are essential to improving the confidence of metabolite identifications. Equally important is the measurement of a higher number of single cells to empower statistical analysis. By enabling the assaying of a large number of small molecules with deeper coverage of the metabolome, HRMS measurements of single cells provide new molecular insights into cell-to-cell differences that may not be detectable in classical cellpopulation averaging experiments, raising the ability to help better design hypothesis-driven studies for health research.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

Acknowledgments

This research was supported by the National Institutes of Health Grant R21 GM114854 (to P.N.), The George Washington University Department of Chemistry Start-Up Funds (to P.N.), and the COSMOS Club Foundation award (to R.M.O).

REFERENCES

- 1. Sutherland MJ, Ware SM. American Journal of Medical Genetics Part C-Seminars in Medical Genetics. 2009; 151C:307–317.
- 2. Blum M, Schweickert A, Vick P, Wright CVE, Danilchik MV. Dev. Biol. 2014; 393:109–123. [PubMed: 24972089]
- 3. Davidson EH. Development. 1990; 108:365–389. [PubMed: 2187672]
- 4. King ML, Messitt TJ, Mowry KL. Biol. Cell. 2005; 97:19–33. [PubMed: 15601255]
- 5. Heasman J. Development. 2006; 133:1205–1217. [PubMed: 16527985]
- 6. Abrams EW, Mullins MC. Curr. Opin. Genet. Dev. 2009; 19:396–403. [PubMed: 19608405]
- 7. De Domenico E, Owens NDL, Grant IM, Gomes-Faria R, Gilchrist MJ. Dev. Biol. 2015; 408:252– 268. [PubMed: 26100918]
- 8. Cuykendall TN, Houston DW. Dev. Dyn. 2010; 239:1838–1848. [PubMed: 20503379]
- 9. Vandenberg LN, Lemire JM, Levin M. Comm. Integr. Biol. 2013; 6:e27155.
- 10. Namigai EKO, Kenny NJ, Shimeld SM. Genesis. 2014; 52:458–470. [PubMed: 24510729]
- 11. Tingler M, Ott T, Tozser J, Kurz S, Getwan M, Tisler M, Schweickert A, Blum M. Genesis. 2014; 52:588–599. [PubMed: 24585437]
- 12. Coutelis JB, Gonzalez-Morales N, Geminard C, Noselli S. EMBO Rep. 2014; 15:926–937. [PubMed: 25150102]
- 13. Fukumoto T, Kema IP, Levin M. Curr. Biol. 2005; 15:794–803. [PubMed: 15886096]
- 14. Lombard-Banek C, Moody SA, Nemes P. Angew. Chem., Int. Ed. 2016; 55 in print.
- 15. Onjiko RM, Moody SA, Nemes P. Proc. Natl. Acad. Sci. U. S. A. 2015; 112:6545–6550. [PubMed: 25941375]
- 16. Zenobi R. Science. 2013; 342:1243259. [PubMed: 24311695]
- 17. Trouillon R, Passarelli MK, Wang J, Kurczy ME, Ewing AG. Anal. Chem. 2013; 85:522–542. [PubMed: 23151043]
- 18. Rubakhin SS, Romanova EV, Nemes P, Sweedler JV. Nat Methods. 2011; 8:S20–S29. [PubMed: 21451513]
- 19. Borland, LM.; Kottegoda, S.; Phillips, KS.; Allbritton, NL. Annu Rev Anal Chem (Palo Alto Calif). Vol. 1. Palo Alto: Annual Reviews; 2008. p. 191-227.
- 20. Dittrich P, Ibanez AJ. Electrophoresis. 2015; 36:2196–2206.
- 21. Kurczy ME, Piehowski PD, Van Bell CT, Heien ML, Winograd N, Ewing AG. Proc. Natl. Acad. Sci. U. S. A. 2010; 107:2751–2756. [PubMed: 20133641]
- 22. Tian H, Fletcher JS, Thuret R, Henderson A, Papalopulu N, Vickerman JC, Lockyer NP. J Lipid Res. 2014; 55:1970–1980. [PubMed: 24852167]
- 23. Ibanez AJ, Fagerer SR, Schmidt AM, Urban PL, Jefimovs K, Geiger P, Dechant R, Heinemann M, Zenobi R. Proc. Natl. Acad. Sci. U. S. A. 2013; 110:8790–8794. [PubMed: 23671112]
- 24. Bouschen W, Schulz O, Eikel D, Spengler B. Rapid Communications in Mass Spectrometry. 2010; 24:355–364. [PubMed: 20049881]
- 25. Liu X, Hummon AB. Anal. Chem. 2015; 87:9508–9519. [PubMed: 26084404]
- 26. Walker BN, Antonakos C, Retterer ST, Vertes A. Angew. Chem., Int. Ed. 2013; 52:3650–3653.
- 27. O'Brien PJ, Lee M, Spilker ME, Zhang CC, Yan Z, Nichols TC, Li W, Johnson CH, Patti GJ, Siuzdak G. Cancer & metabolism. 2013; 1:4. [PubMed: 24280026]
- 28. Gonzalez-Serrano AF, Pirro V, Ferreira CR, Oliveri P, Eberlin LS, Heinzmann J, Lucas-Hahn A, Niemann H, Cooks RG. PLoS One. 2013; 8

- 29. Pirro V, Oliveri P, Ferreira CR, Gonzalez-Serrano AF, Machaty Z, Cooks RG. Anal. Chim. Acta. 2014; 848:51–60. [PubMed: 25263116]
- 30. Ferreira CR, Pirro V, Eberlin LS, Hallett JE, Cooks RG. Anal. Bioanal. Chem. 2012; 404:2915– 2926. [PubMed: 23052870]
- 31. Shrestha B, Sripadi P, Reschke BR, Henderson HD, Powell MJ, Moody SA, Vertes A. PLoS One. 2014; 9:e115173. [PubMed: 25506922]
- 32. Coello Y, Jones AD, Gunaratne TC, Dantus M. Anal. Chem. 2010; 82:2753–2758. [PubMed: 20210322]
- 33. Stolee JA, Shrestha B, Mengistu G, Vertes A. Angew. Chem., Int. Ed. 2012; 51:10386–10389.
- 34. Shrestha, B.; Vertes, A. Plant Metabolism. Sriram, G., editor. Vol. 1083. Humana Press; 2014. p. 31-39.
- 35. Pan N, Rao W, Kothapalli NR, Liu R, Burgett AW, Yang Z. Anal. Chem. 2014; 86:9376–9380. [PubMed: 25222919]
- 36. Fujii T, Matsuda S, Tejedor ML, Esaki T, Sakane I, Mizuno H, Tsuyama N, Masujima T. Nat. Protoc. 2015; 10:1445–1456. [PubMed: 26313480]
- 37. Zhang LW, Foreman DP, Grant PA, Shrestha B, Moody SA, Villiers F, Kwake JM, Vertes A. Analyst. 2014; 139:5079–5085. [PubMed: 25109271]
- 38. Buchberger, A.; Yu, Q.; Li, LJ. Annual Review of Analytical Chemistry, Vol 8. Cooks, RG.; Pemberton, JE., editors. Vol. 8. Palo Alto: Annual Reviews; 2015. p. 485-509.
- 39. Cheng W, Klauke N, Sedgwick H, Smith GL, Cooper JM. Lab Chip. 2006; 6:1424–1431. [PubMed: 17066165]
- 40. Lapainis T, Rubakhin SS, Sweedler JV. Anal. Chem. 2009; 81:5858–5864. [PubMed: 19518091]
- 41. Nemes P, Knolhoff AM, Rubakhin SS, Sweedler JV. Anal. Chem. 2011; 83:6810–6817. [PubMed: 21809850]
- 42. Aerts JT, Louis KR, Crandall SR, Govindaiah G, Cox CL, Sweedler JV. Anal. Chem. 2014; 86:3203–3208. [PubMed: 24559180]
- 43. Moody SA. Methods Mol. Biol. 2000; 135:331–347. [PubMed: 10791329]
- 44. Klein SL. Dev. Biol. 1987; 120:299–304. [PubMed: 3817297]
- 45. Moody SA, Kline MJ. Anat. Embryol. 1990; 182:347–362. [PubMed: 2252221]
- 46. Grant PA, Herold MB, Moody SA. Journal of visualized experiments : JoVE. 2013
- 47. Knolhoff AM, Nautiyal KM, Nemes P, Kalachikov S, Morozova I, Silver R, Sweedler JV. Analytical Chemistry. 2013; 85:3136–3143. [PubMed: 23409944]
- 48. Nemes P, Knolhoff AM, Rubakhin SS, Sweedler JV. ACS Chem Neurosci. 2012; 3:782–792. [PubMed: 23077722]
- 49. Nemes P, Rubakhin SS, Aerts JT, Sweedler JV. Nat. Protoc. 2013; 8:783–799. [PubMed: 23538882]
- 50. Xia JG, Sinelnikov IV, Han B, Wishart DS. Nucleic Acids Res. 2015; 43:W251–W257. [PubMed: 25897128]
- 51. Want EJ, O'Maille G, Smith CA, Brandon TR, Uritboonthai W, Qin C, Trauger SA, Siuzdak G. Anal. Chem. 2006; 78:743–752. [PubMed: 16448047]
- 52. Masson P, Alves AC, Ebbels TM, Nicholson JK, Want EJ. Anal. Chem. 2010; 82:7779–7786. [PubMed: 20715759]
- 53. Rabinowitz JD, Kimball E. Anal. Chem. 2007; 79:6167–6173. [PubMed: 17630720]
- 54. Vastag L, Jorgensen P, Peshkin L, Wei R, Rabinowitz JD, Kirschner MW. PLoS One. 2011; 6:e16881. [PubMed: 21347444]
- 55. Bennett BD, Yuan J, Kimball EH, Rabinowitz JD. Nature protocols. 2008; 3:1299–1311. [PubMed: 18714298]
- 56. Zhu ZJ, Schultz AW, Wang JH, Johnson CH, Yannone SM, Patti GJ, Siuzdak G. Nat. Protoc. 2013; 8:451–460. [PubMed: 23391889]
- 57. Wishart DS, Jewison T, Guo AC, Wilson M, Knox C, Liu YF, Djoumbou Y, Mandal R, Aziat F, Dong E, Bouatra S, Sinelnikov I, Arndt D, Xia JG, Liu P, Yallou F, Bjorndahl T, Perez-Pineiro R,

Eisner R, Allen F, Neveu V, Greiner R, Scalbert A. Nucleic Acids Res. 2013; 41:D801–D807. [PubMed: 23161693]

- 58. Kazarian AA, Hilder EF, Breadmore MC. J Sep. Sci. 2011; 34:2800–2821. [PubMed: 22006737]
- 59. Levin M, Thorlin T, Robinson KR, Nogi T, Mercola M. Cell. 2002; 111:77–89. [PubMed: 12372302]
- 60. Adams DS, Robinson KR, Fukumoto T, Yuan SP, Albertson RC, Yelick P, Kuo L, McSweeney M, Levin M. Development. 2006; 133:1657–1671. [PubMed: 16554361]
- 61. Jaffe EH, Marty A, Schulte A, Chow RH. J Neurosci. 1998; 18:3548–3553. [PubMed: 9570786]
- 62. Deng JW, Yang YY, Xu MZ, Wang XW, Lin L, Yao ZP, Luan TG. Anal. Chem. 2015; 87:9923– 9930. [PubMed: 26360344]
- 63. Sheldon MT, Mistrik R, Croley TR. J Am. Soc. Mass Spectrom. 2009; 20:370–376. [PubMed: 19041260]
- 64. Vaniya A, Fiehn O. Trac-Trends Anal. Chem. 2015; 69:52–61.

Figure 1.

Experimental strategy to identify and quantify small molecules in D1 blastomeres in the 8 cell Xenopus embryo. Different D1 blastomeres were extracted using solvents with complementary physicochemical properties to enhance the detectable portion of the singlecell metabolome and assess metabolic differences between blastomeres on the left and right sides of the embryo. Scale bar = $200 \mu m$.

Figure 2.

Enhancing small-molecule detection from single blastomeres using multi-solvent extraction. (A) The polar_{pH4}, apolar_{pH5}, and apolar_{pH8} solvents allowed us to detect a complementary set of small molecules in the cells. Numbers indicate different compounds. **(B)** These solvents also provided complementary performance in electrophoretic separation; 28 select small molecules are shown. Higher signal-to-noise ratio and separation power was achieved using apolar_{pH5}, and apolar_{pH8} solvents. Theoretical plate numbers (N) are provided for arginine. Key: CR, creatine; GSH, glutathione; GSSG, oxidized glutathione; HPX, hypoxanthine; S-adenosylmethionine (9); ornithine (10); lysine (11). Identified compounds are listed in Table S2.

Figure 3.

HCA-heat map identifying differential metabolite extraction from single D1R blastomeres using the 3 solvents. The plot was calculated for the 70 most statistically significant differences. Individual blastomeres with unique cell identifiers (Cell ID) are shown on the horizontal axis (see bottom axis). Different molecular features are shown on the vertical axis (see right axis). Dashed boxes exemplify quantitative chemical differences between the extracts. Metabolites with statistically significant differences in abundance based on Fisher's LSD are labeled (see also Table S3). Metabolites are shown in three-letter codes. Key: Cit,

citrulline; GB, glycine betaine. Molecular features that have yet to be identified are not shown (see all labels in Fig. S2).

Author Manuscript Author Manuscript $\mathbf A$

 \bf{B}

Figure 4.

Differential metabolite enrichment between D1 blastomeres on the left (D1L) and right (D1R) sides of the 8-cell Xenopus embryo. **(A)** Volcano plot compares signal differences between D1L/D1R cell extracts prepared using the three solvents. Dashed lines (grey) mark thresholds for statistical significance ($p < 0.05$) and biological importance (fold change 1.5). **(B)** Relative comparison of differentially enriched, identified metabolites (see data in Table S4).

Onjiko et al. Page 18

Figure 5.

Pathway analysis for metabolites with asymmetric distribution between left and right D1 blastomeres. Correlation between p value from pathway enrichment analysis and pathway impact from pathway topology analysis in MetaboAnalyst identified arginine-proline as the most represented pathway (left panel). Identified metabolites underlying this pathway were mapped against KEGG using *Danio rerio* (zebrafish) as the model organism (right panel). Key: Numbers correspond to KEGG metabolite identifiers.

Table 1

Quantification of metabolite differences between left and right D1 blastomeres of the 8-cell Xenopus embryo.

* Note: Negative fold change values indicate down-production and were calculated by inverting ratios lower than 1.