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In situ monitoring of the shikimate pathway: a combinatorial approach of Raman reverse stable isotope probing and hyperspectral imaging†

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Sensing and visualization of metabolites and metabolic pathways *in situ* are significant requirements for tracking their spatiotemporal dynamics in a non-destructive manner. The shikimate pathway is an important cellular mechanism that leads to the *de novo* synthesis of many compounds containing aromatic rings of high importance such as phenylalanine, tyrosine, and tryptophan. In this work, we present a cost-effective and extraction-free method based on the principles of stable isotope-coupled Raman spectroscopy and hyperspectral Raman imaging to monitor and visualize the activity of the shikimate pathway. We also demonstrated the applicability of this approach for nascent aromatic amino acid localization and tracking turnover dynamics in both prokaryotic and eukaryotic model systems. This method can emerge as a promising tool for both qualitative and semi-quantitative *in situ* metabolomics, contributing to a better understanding of aromatic ring-containing metabolite dynamics across various organisms.

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Introduction

The shikimate pathway is the bridge between carbon source metabolism and biosynthesis of many aromatic ring-containing metabolites. The major products of this pathway are three aromatic amino acids, *i.e.*, phenylalanine, tyrosine and tryptophan. This pathway is present only in plants and microbes, like bacteria and yeast, but never in animals.^{1,2} It not only produces aromatic amino acids but also many other important aromatic intermediates and aromatic biomolecules such as chorismate, dehydroshikimate, folates, salicylic acid, resveratrol, flavonoids, dopamine, and alkaloids. These molecules have a wide range of applications in developing value-added aromatic compounds for the food, agrochemical, fuel additive, cosmetic, dye, pharmaceutical and other industries (Fig. 1).^{2,3} Over 20% of all fixed carbon flows through this pathway.⁴ The shikimate pathway also plays an important role mediating the production of antiviral drugs such as oseltamivir.⁵ The quinoid core found in benzoquinone and ubiquinone is a derivative of the shikimate pathway in bacteria and yeast.⁶ Novel approaches that can help one to understand or visualize metabolites and metabolic pathways *in situ* (in their native position) or *in vivo* (inside the living system) are still at the inception stage. This is primarily attributed to the metabolome complexity and dynamic nature of biomolecules.

The common conventional approaches for profiling metabolites and monitoring metabolic pathways are gas chromatography-mass spectrometry and liquid chromatography-mass spectrometry (GC-MS and LC-MS).^{4–8} These are highly efficient and sensitive methods for both qualitative and quantitative analyses of the entire pathway at each step until the final product is delivered. However, these approaches require the extraction of metabolites and do not provide information related to the dynamic nature of metabolic pathways.^{7,8} Fluorescence spectroscopy-based methods are an alternative but it involves the introduction of foreign molecules, which can perturb the natural metabolic processes.⁷ Raman spectroscopy is an efficient label-free, reagent-free, and non-destructive analytical tool for analysing biomolecules *in situ*. Stable isotope probes, such as ²H, ¹³C, ¹⁵N, and ¹⁸O, can be used as tags for analysing the *de novo* biosynthesis of different biomolecules simultaneously using Raman spectroscopy. When the parent atom is replaced by its heavier isotope, the reduced mass increases. As the relationship between the wave-number and the square root of reduced mass is not directly proportional, due to the isotopic effect, the spectral signal shifts to a lesser wavenumber known as the red shift of the peak. However, this approach is limited by the fact that stable isotope sources are not abundantly available and their relative cost is higher when compared to that of their unlabelled counterparts. To overcome this, we have used the strategy of Raman reverse stable isotope probing (RrSIP). A reverse isotopic effect is observed when the heavier isotope is replaced by the lighter isotope resulting in the reduction of reduced mass and a shift towards a higher wavenumber (blue shift).^{9,10}

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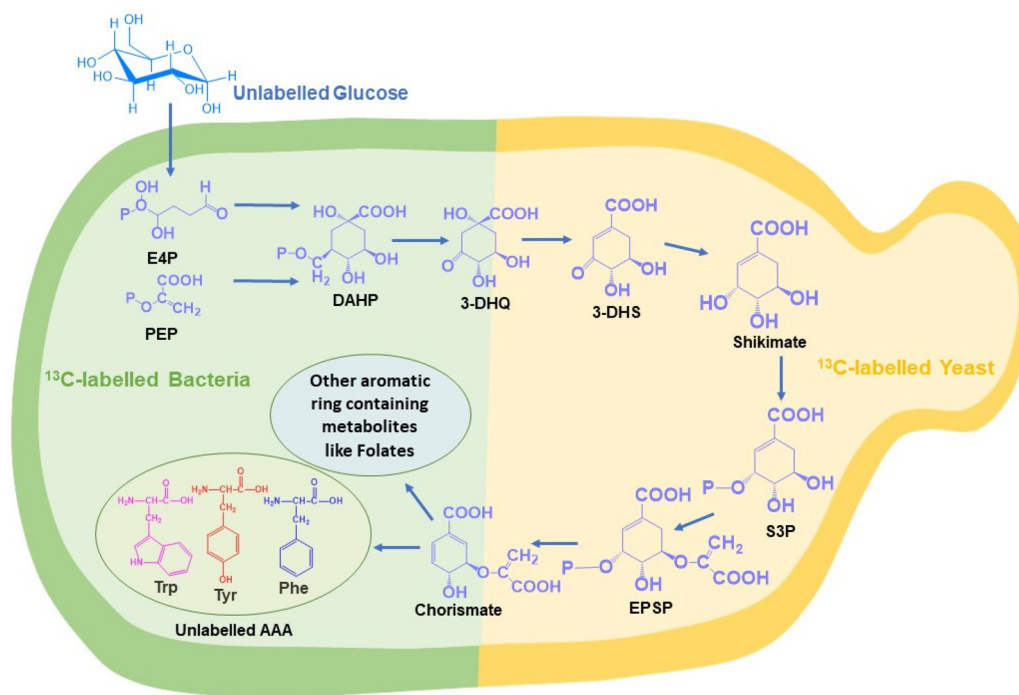


Fig. 1 Reverse labelling of the shikimate pathway occurring in both ^{13}C -labelled bacteria (green) and ^{13}C -labelled yeast (yellow) showing the carbon flow from unlabelled glucose to the synthesis of aromatic amino acid and aromatic ring-containing metabolites. (Abbreviation: E4P – erythrose-4-phosphate; PEP – phosphoenol pyruvate; DAHP – 3-deoxy-o-arabino-hepulosonate 7-phosphate; 3-DHQ – 3-dehydroxyquinic acid; 3-DHS – 3-dehydroshikimate; S3P – shikimate-3-phosphate; EPSP – 5-enolpyruvylshikimate 3-phosphate; Phe – phenylalanine; Tyr – tyrosine; Trp – tryptophan; AAA – aromatic amino acid.)

In this study, for the first time, we report *in situ* monitoring of the activity of the shikimate pathway at a community level in *Escherichia coli* (a prokaryotic microbe) and at a single-cell level in *Saccharomyces cerevisiae* (a eukaryotic microbe). The primary reason for using these two model systems is their well-known utility in industrial fermentation and a wide range of biotechnological applications. We here report the visualization of shikimate pathway activity *via* sensing the spatial distribution of nascent phenylalanine peaks through hyperspectral Raman imaging combined with RrSIP with time. We also performed multivariate curve resolution (MCR) to extract each component with interpretable spectra from a complex mixture of spectral data. This approach of Raman spectroscopy and imaging combined with RrSIP can be a promising analytical tool to explore different metabolic pathways involving aromatic rings.

Experimental methods

Microbial rSIP culture conditions and sample preparation

In our study, the bacterial strain *Escherichia coli* K12 and yeast strain *Saccharomyces cerevisiae* were used. For the primary culture, a single colony was inoculated in a carbon-free growth medium with 5 g L^{-1} of uniformly labelled ^{13}C glucose (Cambridge Isotope Laboratories) as the sole carbon source. This secondary broth culture was incubated overnight accord-

ing to standard microbial culture protocols. An aliquot of the ^{13}C -labelled microbial cells was taken and centrifuged at 7000g. Subsequently, the cells were washed twice with phosphate-buffered saline (PBS, Sisco Research Laboratories) to eliminate any residual traces of the ^{13}C -labelled growth medium. Immediately after the washing steps, the cell pellets were reintroduced into the growth medium, this time with 5 g L^{-1} of unlabelled (^{12}C) glucose (Sigma-Aldrich) as the exclusive carbon source. This culture was then subjected to standard incubation protocols. An inhibition assay was conducted using glyphosate, a well-known shikimate pathway inhibitor (Sigma-Aldrich). A working concentration of $700\text{ }\mu\text{g mL}^{-1}$ glyphosate was introduced into M9 minimal media containing 5 g L^{-1} unlabelled carbon before inoculating the ^{13}C -labelled cells.¹¹ Following a similar procedure to that with the mentioned glyphosate treatment, a supplement of three aromatic amino acids of concentration $30\text{ }\mu\text{g mL}^{-1}$ was added to the growth medium. Subsequently, growth-monitoring experiments were performed for all cultures using the optical density method at 600 nm (OD_{600}). The values were recorded at various time points starting from 0 hour to 24 hours as shown in Fig. S1 and S2.†

Raman spectral measurement and analysis

Aliquots of microbial samples were collected from both control and treated groups at various time intervals, *i.e.* 0, 1, 2, 4, 8, 16 and 24 hours. Appropriate dilutions were performed to



maintain approximately equal numbers of cells. The sample was immediately centrifuged at 7000 rpm for 5 min, followed by washing twice with PBS. A similar procedure was performed with yeast samples. The cells were placed on a clean ethanol-washed slide and Raman spectra were acquired using a WITec confocal micro-Raman spectrometer, equipped with a 532 nm laser source and 100× objective lens with a numerical aperture of 0.8. A single spectrum was averaged over 5 accumulations with 10-second exposure time in the spectral range of 200 to 3200 cm^{-1} with 600 gr mm^{-1} diffraction grating. The Raman spectral data were pre-processed using MATLAB R2021B. Smoothing was applied using the Savitzky Golay filter; a 5th order polynomial was used for baseline correction and unit normalization was performed. All spectral plots and figures including two dimensional correlation maps were generated using Origin Pro 2023b software.

Single-cell Raman imaging and MCR analysis

For Raman imaging, yeast cells were collected using the procedure mentioned earlier. The cells were diluted to an optimal concentration to make sure single cells can be obtained. After dilution, 10 μL was spread on an ethanol-washed calcium fluoride substrate and was subjected to Raman imaging using the above mentioned WITec confocal micro-Raman spectrometer. The step size of the scan was maintained at 0.33 microns with 5 s laser exposure time. The Raman hyperspectral data were pre-processed and the Raman image was generated using in-house scripts in MATLAB 2021B software. Multivariate curve resolution analysis (MCR) was performed using RamApp.¹² MCR analysis was initiated to resolve the complex spectral matrix into two components while these two MCR components retain their interpretable spectral information.

Results and discussion

In situ sensing of shikimate pathway activity and monitoring the *de novo* synthesis of phenylalanine using Raman spectroscopy

The shikimate pathway is directly related to the biosynthesis of aromatic amino acids. Therefore, the signal for phenylalanine was taken as the representative signature peak. The signal shows a strong unique peak in the Raman spectra of cells at 1003 cm^{-1} for unlabelled phenylalanine and at 967 cm^{-1} for ¹³C-labelled phenylalanine, as detailed in Table 1. This peak

arises due to the ring-breathing vibration. In the Raman reverse stable isotope probing approach, the ¹³C-labelled *E. coli* cells were allowed to grow in unlabelled media and their Raman spectra at various time points of 0, 1, 2, 4, 8, 16 and 24 h post-incubation were acquired, as shown in Fig. 2(A). Initially, at 0 h, the peak intensity at 967 cm^{-1} is high and very low at 1003 cm^{-1} . This indicates that nearly the entire aromatic amino acid pool inside cells is ¹³C-labelled. However, with the increase in the incubation time, we observed dynamic changes in the peak intensity. The peak intensity at 967 cm^{-1} gradually decreases while simultaneously there is a gradual increase in peak intensity at 1003 cm^{-1} . At 1 and 2 h, we observed a small signal at 1003 cm^{-1} ; this blue shift in the phenylalanine peak indicates the incorporation of ¹²C into the newly synthesized phenylalanine from the supplemented unlabelled glucose, which suggests positive shikimate pathway activity in the *E. coli* cells. At 4 h post-incubation, we observed that the intensity of both bands is nearly equal. At the later time points of 8, 16 and 24 h, a significant decrease in the peak intensity at 967 cm^{-1} and an exponential increase in the peak intensity at 1003 cm^{-1} can be observed. This blue shift of the peak with time can be interpreted as the old pool of aromatic amino acids becoming moderately substituted by the newly synthesised aromatic amino acids *via* the active shikimate pathway in the cells. This can be better visualized in the line plots shown in Fig. 2(B) and (C), which represent dynamic intensity *vs.* time relationships showing the incorporation of ¹²C into the phenylalanine pool of the cells. This enables us to track the carbon flow from the supplement carbon source to the aromatic amino acid through the shikimate pathway. The time-dependent change in the ratiometric intensity of 1003/967 cm^{-1} is plotted in Fig. 2(C), which shows the turnover of the nascent unlabelled phenylalanine in the cells.

The next analysis was completed to establish and validate that the origin of the blue-shifted phenylalanine reference peak is *via* the shikimate pathway only. To achieve this, we monitored the growth of cells with and without glyphosate treatment. Glyphosate (*N*-[phosphonomethyl]-glycine) is a well-known inhibitor of the shikimate pathway in *E. coli*.^{13,14} To check the inhibitory action of glyphosate, OD₆₀₀ values from the treated culture were plotted against the incubation time, as shown in Fig. S1.† Post-treatment, we can see that the OD₆₀₀ values do not increase with the incubation time, confirming the inhibitory effect of glyphosate on the phenylalanine synthesis of *E. coli*. Post-treatment Raman spectra from the cells were recorded at time points of 0, 1, 2, 4, 8, 16

Table 1 Band assignment for biomolecular bonds associated with shikimate pathway sensing

¹³ C-labelled peaks (cm^{-1})	Unlabelled peaks (cm^{-1})	Assignment	Significance	Ref.
622	644	Tyrosine	622 cm^{-1} is overlapped by unlabelled phenylalanine and ¹³ C-labelled tyrosine. Possible qualitative marker	15–18
967	1003	Phenylalanine	Evident qualitative and quasi-quantitative marker	10, 15–17, 19–21
1317	1337	Tryptophan	Band overlap by adenine and guanine. Possible qualitative marker	15–17



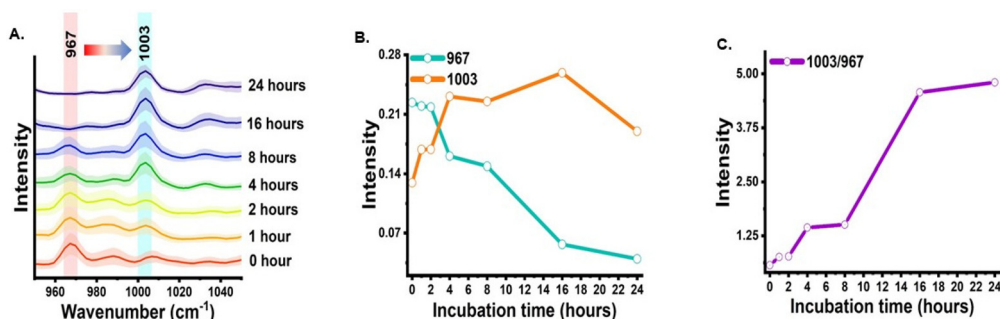


Fig. 2 Time-dependent sensing of shikimate pathway activity using the representative peak of phenylalanine via RrSIP. (A) Mean Raman spectra with standard deviation of *E. coli* at a specific range showing the blue shift of the phenylalanine peak from 967 cm⁻¹ toward 1003 cm⁻¹ with time. (B) Intensity plot showing incorporation of ¹³C into newly synthesized phenylalanine. (C) Intensity ratio plot showing the turnover of nascent phenylalanine replacing the old pool of phenylalanine with time.

and 24 h post-incubation and are shown in Fig. 3(A). As can be seen, the blue shift of the phenylalanine peak stopped due to the disruption in shikimate pathway activity. There are negligible changes in peak intensity at 1003 cm⁻¹ with the increase in incubation time, which is expected to derive from the nascent phenylalanine when compared with the peak intensity of the untreated group, as shown in Fig. 2(A). This observation suggests that the assigned peak at 1003 cm⁻¹ indeed has its origin from shikimate pathway activity.

Furthermore, we also performed a complementary analysis by reversing the inhibitory effect of glyphosate on the shikimate pathway of the cells. The glyphosate-treated growth medium was exogenously supplemented with the three unlabelled aromatic amino acids that are essential for cell sur-

vivability. The Raman spectra of all three supplemented unlabelled aromatic amino acids were recorded (Fig. S3†) to ensure the purity and peaks of interest. The Raman spectra from the cells were recorded at various time points of 0, 1, 2, 4, 8, 16 and 24 h, as shown in Fig. 3(B). We successfully monitored the incorporation of exogenous aromatic amino acids from the growth media into the glyphosate-treated cells by Raman spectroscopy in an extraction-free manner. At 1 h, we noticed a rapid increment of intensity at 1003 cm⁻¹ when compared to that of the untreated cells. However, with the incubation time, the intensity increases gradually. It was seen that this increase in the peak at 1003 cm⁻¹ was not accompanied by the peak signal diminishing at 967 cm⁻¹ after 1 h and so on. This indicates that there is no blue shift in the phenyl-

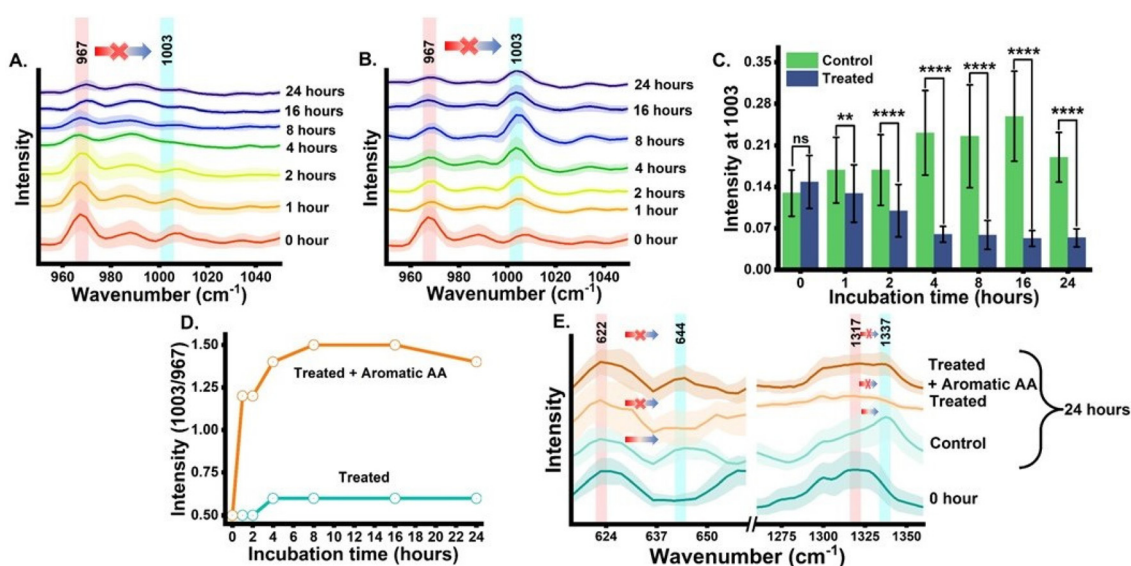


Fig. 3 Validating the peak from the nascent aromatic amino acid in *E. coli*. (A) Time-dependent mean Raman spectra with standard deviation of *E. coli* treated with glyphosate. (B) Time-dependent mean Raman spectra with standard deviation of glyphosate-treated *E. coli* supplemented with exogenous aromatic amino acid. (C) Intensity bar plot at 1003 cm⁻¹ of control versus the treated group showing statistical significance (ns ($p > 0.05$), * ($p \leq 0.05$), ** ($p \leq 0.01$), *** ($p \leq 0.001$), and **** ($p \leq 0.0001$)). (D) Intensity line plot at 1003 cm⁻¹ of the glyphosate-treated group vs. the aromatic amino acid-supplemented glyphosate-treated group. (E) Mean Raman spectra at 0 h and 24 h showing the blue shift of tyrosine and tryptophan peaks under different treatment conditions at 0 and 24 h.



Raman hyperspectral imaging combines spectroscopy with imaging, which provides information not only about the bio-

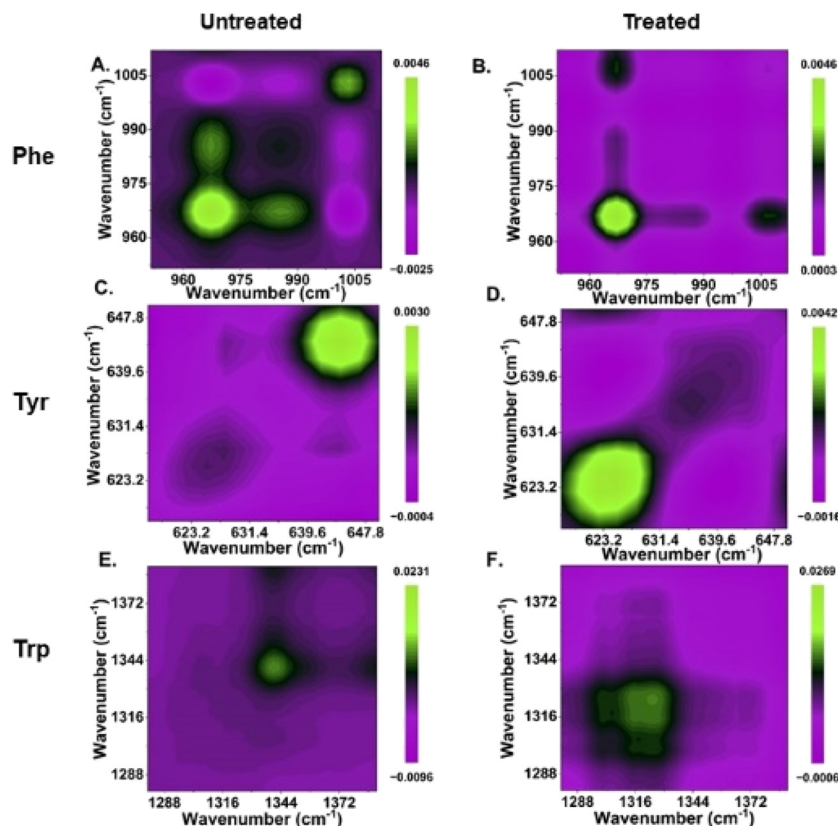


Fig. 4 Two-dimensional synchronous correlation map of untreated and treated cells. (A and B) phenylalanine; (C and D) tyrosine; (E and F) tryptophan. Each map includes the mean Raman spectra of different incubation time points to provide the spectral dynamics over the perturbation range. The green signal represent positive while magenta signal represent negative.

molecular components but also their spatial distribution within the subcellular resolution at different pixel positions of the Raman image. We mapped the peak intensity of phenylalanine at 967 and 1003 cm^{-1} from the single *S. cerevisiae* cell at various time points of 0, 2, 4, 8 and 24 h, as shown in Fig. 6 (A). The time points for Raman imaging were decided by analysing the notable changes in the mean Raman spectra of the single cell, as shown in Fig. 5(A). The bright field and the corresponding Raman image of the C–H band show the cellular boundary. The Raman image of the cell at 0 h shows a highly intensified image at 967 cm^{-1} with respect to that at 1003 cm^{-1} , indicating the initial phenylalanine pool to be ^{13}C -labelled. This distribution of phenylalanine comes from the free phenylalanine of the cytoplasm as well as the protein structures. The Raman image of the cell at 2 h post-inoculation shows the nearly equivalent spatial intensity distribution of phenylalanine at 967 and 1003 cm^{-1} . This confirms positive shikimate pathway activity and nearly equivalent distribution of the ^{13}C -labelled and unlabelled phenylalanine in the cellular phenylalanine pool, as shown in the mean Raman spectra from a single cell in Fig. 5(A). Likewise, with the increase in the incubation time, we see that the spatial distribution of the ^{13}C -labelled phenylalanine diminishes while that of the unlabelled nascent phenylalanine increases confirming the

visualization of the phenylalanine turnover dynamics. We also noticed that a newly formed bud of the budding yeast showed an intensified signal at 1003 cm^{-1} , giving insight into the carbon flow and newly synthesized aromatic amino acids in the bud connected to the mother cell. The Raman image was constructed targeting the distribution of the single peak intensity in each pixel. Multivariate curve resolution (MCR) was performed to resolve the complex mixture of spectral data of the Raman image into various components by conserving the chemical meaning in each component. We observed that the two components had most of the information of interest for understanding the biomolecular spectral pattern. For quenching out the biomolecular Raman spectral pattern from the mixed complex hyperspectral Raman data, MCR analysis with two components was performed. Among the two MCR components under study, one component had phenylalanine and other biomolecular information of the cell of interest and the other one had a negligibly interpretable biomolecular spectral pattern that consists of background information, as shown in Fig. 6(B). An MCR 1 component map at various time points showed a similar cell boundary and pattern when compared to the bright field images shown in Fig. 6(A). The corresponding MCR1 component loading plot can be seen with the phenylalanine peak intensity marked in Fig. 6(B). We can see the



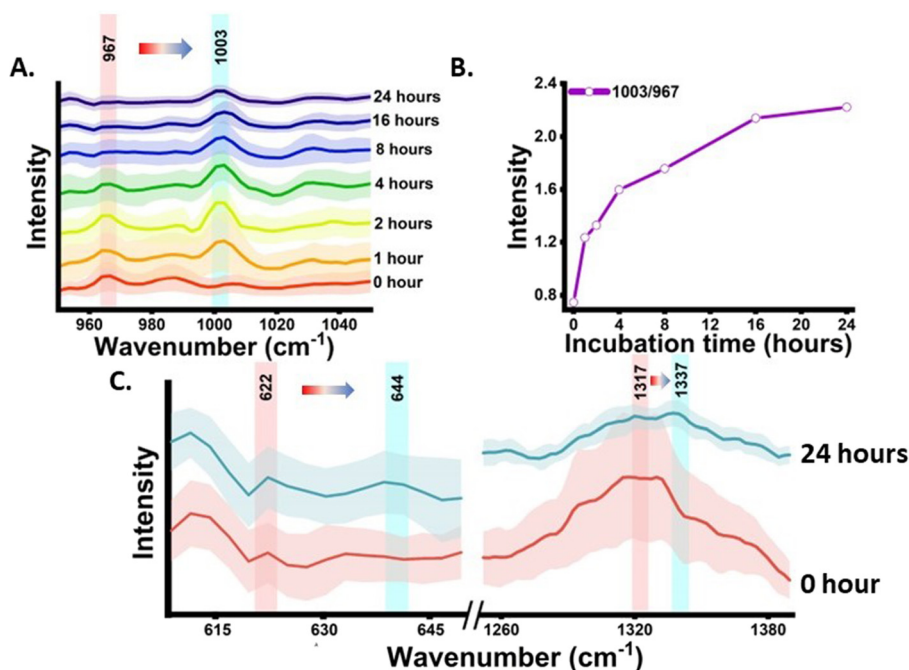


Fig. 5 Time-dependent sensing of shikimate pathway activity using the representative peak of phenylalanine via RrSIP. (A) Mean Raman spectra with standard deviation of a single *S. cerevisiae* cell at a specific range showing the blue shift of the phenylalanine peak from 967 cm^{-1} toward 1003 cm^{-1} with time. (B) Raman intensity ratio plot showing the turnover of nascent phenylalanine replacing the old pool of phenylalanine with time in a single *S. cerevisiae* cell. (C) Mean Raman spectra showing the blue shift of tyrosine and tryptophan peaks at 0 and 24 h.

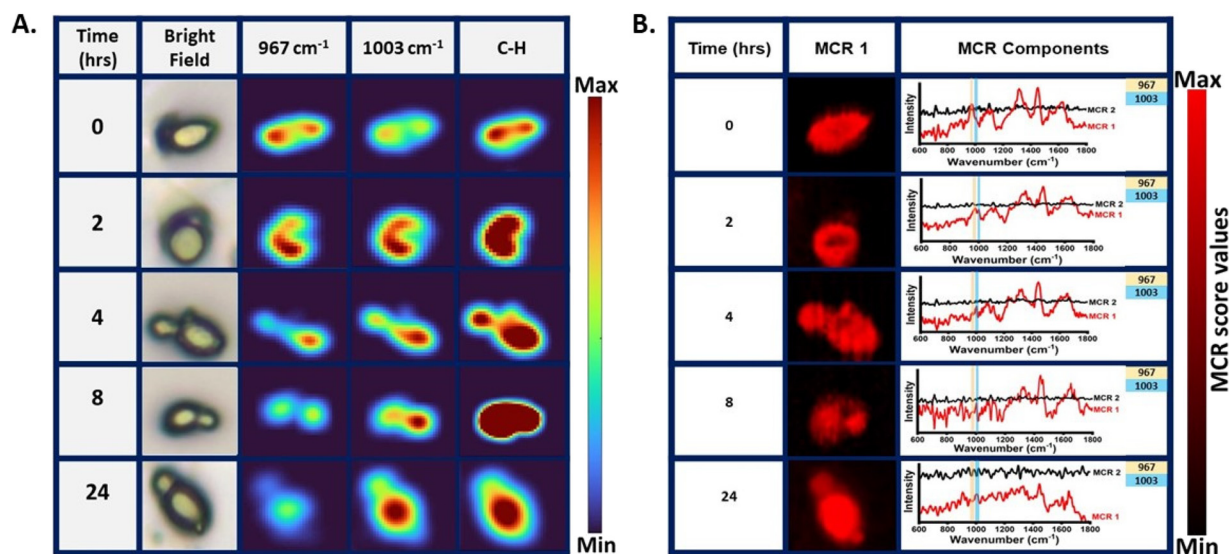


Fig. 6 Raman image of a single yeast cell. (A) Raman single-cell level visualization of phenylalanine turnover and shikimate pathway activity with incubation time in the yeast cell showing a bright field image for reference and the corresponding Raman image at 967 cm^{-1} (^{13}C -labelled phenylalanine), 1003 cm^{-1} (unlabelled nascent phenylalanine) and C–H stretch (defining Raman image of the cell boundary). (B) Raman image constructed from multivariate curve resolution analysis for MCR 1 and the corresponding loading plots of MCR components in the bio-fingerprint region from 600 to 1800 cm^{-1} showing the blue shift in the phenylalanine band with time in the MCR 1 component.

dynamics in the phenylalanine bands at 967 cm^{-1} and 1003 cm^{-1} with time at the single-cell level.

Overall findings suggest that Raman spectroscopy and hyperspectral imaging combined with reverse stable isotope probing could serve as an extraction-free and non-destructive

approach for metabolic pathway monitoring. Previously, Weng *J. et al.* have reported high single-cell phenotypic heterogeneity of carbon uptake and hydrogen uptake in carotenoids using Raman stable isotope probing (RSIP).²² Li *J. et al.* using Raman active cell sorting SIP identified, sorted and isolated



the active toluene degrader single cell *Pigmentiphaga*.²³ Li Menqui *et al.* have shown the potential of RSIP using phenylalanine and thymine as Raman spectral biomarkers to track the carbon flow in the predator–prey model at the single-cell level.²⁰ Although these studies explored the possibilities of RSIP, we considered RrSIP a cost-effective approach with respect to RSIP, as preparing the labelled cell for inoculation, and growing in unlabelled media will require an extremely low amount of stable isotope. Additionally, employing RrSIP reveals a blue shift in the Raman spectra, signifying the migration of labelled metabolite peaks toward the position of unlabelled metabolite peaks over time. This shift can be readily confirmed, as the positions of unlabelled bands in the cell spectra have been thoroughly investigated in previous studies.^{10,21} In one of our recent studies, we reported the utility of this approach for global proteome monitoring.¹⁰ Due to the easy availability and low cost of unlabelled supplements, it is an efficient approach for requirements to working with large growth medium volumes. This offers an attractive opportunity in the future for sensing and visualizing metabolites and metabolic pathways inside cells *in situ* or *in vivo*. As cells have carbon everywhere at the genome, proteome, and metabolome levels, carbon-based RrSIP can give insight into sensing the *de novo* synthesis and turnover of different metabolites without any requirement for extraction. In our work, we have described its potential in prokaryotic and eukaryotic model systems. A remarkable finding in this work is the observation that the phenylalanine peaks at 967 and 1003 cm⁻¹ are established as Raman metabolic spectral markers for studying shikimate pathway activity, *de novo* phenylalanine synthesis, and the turnover dynamics of nascent phenylalanine over time. We have verified that the signal arising from newly synthesized unlabelled phenylalanine originates solely from the shikimate pathway *via* an inhibition assay and bypasses the effect of inhibition with exogenous supplements.

One of the multiple beneficial attributes of this approach is that along with the 1D temporal spectral information, spatial information can also be derived using Raman hyperspectral imaging. This provides an image representation of the sample with different pseudo-colours that correspond to the intensity of the biomolecular peak for visualization and quantification. Along with the identification, it also gives information about the heterogeneity of the scanned area without the requirement of endogenous and exogenous labels or markers, unlike conventional fluorescence-based methods. The endogenous labels like fluorescent proteins and exogenous labels like dyes (Nile red and BODIPY) or quantum dots can interfere with the metabolic activity or can cause photobleaching or phototoxicity.^{7,24–29} Raman hyperspectral imaging combined with multivariate curve resolution validates the sensitivity and specificity of our approach. Visualizing the spatial distribution and extracting the spectral pattern of phenylalanine from intricate spectral data not only contributes to the validation of our study but also enhances our ability to explore shikimate pathway activity at various incubation times *in situ*. Furthermore, we have identified peak positions linked to tyro-

sine and tryptophan, which can serve as potential qualitative markers. This identification has been further corroborated through the application of two-dimensional correlation spectroscopy, adding validation to our findings.

Conclusions

Overall, the findings of our study demonstrate the immense potential of this combinatorial approach in objectively unraveling the dynamics and complexities of metabolic pathways. The presented approach is cost-effective and has potential to be translated into a preferred method for monitoring the production of commercially or medically relevant metabolites and byproducts.

Author contributions

Jiro Karlo – conceptualization, investigation, methodology, formal analysis, software, visualization, and writing – original draft. Aryan Gupta – investigation, methodology, and formal analysis. Surya Pratap Singh – resources, funding acquisition, supervision, and writing – review & editing.

Conflicts of interest

The authors have no conflicts of interest to declare.

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