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4-Hydroxy-2-Oxoglutaric Acid, A Key Metabolite Involved in Trypsin-Regulation of Arginine Metabolism in *Hylocereus undatus* during Storage

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Received: 06 February 2024 Accepted: 28 March 2024 Published: 28 May 2024

ABSTRACT

Trypsin, a novel superoxide scavenger, significantly enhances the storage quality of *Hylocereus undatus* (*H. undatus*). To elucidate the preservation mechanism of trypsin on *H. undatus*, a widely targeted metabolomic analysis, and transcriptomics analysis were conducted. Firstly, a total of 453 metabolites were identified, with organic acids and their derivatives constituting the largest proportion (25%). Amino acids and their metabolites, prominent among organic acids, were further analyzed. Among them, 73 metabolites were associated with amino acids, and 37 exhibited significant differences. The most enriched Kyoto Encyclopedia of Genes and Genomes (KEGG) pathway was arginine biosynthesis (map00220), with polyamine metabolites showing the most pronounced differences, particularly spermine (FC = 1.7594). Compared with the control group, 4-hydroxy-2-oxoglutaric acid was significantly upregulated (FC = 2.117) in the process of spermine biosynthesis. Furthermore, the results of Gene Ontology (GO) and KEGG enrichment analysis of the *H. undatus* transcriptome profile revealed that trypsin treatment led to 187 differentially expressed genes associated with arginine. Both GO and KEGG analyses exhibited significant enrichment in the spermine biosynthetic process (GO:0006597) (map:00220) within the arginine biosynthesis pathway. Moreover, most enzymes and metabolites within the spermine biosynthesis pathway in *H. undatus* were upregulated. The results of the PPI network highlighted that ADC, SPDS, and SAMDC, among others, were pivotal proteins involved in trypsin-regulated arginine metabolism and spermine synthesis. This study revealed that trypsin could significantly delay postharvest senescence of *H. undatus* at room temperature. This effect might be attributed to trypsin triggering the synthesis of 4-hydroxy-2-oxoglutaric acid in the fruit peel, thereby promoting the biosynthesis of spermine and other polyamines.

KEYWORDS

Trypsin; arginine; transcriptomics; widely targeted metabolomics

1 Introduction

Hylocereus undatus (*H. undatus*), belonging to the Cactaceae family, contained plant-based albumin and anthocyanins, along with abundant vitamins and water-soluble dietary fiber—components rarely found in most plants. It boasted effects such as preventing arteriosclerosis, detoxifying and protecting the stomach, as well as skin whitening and weight loss [1]. These qualities have earned it significant popularity among



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consumers. However, after being harvested, *H. undatus* often succumbed to a range of diseases, leading to fruit damage and a loss of commercial value [2]. Therefore, research aimed at enhancing postharvest fruit quality serves as a reference for future studies on the storage of fruits and vegetables.

Most researchers considered reactive oxygen species (ROS) to be the primary cause of inducing fruit decay and deterioration [3]. Under normal circumstances, the production and elimination of ROS within plant cells remained balanced, preventing damage to the fruit. However, as the storage period lengthens, the content of oxygen free radicals increases, disrupting the equilibrium of ROS within cells. This led to lipid peroxidation of membranes, damage to biomolecules such as proteins and nucleic acids, and affected the normal physiological metabolism of the fruit. Ultimately, this resulted in the decay and deterioration of the fruit [4,5]. Plants had complex antioxidant mechanisms to eliminate excessive ROS, and the strength of their antioxidant capacity primarily depended on the content of various small molecules and the activity of antioxidant enzymes [6]. Research indicated that polyamines (PAs), such as putrescine, spermidine, and spermine, can function as antioxidants within plants under certain stressful conditions [7].

The antioxidant effects of PAs were attributed to their inherent polycationic nature, which enabled them to participate in regulating cellular ion balance and interacting with polyanionic molecules. This interaction prevented the degradation of large molecules and protected cells from oxidative damage. They collectively mitigated plant cell aging caused by superoxide radicals. Additionally, Liu et al. [8] demonstrated that PAs enhance plant drought tolerance by modulating K⁺ channels to regulate stomatal opening; Cuevas et al. [9] used *Arabidopsis* mutants to show that endogenous PAs levels can enhance a plant cell's response to low-temperature conditions; Roy et al. [10] found that rice transformed with the oat arginine decarboxylase (ADC) gene led to PAs accumulation in line with increased ADC activity, subsequently enhancing plant salt tolerance. However, as of now, the influence of PAs compounds present in *H. undatus* peel on its quality during storage remains to be elucidated.

Amino acids or certain PAs synthesized from amino acid precursors could enhance a plant's adaptability to stress conditions. Amino acids are important physiological substances within plants, playing a vital regulatory role in growth processes. When faced with stress and adversity, plants control the absorption, synthesis, and degradation of amino acids to mitigate the harm caused by these conditions. In recent years, research on how amino acids enhance plant stress resistance has garnered widespread attention. Proline and amino butyric acid, in particular, have been extensively studied, and some progress has been made in understanding their mechanisms [11]. While arginine was typically considered a crucial nitrogen storage nutrient for reuse, recent studies have indicated that arginine's enhancement of the abiotic stress tolerance of plants may be achieved through the production of PAs and nitric oxide (NO) through metabolism [7]. Furthermore, nitrogen nutrition could be supplied by other amino acids, but these other amino acids couldn't generate PAs and NO [11]. Nitric oxide synthase (NOS), ADC, and arginase were key enzymes in L-arginine catabolism. The relative strengths of the activities of these three enzymes could determine the metabolic direction of arginine. Arginine is transformed through metabolism to form important signaling molecules, potentially holding more significance for plant growth, development, and tolerance formation compared to being just a regular nitrogen nutrient [12,13]. From this perspective, it became evident that both arginine and its metabolic products play a crucial physiological role throughout the entire lifecycle of plants.

In recent years, preservation methods could be categorized into two main groups: physical [14,15] and chemical methods [16]. However, as awareness of food safety increased, postharvest storage and environmentally friendly preservation of fruits and vegetables became especially important, leading to the gradual development of biological preservation techniques. Our laboratory discovered that trypsin has a positive effect on superoxide scavenging [17,18]. Trypsin, an enzyme belonging to the serine protease

family, has been reported to effectively scavenge superoxide anions and provide excellent protection to membrane lipids, thereby preventing cell damage [19–21]. Additionally, trypsin could enhance the biocontrol function of pseudohyphal yeast, inhibit mold growth on the surface of fruits and vegetables, and extend their shelf life. However, whether amino acid metabolism is involved in the process of trypsin-mediated preservation of fruits and vegetables has not been addressed in any reports.

This study integrated metabolic and transcriptomic data from *H. undatus* peel samples treated with trypsin and control groups, across different storage times. Within the metabolic dataset, amino acid-related differentially expressed metabolites were screened. Variable Importance in Projection (VIP) value analysis using various algorithms was employed to showcase the expression patterns of metabolites across different samples, as well as the VIP values in multivariate statistical analysis and the *p*-values in univariate statistics. This analysis was conducted to uncover significant metabolic pathways and their crucial differential metabolites, and functional analysis of these metabolites was performed. In combination with transcriptomic data, core regulatory genes were identified, and interaction networks among proteins were studied. At the RNA level, the regulatory mechanism of trypsin-induced amino acid metabolism was investigated. This research aimed to further elucidate the preservation mechanism of *H. undatus* postharvest by trypsin, providing additional theoretical support to the field of biopreservation techniques.

2 Materials and Methods

2.1 *H. undatus* Treatment Methods

Thirty fruits (approximately 15 cm in diameter) of *H. undatus* were divided into two groups: the trypsin group and the control group. The pericarp of each fruit was evenly sprayed with either a trypsin solution (2.41×10^{-6} mol/L) or distilled water. Each group, including the experimental and control groups, consisted of three biological replicates. The samples were placed at room temperature (25°C) and observed, photographed, and recorded, the water loss rate was measured, and color difference was assessed every 24 h. After one week of storage, *H. undatus* pericarp samples were immediately frozen in liquid nitrogen and stored at -80°C for subsequent analysis.

2.2 Quality of Fruit

The weight loss rate and hardness of samples were detected and recorded during 10 days of storage at 25°C. The color of *H. undatus* peels was determined at room temperature by colorimeter (X-Rite Color i5, Germany). The value of ΔE was calculated as the following formula [20,21]:

$$\Delta E = \sqrt{\Delta L^2 + \Delta a^2 + \Delta b^2}$$

2.3 Omic Analyses

Sample collection, sequencing, compound identification for omics analysis, selection of differential compounds and differential genes, and the analysis methods for data such as Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) were conducted following the descriptions outlined in our previously published papers and those of other researchers [20,21].

2.4 Protein-Protein Interaction (PPI) Network Analysis

Using a network modeling approach, a protein-protein interaction network was constructed to elucidate the relationships between proteins. Starting from a network perspective, topology attributes of the network were analyzed to extract significant protein interaction relationships from complex biological data. The protein interaction network construction was carried out using the Majorbio Cloud platform, and the

resulting network was visualized using Cytoscape [22,23]. The MCODE and CytoHubba plugins in Cytoscape were employed for cluster analysis and the identification of key nodes.

2.5 Quantitative Real-Time PCR Analysis

Reverse transcription-qPCR (RT-qPCR) was used to check the result of RNA-seq as described by Yang et al. [24]. Gene β -actin was used as an internal control to normalize gene expression in *H. undatus*. The relative copy numbers of the genes were calculated by the $2^{-\Delta\Delta C_t}$ method [24,25].

2.6 Statistical Analysis

To assess differences between groups, Partial Least Squares Discriminant Analysis (PLS-DA) was performed, and metabolites were selected based on $p(\text{corr})$ combinations. VIP values were calculated using OPLS-DA analysis. All experiments were conducted in triplicate. Statistical analysis was conducted using the SPSS statistical software package (version 11.0.1). Paired sample t -tests were used to determine the significance of differences between samples at specific time points. Significance levels were evaluated as $p < 0.05$ for significant differences and $p < 0.01$ for highly significant differences.

3 Results

3.1 Influence of Trypsin on the Storage Quality of *H. undatus*

During the initial stages of storage, all groups of *H. undatus* exhibited good quality with vibrant color (Fig. 1A). After 6 days of storage, the control group's fruit scales were completely dried out, appearing dull in color, and the fruit bodies were visibly infected by pathogens, showing extensive diseased spots (Fig. 1A). In contrast, the trypsin-treated group displayed partially dried scales, particularly noticeable at the top; however, the fruit skin maintained its vibrant color without evident signs of fungal growth (Fig. 1A).

As the storage time extended, the weight loss rate for all groups exhibited a significant upward trend (Fig. 1B). The average fresh weight reduction of the control group's fruit was relatively high, decreasing by 0.95% per day, while the trypsin-treated group showed weight reduction rates of 0.74% per day (Fig. 1B). The difference between the trypsin-treated and control groups was notably distinct. Throughout the storage process, the rate of change of the E value increased during the initial stages of storage and began to decline after the 4th day. Comparative color analysis indicated that the E value of trypsin-treated fruit changed at a slower rate, particularly in the first 4 days of storage, highlighting the favorable effect of trypsin treatment on maintaining the color stability of *H. undatus* under storage conditions (Fig. 1C). Trypsin treatment suppressed the loss of hardness during the entire storage (Fig. 1D).

The study results suggested that trypsin can effectively delay fruit dehydration, maintain higher hardness, significantly improve fruit quality, and extend shelf life.

3.2 Widely Targeted Metabolomics Analysis

In this study, metabolomics analysis was conducted to investigate amino acid-related metabolic pathways and identify key metabolites, to determine the regulatory effect of trypsin on their synthesis pathways.

3.2.1 Overview of Metabolomics Data

Based on the number of metabolites, a Human Metabolome Database (HMDB) classification was performed (Fig. S1A), where different colors in the pie chart represented various categories of metabolites, and the area depicted the relative proportion of metabolites within each category. Among these, organic acids and their derivatives accounted for 25% of the total, followed by heterocyclic compounds at 17.50%, lipids and lipid-like molecules at 15.83%, aromatic compounds at 10.00%, nucleosides, nucleotides, and their analogs at 10.00%, organic oxygen compounds at 9.17%, phenylpropanes and polyketides at 5.83%, organic nitrogen compounds at 4.17%, alkaloids and their

derivatives at 1.67%, and organic sulfur compounds at 0.83%. Organic acids and their derivatives were the largest category, with amino acids and their metabolites being the primary components. Consequently, amino acid-related metabolites within the category of organic acids were selected for further analysis. A total of 453 differential metabolites were identified, with 219 significantly different metabolites (p -Value < 0.05). The volcano plot illustrated these differences (Fig. S1B), where red circles indicated upregulated metabolites (113), blue circles represented downregulated metabolites (106), and gray circles indicated non-significant differences. To analyze differences between groups, the variability in metabolite composition and abundance among samples was quantitatively analyzed using correlation data between samples. In Fig. S1C, each cell represented the correlation between two samples, with high correlation values close to one indicating high similarity in metabolite composition and abundance between samples. The PLS-DA score plot visually demonstrates the classification effect of the model (Fig. S1D), where the contribution rate of PC1 (trypsin-treated group) was 70.20% and that of PC2 (control group) was 19.70%. The distinct separation of metabolite expression levels and group relationships indicated a significant classification effect.

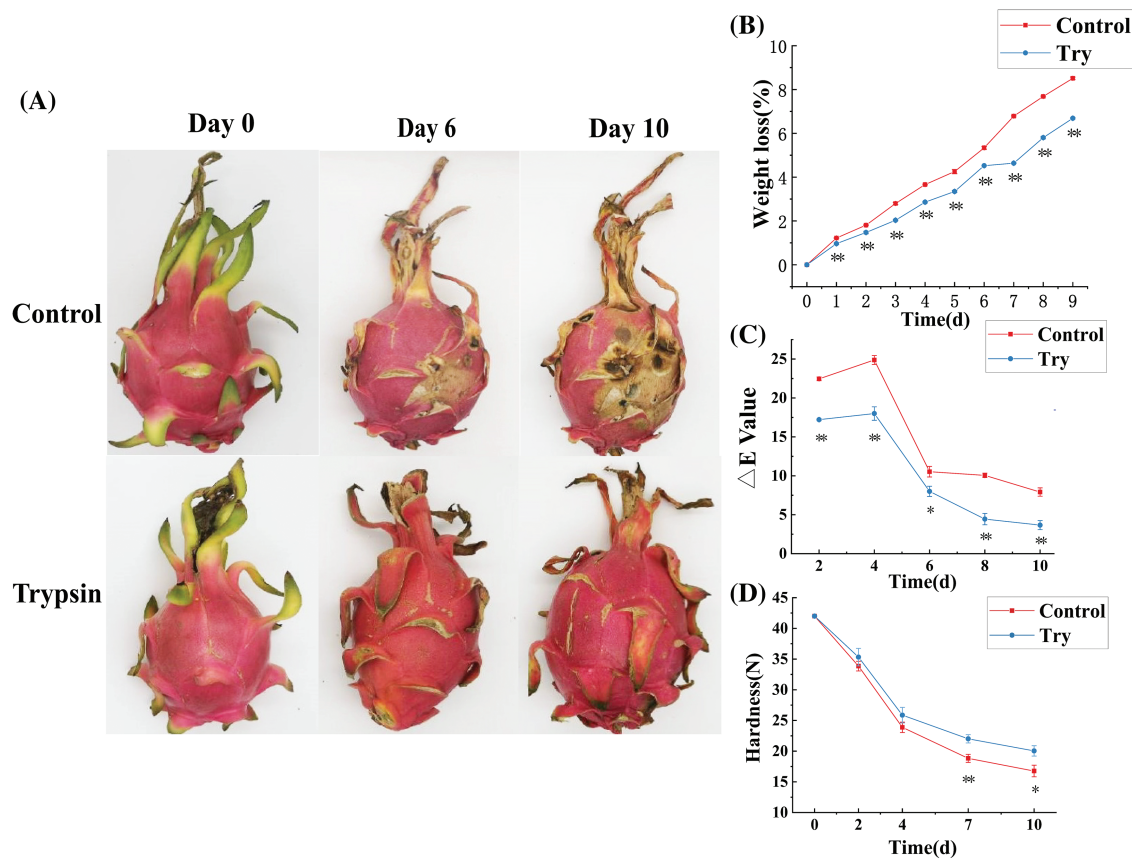


Figure 1: Effect of trypsin treatment on the storage quality of *H. undatus*. (A) The appearance changes of *H. undatus* stored at 25°C. (B) Weight loss of *H. undatus* fruit stored with or without trypsin for 9 d. (C) Changes of the E value of *H. undatus* fruit stored with or without trypsin at 25°C for 10 days. (D) The hardness of *H. undatus* fruit stored with or without trypsin for 10 d. Bars represent standard error. Respectively, * represents $p < 0.05$, and ** represents $p < 0.01$

3.2.2 Analysis of Amino Acid-Related Differential Metabolites

From the category of organic acids and their derivatives, a total of 73 amino acid-related metabolites were screened (Fig. 2A), with 56 metabolites upregulated and 17 metabolites downregulated. Among these, 37 metabolites showed significant differences. As shown in Fig. 2B, there was a positive correlation between L-(–)-tyrosine, spermine, histidine, and glutamine. Butyric acid, methylmalonic acid, citric acid, isopropylidene malic acid, and succinic acid showed a positive correlation, while they displayed a negative correlation with other amino acids.

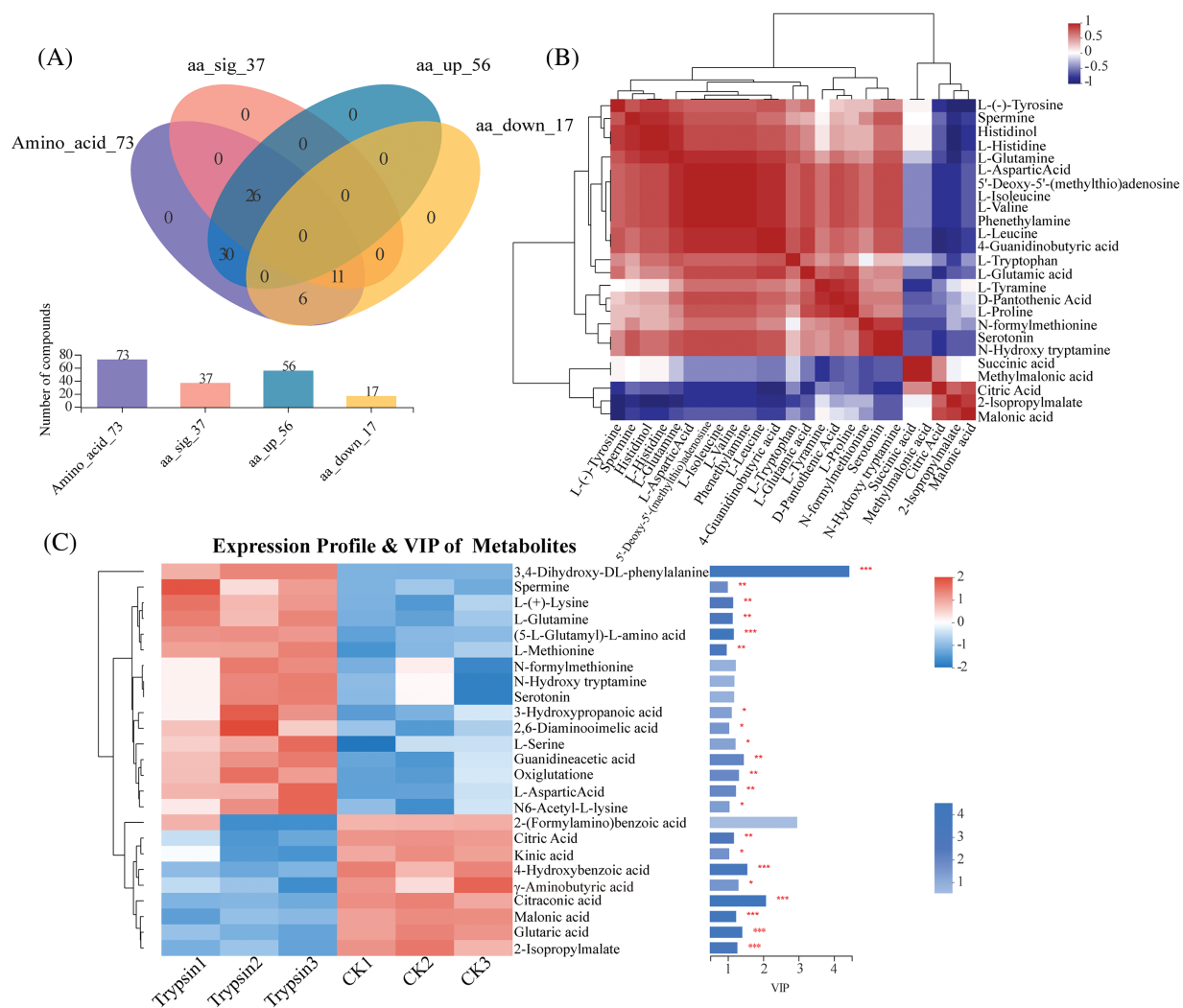


Figure 2: Metabonomic analyses of metabolites regulated by trypsin. (A) Metabolic set analysis Venn diagram. (B) Metabolite correlation analysis diagram. (C) 25 significantly different metabolites (VIP \geq 0.9) expression profile and VIP value. Respectively. * represents $p < 0.05$, and ** represents $p < 0.01$, and *** represents $p < 0.001$. Within the heatmap, red represents a positive correlation, and blue represents a negative correlation

Based on the threshold criteria of $|\log_2(\text{fold change})| \geq 1$ and $\text{VIP} \geq 0.9$, the top 25 differential metabolites were identified (Fig. 2C). Among these 25 differentially expressed metabolites, 16 amino acids were upregulated in the trypsin-treated group compared to the control group. These included 3,4-dihydroxy-DL-phenylalanine, spermine, L-(+)-lysine, glutamine, (5-L-glutaminy)-L-amino acid, L-asparagine, N-formylmethionine, N-hydroxytryptamine, serotonin, 3-hydroxypropionic acid, 2,6-diaminopropionic acid, L-serine, guanidine acetic acid, pentanedial, aspartic acid, and N6-acetyl-L-lysine. On the other hand, 9 amino acids were downregulated, including 2-(methylamino) benzoic acid, citric acid, isoleucine, 4-hydroxybenzoic acid, γ -aminobutyric acid, citric acid, succinic acid, glutaric acid, and isopropylidene malic acid.

3.2.3 Amino Acid-Related KEGG Enrichment Analysis

Through KEGG enrichment analysis of all amino acid-related metabolic sets (Table 1; Fig S2), the most enriched metabolic pathway was arginine biosynthesis (map00220) with 10 metabolites, followed by alanine, aspartate, and glutamate metabolism (map00250) with 10 metabolites, valine, leucine, and isoleucine degradation (map00410) with 9 metabolites, aminoacyl-tRNA biosynthesis (map00970) with 14 metabolites, and arginine and proline metabolism (map00330) with 17 metabolites (Fig. S2). Among these, the arginine and proline metabolism pathways involved the highest number of metabolites.

Table 1: KEGG enrichment analysis of metabolic pathways related to amino acids

Num.	Pathway description	Pathway ID	Ratio	<i>p</i> -value
17	Arginine and proline metabolism	map00330	17/73	0
15	ABC transporters	map02010	15/73	0
14	Aminoacyl-tRNA biosynthesis	map00970	14/73	0
10	Alanine, aspartate and glutamate metabolism	map00250	10/73	0
10	Arginine biosynthesis	map00220	10/73	0
10	Phenylalanine metabolism	map00360	10/73	0
9	Beta-alanine metabolism	map00410	9/73	0
9	Cysteine and methionine metabolism	map00270	9/73	0
7	Glutathione metabolism	map00480	7/73	0
7	Tryptophan metabolism	map00380	7/73	0.0001

The results indicated that not only the arginine metabolism pathway was significant, but its biosynthesis pathway was also highly significant. We speculated that arginine might be a key amino acid regulated by trypsin. Arginine metabolism leads to the production of PAs and NO. Generally, the accumulation of PAs decreases as fruits mature and age [26]. However, we found that compared to untreated *H. undatus* fruits, trypsin-treated *H. undatus* showed significant upregulation of spermine ($\text{VIP} = 0.9872$) during storage (Fig. 2C). This suggested that spermine could play an important role in the preservation effects of trypsin on *H. undatus*, as it is involved in regulating various physiological processes.

3.3 Transcriptomic Analysis

3.3.1 Sequencing Quality Analysis

Through the analysis of amino acid-related metabolites in the metabolomics data, we observed significant changes in the arginine and proline metabolism pathways. However, the differences in metabolites related to the proline pathway were not significant. In contrast, the arginine metabolism

pathway, including its biosynthesis, showed significant changes. This led us to speculate that arginine might be a key amino acid regulated by trypsin. To further investigate how trypsin regulates the biosynthesis and metabolism of arginine to achieve its preservation mechanism, we conducted an in-depth analysis of the transcriptome data.

In this study, we constructed two groups, an experimental group and a control group, each with three biological replicates, totaling six transcriptome libraries for transcriptional analysis. After sequencing, the Q20 (probability of base sequencing accuracy over 99.9%) accounted for over 98% of the entire read length, and Q30 (probability of base sequencing accuracy over 99.99%) accounted for over 94% of the entire read length. The GC content was 45%, and the sequencing error rate was below 0.1%. The sequencing results exhibited high quality, meeting the requirements for subsequent analyses.

3.3.2 GO Enrichment Analysis of Arginine-Related Genes

We conducted a GO annotation analysis on the set of arginine-related genes (ARGs) (Fig. 3), which included a total of 187 differentially expressed genes (DEGs). The DEGs were annotated using GO terms from categories including biological processes, cellular components, and molecular functions. More than 100 DEGs were associated with biological biosynthesis processes (GO:0009058) and metabolic processes (GO:0008152). The most significantly enriched GO term was the spermine biosynthetic process (GO:0006597), followed by spermine metabolic process (GO:0008215), S-adenosylmethionine decarboxylase activity (GO:0004014), spermine metabolic process (GO:0008216), spermine biosynthetic process (GO:0008295), and arginine biosynthetic process (GO:0006526), among others.

The GO terms of the ARGs regulated by trypsin were organized into a Directed Acyclic Graph (DAG) that depicted the relationships between GO terms (Fig. 3B). The relationships between terms are described as “is_a” and “part_of.” For instance, the GO biological process “arginine biosynthetic process (GO:0006526)” is a parent term of “arginine metabolic process (GO:0006525),” which, in turn, is a parent term of “spermine metabolic process (GO:0008215)” and “spermine metabolic process (GO:0008216).” Similarly, the MF term “S-adenosylmethionine decarboxylase activity (GO:0004014)” is a child term of “carboxy-lyase activity (GO:0016831).” The root term for BP ontology is traced back to the ancestral term GO:0008150.

3.3.3 KEGG Enrichment Analysis of Arginine-Related Genes

By conducting a combined KEGG enrichment analysis on both transcriptomic and metabolomic data, we identified the top 10 enriched KEGG pathways related to arginine ($p < 0.01$) (Fig. 3C, Table 2). The most enriched pathway in the transcriptomic data was “Arginine and proline metabolism (map00330)” with 69 genes enriched, followed by “Arginine biosynthesis (map00220)” with 51 genes enriched, which was consistent with the metabolomic results.

Furthermore, the expression changes of genes in the transcriptomic data within the enriched KEGG pathways were depicted in the chord diagram of the KEGG enrichment analysis (Fig. 3D). In the arginine and proline metabolism pathway (map00330), genes such as argininosuccinate synthase (ASS), spermidine synthase (SPDS), and S-adenosylmethionine decarboxylase precursor (SAMDC), which were involved in the biosynthesis of spermine, were significantly regulated either up or down, suggesting their pivotal roles as central nodes and complex regulatory mechanisms through interactions in signaling pathways. Other enriched pathways included “Arginine biosynthesis (map00220),” “Beta-alanine metabolism (map00410),” “Alanine, aspartate, and glutamate metabolism (map00250),” “Limonene and pinene degradation (map00903),” “Histidine metabolism (map00340),” “Tryptophan metabolism (map00380),” “Lysine degradation (map00310),” “Ascorbate and aldarate metabolism (map00053),” and “Valine, leucine, and isoleucine degradation (map00280).”

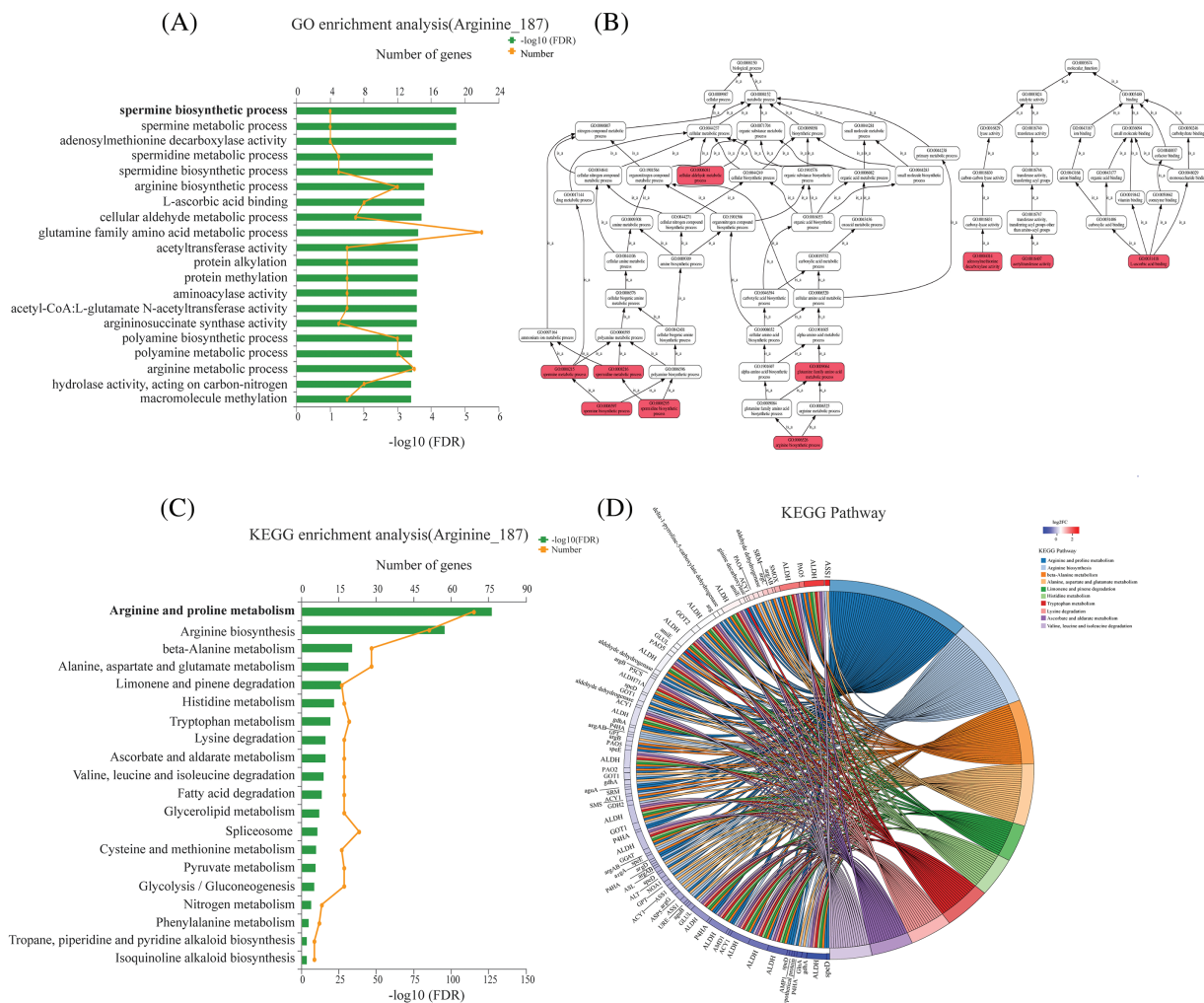


Figure 3: GO and KEGG enrichment analysis of ARGs. (A) GO enrichment analysis diagram of ARGs. (B) Relationships between GO terms in a directed acyclic graph (DAG). (C) KEGG enrichment analysis diagram of ARGs. (D) Chordal graph of KEGG enrichment of ARGs

Table 2: KEGG enrichment of ARGs

Num.	Pathway id	Description	Ratio_in_study	p-value_corrected
69	map00330	Arginine and proline metabolism	69/173	3.5223E-128
51	map00220	Arginine biosynthesis	51/173	7.46576E-97
28	map00410	Beta-alanine metabolism	28/173	8.45645E-35
28	map00250	Alanine, aspartate and glutamate metabolism	28/173	3.10406E-32
16	map00903	Limonene and pinene degradation	16/173	1.8982E-27
17	map00340	Histidine metabolism	17/173	6.53342E-23
19	map00380	Tryptophan metabolism	19/173	2.48574E-20
17	map00310	Lysine degradation	17/173	4.41538E-17

(Continued)

Table 2 (continued)				
Num.	Pathway id	Description	Ratio_in_study	p-value_corrected
17	map00053	Ascorbate and aldarate metabolism	17/173	7.13998E-17
17	map00280	Valine, leucine and isoleucine degradation	17/173	7.36386E-16

3.4 Integrated Metabolomic-Transcriptomic Analysis

Throughout plant evolution, various strategies have evolved to cope with stress, and among them, amino acids play a crucial role as soluble substances in maintaining cellular osmotic pressure, stabilizing cell membranes, and balancing water levels. Protease enzymes, such as trypsin, can impact plant secondary metabolites, especially amino acids, PAs, and other compounds found in *H. undatus*. Our understanding of the effects of trypsin on amino acids in *H. undatus* is still limited. Integrative multi-omics analysis can help uncover potential molecular mechanisms.

Our study revealed that trypsin promotes the production of metabolites related to arginine metabolism in *H. undatus*. Fig. 4 depicts the biosynthesis and metabolism of arginine, demonstrating that trypsin enhances both the biosynthesis and metabolism of arginine. Metabolites associated with arginine metabolism were significantly elevated, including arginine, spermine, agmatine, and putrescine, with spermine showing the most pronounced upregulation. Trypsin increased its abundance by a fold change (FC) of 1.7594. During the biosynthesis of spermine, the upregulation of 4-hydroxy-2-oxoglutaric acid exhibits the most significant change (FC = 2.117). Further VIP value analysis was performed on the 15 metabolites involved in spermine biosynthesis (Fig. S3), and 4-hydroxy-2-oxoglutaric acid had the highest VIP value (VIP = 1.1788), with $p < 0.01$, indicating that 4-hydroxy-2-oxoglutaric acid plays a crucial role as a key metabolite in the regulation of spermine metabolism by pancreatic protease. PAs are important regulators that play a crucial role in plant growth, morphogenesis, delaying senescence, and environmental adaptation. Under trypsin treatment, the PAs content in *H. undatus* pericarp was higher compared to the control group.

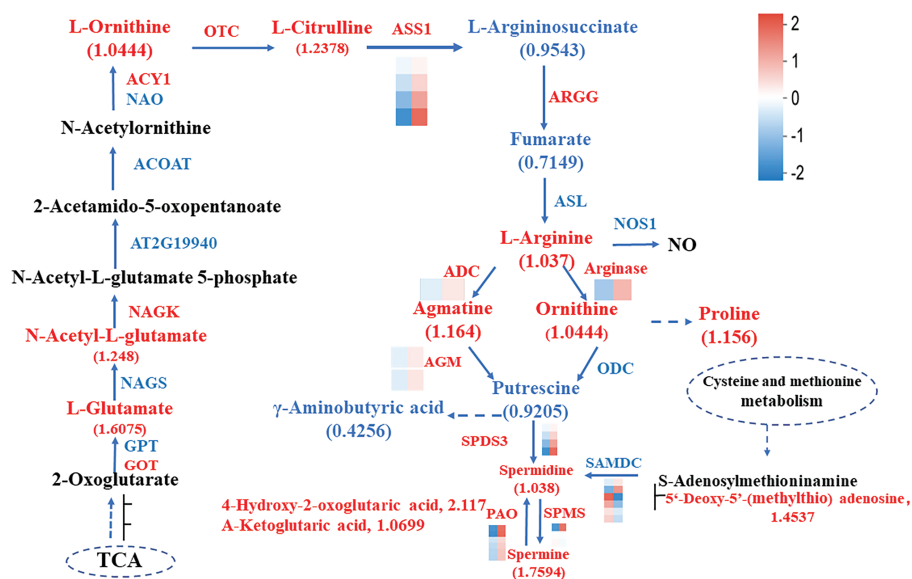


Figure 4: Schematic representation of the arginine biosynthesis pathway as affected by trypsin in *H. undatus*. Note: red represents up-regulation and blue represents down-regulation. Each column of the color legend in the figure represents a sample, from left to right are the control group and the treatment group. For the specific change trend of the expression amount, see the number label under the color bar at the top right

These findings suggested that trypsin promotes arginine metabolism in *H. undatus* pericarp, leading to an increase in PAs content, with spermine being notably affected. The biosynthesis and metabolism of PAs involve complex processes. In the arginine metabolism pathway (Fig. 4), three key enzymes were involved: ADC, NOS, and arginase. In the transcriptomic data, we observed significant upregulation of ADC and arginase compared to the control group, which determined the direction of arginine metabolism. Therefore, it was speculated that during *H. undatus* storage, trypsin regulation mainly promotes arginine metabolism, subsequently stimulating the synthesis of related proteins, thereby achieving a preservation effect.

3.5 Protein-Protein Interaction (PPI) Network Analysis

3.5.1 PPI Network Analysis of Proteins Related to Arginine

Next, protein-protein interaction (PPI) network analysis was employed to investigate the mechanisms of interaction among proteins encoded by ARGs. The analysis aimed to identify key proteins of arginine-polyamine metabolism involved in the regulation of trypsin.

A PPI subnetwork was constructed, consisting of the arginine-related proteins (ARPs) encoded by 187 ARGs. This subnetwork comprised 92 nodes and 799 edges, highlighting 49 upregulated nodes and 43 downregulated nodes, represented by triangles and rounded rectangles, respectively (Fig. 5A).

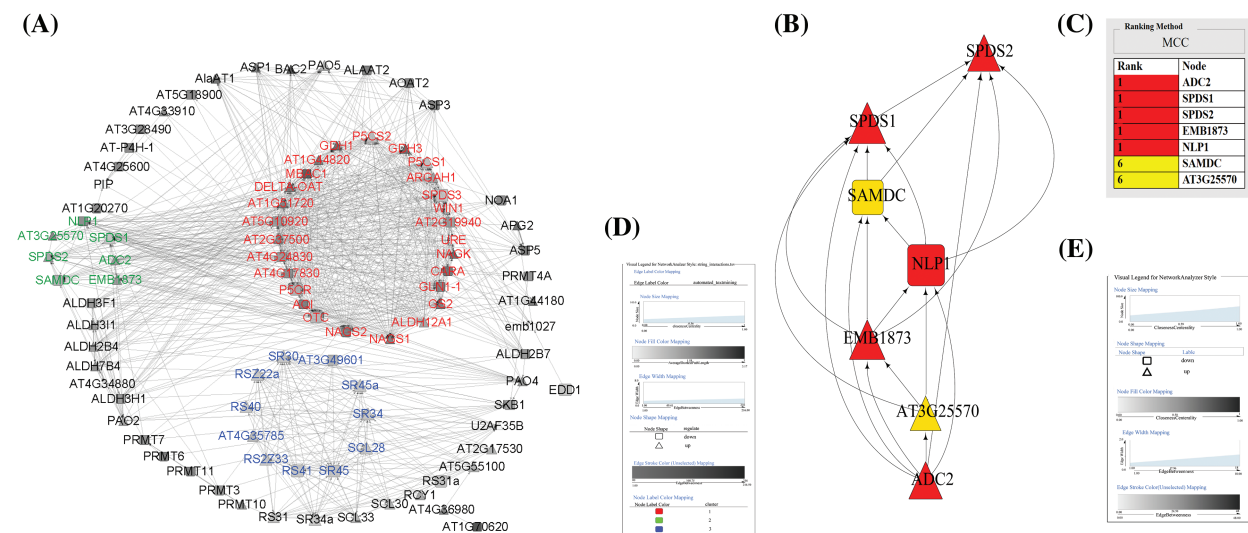


Figure 5: Expression correlation network analysis. (A) PPI networks of trypsin-induced ARPs constructed by Cytoscape. (B) PPI network of differentially expressed genes in cluster 3. (C) Cluster ranking information. (D and E) Legends for the figure. Note: red indicates cluster 1, blue indicates cluster 2, and green indicates cluster 3

3.5.2 Cluster Analysis in the PPI Network of ARGs

Cluster analysis was performed using the Cytoscape plugin “MCODE,” resulting in the identification of 8 clusters within the ARGs’ PPI network. These clusters contained 27, 11, 7, 6, 6, 5, 3, and 3 nodes, respectively. For analysis, we focused on the nodes from the top three protein clusters, which were marked with different colors (Fig. 5A).

Cluster 1 primarily consisted of enzymes related to arginine and proline biosynthesis, such as arginine succinyltransferase (AT4G), ornithine carbamoyltransferase (OTC), pyrroline-5-carboxylate reductase (P5CR), and others. Cluster 2 involved serine/arginine-rich proteins (SR proteins) enriched in serine and arginine residues, which are important splice factors in plants’ response to abiotic stress. Notable proteins

in this cluster included serine/arginine-rich splicing factor (SR45) and serine/arginine-rich SC35-like splicing factor (SCL28). Cluster 3 was associated with enzymes involved in the generation of PAs during arginine metabolism, including SPDS1, ADC, guanidinobutyrase (EMB1873), and others.

The Cytoscape plugin “MCODE” aided in illustrating the affiliation of ARGs and DEGs within the PPI network. It was evident that many key enzymes related to arginine biosynthesis and metabolisms, such as arginine succinate synthase (ASS), arginine succinate lyase (AT5G10920), ADC, SPDS, and SAMDC were clustered in protein clusters 1 and 3.

3.5.3 Core Protein Interaction Network Analysis of ARGs

To delve into the potential regulatory mechanisms of trypsin, it was crucial to identify the central proteins within the ARPs network. Studying the interactions among ARPs can provide insights into their roles in trypsin-induced arginine metabolism.

To explore the hub of ARPs metabolism, the nodes from cluster 3 within the DEG overall network were ranked using the cytoHubba method “Maximum Clique Centrality (MCC).” The results from cluster 3 indicated a two-tier structure (Fig. 5B): ADC2, SPDS1, SPDS2, guanidinobutyrase (EMB1873), and N-acetylputrescine oxidase (NLP) comprised the first tier, while SAMDC and AT3G25570 constituted the second tier. This ordering, based on the cytoHubba plugin, further underscored the central roles of ADC, SPDS, and SAMDC within the ARPs metabolism network.

In plants, the initial reaction of PAs biosynthesis involves arginine and ornithine undergoing decarboxylation by ADC and ornithine decarboxylase (ODC) to form putrescine. Subsequently, putrescine is converted to spermidine by SPDS3, and spermine is synthesized by adding an aminopropyl group derived from S-adenosylmethionine decarboxylation. From the sorting results in Fig. 5, we can observe that ADC and SPDS were the core genes of cluster 3 and are both significantly upregulated.

Based on this, we can speculate that trypsin regulates arginine metabolism and polyamine biosynthesis primarily through core proteins like ADC and SPDS. Many other proteins were controlled by these core proteins, which amplified trypsin’s preservative effect on fruit skin, thus extending the storage period of the fruit.

3.6 Accuracy of the RNA-Seq Data Verification by RT-qPCR

The accuracy of transcriptome data of four key genes was checked by RT-qPCR (Fig. 6A). The square value of R ($R^2 = 0.9389$) was obtained between RT-qPCR and transcriptome spectrum (Fig. 6B). Table S1 showed the RT-qPCR experimental information of four key genes.

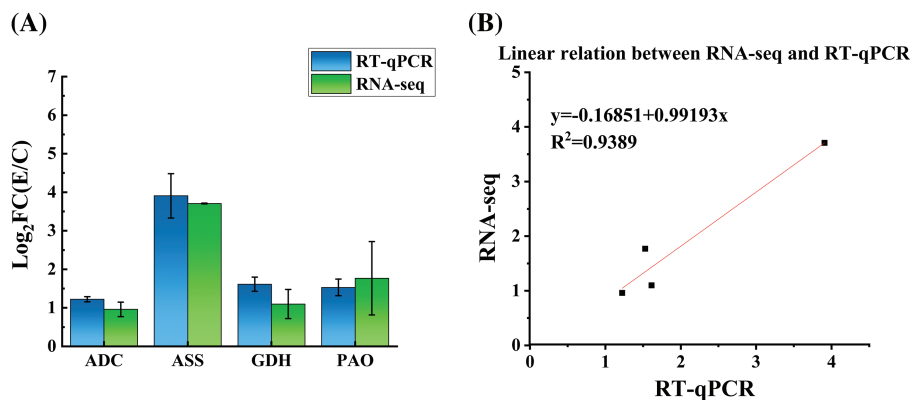


Figure 6: Expression of four genes treated with trypsin in transcriptome data and RT-qPCR. (A) Expression patterns of the selected 4 genes, as determined by transcriptomic and RT-qPCR. (B) Linear relationship of the expression ratio of selected genes measured by transcriptomic and RT-qPCR

4 Discussion

In this study, we investigated the effect of trypsin on the storage quality of *H. undatus* and elucidated its regulatory mechanisms through widely targeted metabolomics and transcriptomics analysis. We found that trypsin treatment can significantly reduce the dehydration rate of *H. undatus*, and maintain the color of the fruit skin, thereby improving the storage quality of the fruit. By integrating metabolomics and transcriptomics data, we revealed the potential mechanisms underlying the enhancement of *H. undatus* preservation through trypsin-induced arginine metabolism and PAs biosynthesis.

Our study indicated that the fruit preservation effect of trypsin may be mainly attributed to the promotion of arginine metabolism and the increase of PAs biosynthesis. The most important plant PAs are putrescine, spermidine, and spermine [7]. Compared with putrescine and spermidine, spermine has more amino groups and higher physiological activity in plants. Based on the results of widely targeted metabolomics studies, we reached a similar conclusion: during trypsin-regulated arginine metabolism, putrescine was downregulated, while spermidine and spermine showed an upregulation trend, with spermine showing a significant increase (FC = 1.7594). The increased activity of spermine inhibits the respiration of the fruit, exerting an antioxidant effect [7], thereby further extending the storage time of *H. undatus*. 4-hydroxy-2-oxoglutaric acid showed the most significant upregulation (FC = 2.117; VIP = 1.1788) in spermine biosynthesis, indicating that 4-hydroxy-2-oxoglutaric acid is the most critical metabolite in trypsin-regulated arginine metabolism.

Through transcriptomic analysis, we have extensively studied the regulatory mechanisms of trypsin treatment on the arginine metabolism pathway. Transcriptomic data revealed that under trypsin treatment, the key gene ADC in the arginine metabolism pathway was upregulated. ADC catalyzes the cleavage of the carboxyl group in arginine molecules, resulting in the production of spermine and carbon dioxide. This further confirmed the important role of arginine metabolism in *H. undatus* preservation. We speculated that trypsin may promote the synthesis of 4-hydroxy-2-oxoglutaric acid by regulating arginine metabolism, thereby facilitating the production of spermine and various other polyamines. Which in turn participates in the regulation of protein synthesis and degradation, maintaining cellular metabolic balance, enhancing plant antioxidant activity, and mitigating damage to further delay fruit senescence.

In the biosynthesis pathway of spermine, 4-hydroxy-2-oxoglutaric acid is one of the precursors which is eventually transformed into spermine via a series of enzyme-catalyzed reactions. Therefore, the accumulation level of 4-hydroxy-2-oxoglutaric acid can indirectly reflect the activity and efficiency of spermine biosynthesis. Researchers found that under drought stress conditions, 4-hydroxy-2-oxoglutaric acid accumulated significantly in tobacco roots by analyzing the metabolite composition [27]. This result indicated that the metabolic state of plants changes under drought stress, and the accumulation of 4-hydroxy-2-oxoglutaric acid may be an adaptive strategy to cope with the metabolic and physiological changes caused by drought stress. It has also been suggested that 4-hydroxy-2-oxoglutaric acid may serve as a useful marker for drought stress in Solanaceae plants [28]. However, the role of 4-hydroxy-2-oxoglutaric acid in post-harvest fruit senescence is still poorly understood and further research is needed to elucidate its mode of action and interaction with other metabolic pathways during fruit senescence. Our study provides new insights into the role of trypsin-regulated spermine metabolism during the storage of *H. undatus* and contributes to a better understanding of the multifunctionality of 4-hydroxy-2-oxoglutaric acid in different biological processes.

In summary, our study revealed the potential mechanism by which trypsin promotes the preservation of *H. undatus*. We speculate that 4-hydroxy-2-oxoglutaric acid regulates the biosynthesis of polyamines, including spermine, enhancing the plant's adaptability to abiotic stress. Spermine plays a role in abiotic stress through various mechanisms, such as regulating stomatal closure, alleviating oxidative damage, and maintaining cell osmotic balance [29,30]. This provides a new perspective and theoretical support for

research in the field of fruit preservation. However, further research is needed to elucidate the molecular details of trypsin regulation and its application prospects under different plant and environmental conditions.

5 Conclusion

Among the 453 compounds analyzed, a total of 75 differential metabolites related to amino acids were identified. The arginine metabolism pathway emerged as the primary route through which trypsin regulates amino acids, contributing to the quality improvement of *H. undatus* during storage. Transcriptional analysis identified 186 ARGs. The most prominently regulated amino acid-related metabolite was 4-hydroxy-2-oxoglutaric acid. The “arginine biosynthesis and metabolism pathway” appeared to be the central route through which trypsin regulates amino acids, as demonstrated by the results from the PPI network protein clusters, including ADC, SPDS, and SAMDC, which served as pivotal proteins in trypsin-regulated arginine metabolism. This observation aligned well with the findings from KEGG enrichment chord diagrams.

As a novel bio-preservative, trypsin promotes arginine metabolism, and increases the content of 4-hydroxy-2-oxoglutaric acid, thereby regulating the biosynthesis of polyamines including spermine, and extending the shelf life of fruits. The results of this study provided valuable theoretical support for the post-harvest preservation of *H. undatus*.

Acknowledgement: We are grateful for the free online platform of Majorbio I-Sanger Cloud Platform (www.majorbio.com).

Funding Statement: This work was supported by the National Key Research and Development Program of China (2017YFC1600802).

Author Contributions: The authors confirm contribution to the paper as follows: study conception and design: Bairu Li, Xin Li; data collection: Jingyu Jia; analysis and interpretation of results: Hemin Wang, Jiaju Sun, Enyan Chen; draft manuscript preparation: Bairu Li, Xin Li. All authors reviewed the results and approved the final version of the manuscript.

Availability of Data and Materials: All data and materials used in this research were publicly available. Raw sequence data from this study have been submitted to the NCBI sequence read archive under the BioProject Accession (PRJNA 509494).

Ethics Approval: Not applicable.

Conflicts of Interest: The authors declare that they have no conflicts of interest to report regarding the present study.

Supplementary Materials: The supplementary material is available online at <https://doi.org/10.32604/phyton.2024.050450>.

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