



Screening the Candidate Postmortem Marker Genes for Estimating Postmortem Intervals in Nurse Bees (*Apis mellifera*)

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Abstract

The honey bee (*Apis mellifera*) plays an essential role in global agriculture; however, colony losses have been frequently reported. Because several mRNAs remain active for a limited period after death, quantitative real-time PCR (qRT-PCR) analysis of mRNA degradation may be useful for estimating postmortem intervals (PMIs) in dead bees. In this study, the expression dynamics of thirteen candidate postmortem marker genes were analyzed in nurse bee carcasses from 1 to 7 days postmortem. Among them, *RPS5* and *RPS18* showed significant decreases in expression within the early postmortem phase (1–3 days). Based on cubic regression models, *RPS5* ($R^2 = 0.997$) and *RPS18* ($R^2 = 0.982$) demonstrated strong correlations with PMI. Furthermore, expression ratios using stably expressed genes as denominators revealed that *RPS5/SDH* and *RPS18/SDH* achieved the highest coefficients of determination ($R^2 = 0.996$), distinguishing carcasses between 1–2 and 3–7 days postmortem. These findings suggest that *RPS5*, *RPS18*, and their ratios with *SDH* can serve as molecular markers for PMI estimation in nurse bees. This molecular approach may support forensic analyses of honey bee mortality under both laboratory and field conditions.

Keywords

Apis mellifera, Postmortem interval, mRNA degradation, *RPS5*, *RPS18*, *SDH*

INTRODUCTION

The western honey bee (*Apis mellifera*) is one of the most important pollinators contributing to agricultural productivity and global ecosystem stability. The Inter-governmental Science-Policy Platform on Biodiversity and Ecosystem Services (IPBES) reported that approximately 35% of global crop production depends on pollination services, with an estimated economic value of 235–577 billion USD (Jung, 2008; IPBES, 2016). In Korea, honey bees are responsible for pollinating nearly 50% of fruit and vegetable crops, providing pollination services valued at approximately 4 billion USD (Jung, 2008). Similarly, Reilly *et al.* (2020) estimated that honey bee-mediated pollination contributes about 6.4 billion

USD annually to U.S. crop yields, with almonds alone accounting for 4.2 billion USD.

However, these contributions have declined in parallel with the global reduction in honey bee populations (Sumner and Boriss, 2006; Meixner, 2010). Multiple factors—including exposure to pesticides and miticides, parasitic infestation, pathogens, and abnormal climatic conditions—have been identified as major stressors that threaten colony health and contribute to colony collapse (Meixner, 2010; Kim, 2022; Lee *et al.*, 2022). Consequently, numerous studies have focused on the physiological effects of these stressors on surviving bees to identify potential lethal mechanisms (Zaobidna *et al.*, 2017; Dolezal and Toth, 2018; Butolo *et al.*, 2021; You *et al.*, 2025).

Nevertheless, beekeepers often recognize colony losses only after large-scale mortality events occur, at which point most individuals are already dead. Thus, analyzing dead bees rather than survivors may provide more practical insight into the underlying causes of colony collapse. Following death, cellular metabolism ceases and autolytic degradation begins (Brooks and Sutton, 2017; Wenzlow *et al.*, 2023). However, several studies have demonstrated that certain messenger RNAs (mRNAs) remain detectable for a limited period after death (Kimura *et al.*, 2011; Pozhitkov *et al.*, 2017; Shafeeq *et al.*, 2020) reported that stress- and development-related genes continued to be expressed for 48–96 hours in post-mortem mouse and zebrafish tissues, whereas Kimura *et al.* (2011) observed that circadian genes in mice maintained rhythmic expression patterns for up to 48 hours after death. In insects, Shafeeq *et al.* (2020) found that *hsp70* expression increased for five days postmortem in *Plodia interpunctella*, while *ultraspiracle* transcripts remained stable.

Because the persistence of mRNA varies depending on the postmortem interval (PMI), these molecular changes have been proposed as biomarkers for estimating PMI (Scrivano *et al.*, 2019; Cianci *et al.*, 2024). Several analytical methods—such as the ΔC_q method, C_q ratio, and direct quantification of transcript abundance—have been employed to model the relationship between mRNA degradation and PMI (Kimura *et al.*, 2011; Ali *et al.*, 2017; Alshehhi and Haddrill, 2019; Wang *et al.*, 2021). However, mRNA stability is affected by multiple variables, including tissue type, temperature, humidity, and time since death (Bauer, 2007; Sampaio-Silva *et al.*, 2013). Moreover, the suitability of specific postmortem marker genes differs among species, underscoring the need to identify honey bee-specific markers for PMI estimation.

In this context, the present study aimed to identify reliable molecular markers that can be used to estimate PMI in nurse bee (*A. mellifera*) carcasses. To this end, the expression levels of ten housekeeping genes [*RPS5* (40S ribosomal protein S5), *RPS18* (40S ribosomal protein S18), *GAPDH* (Glyceraldehyde-3-phosphate dehydrogenase), *ARF1* (ADP-ribosylation factor 1), *PPI* (Peptidyl-prolyl cis-trans isomerase), *PGK* (Phosphoglycerate kinase), *SDH* (Succinate dehydrogenase flavoprotein subunit), *TBP* (TATA-box-binding protein), *EF1* (Elongation

factor 1-alpha F2), and *EcR* (Ecdysone receptor)] and four putative postmortem genes—*LOC410857* (protein lethal (2) essential for life), *DAAM* (disheveled-associated activator of morphogenesis-like protein), and *Prx2* (Peroxiredoxin 1)—were examined using quantitative real-time PCR (qRT-PCR). Expression ratios between candidate genes and stable reference genes were further analyzed to enhance resolution across PMI time points. Mathematical regression models were then established to determine the most suitable markers and to propose an accurate method for estimating PMI in honey bee carcasses.

MATERIALS AND METHODS

1. Sample preparation

Colonies of the western honey bee (*A. mellifera*) were maintained at the experimental apiary in Sangju-si, Korea. Nurse bees were collected based on their labor-specific behaviors (Johnson, 2010) and stabilized in cages (Bugdorm-1, MegaView Science Co., Ltd., Taichung City, Taiwan) supplied with 50% sucrose solution. The bees were maintained for 24 h in a dark incubator at 32°C (JS Research Inc.) prior to experiments. To obtain carcasses for different postmortem intervals (PMIs), the nurse bees were euthanized with CO₂ and maintained under controlled laboratory conditions. Carcasses were sampled at 1, 2, 3, 4, 5, 6, and 7 days postmortem, ensuring no fungal growth on the body surface. Samples were immediately stored at –80°C until RNA extraction.

2. Candidate postmortem marker genes and primer design

Thirteen candidate postmortem marker genes were selected to evaluate their potential for PMI estimation: *RPS5*, *RPS18*, *GAPDH*, *ARF1*, *PPI*, *PGK*, *SDH*, *TBP*, *EF1*, *LOC410857*, *DAAM*, *Prx2*, and *EcR* gene was used as a reference gene based on its previously validated expression stability under postmortem conditions (Kim and Kim, 2025a). Primer sequences for ten genes (*RPS5*, *RPS18*, *GAPDH*, *ARF1*, *RAB1a*, *PPI*, *PGK*, *SDH*, *TBP*, and *EF1*) were obtained from previous reports (Moon *et al.*, 2018; Jeon *et al.*, 2020; Kim *et al.*, 2022; Kim and Kim, 2025a, 2025b). Primers for *LOC410857*, *DAAM*,

Table 1. Information of primer sets for candidate PMI marker gene and reference gene

Gene		Primer							Reference
Symbol (Gene name)	Accession No.	Label	Sequence (5'→3')	Size (bp)	GC (%)	Tm (°C)			
^a <i>RPS5 (40S ribosomal protein S5)</i>	XM_006570237	F	GATGTTTCTCCGTTACGACGAGT	23	48	62.9		Jeon <i>et al.</i> (2020)	
		R	GAGTTCAATCGGCTAAACATTCGG	23	48	62.9			
^a <i>RPS18 (40S ribosomal protein S18)</i>	XM_625101	F	GATTCGCCGATTGGTTTTGAATAG	24	38	60.3		Moon <i>et al.</i> (2018)	
		R	AACCCCAATAATGACGCAAAACC	22	45	60.1			
^a <i>GAPDH (Glyceraldehyde-3-phosphate dehydrogenase)</i>	XM_393605	F	CACCTTCTGCAAAATTAATGGCG	22	45	60.1		Moon <i>et al.</i> (2018)	
		R	ACCTTTGCCAAGTCTAACTGTAA	24	38	60.3			
^a <i>ARF1 (ADP-ribosylation factor 1)</i>	LOC409481	F	GGGCTTCATTTCTCCGCAA	20	55	60.5		Kim <i>et al.</i> (2022)	
		R	AGAGCCAATCAAGACCCTCG	20	55	60.5			
^a <i>RAB1a (Ras-related protein Rab-1A)</i>	LOC102654987	F	CTTAGAGTGGTCTCTCCATC	20	55	60.5		Kim <i>et al.</i> (2022)	
		R	CAGCAGCATCCAGATTTAGAGG	22	50	62.1			
^a <i>PPI (Peptidyl-prolyl cis-trans isomerase)</i>	XM_393381	F	GCAGCTTCCATAGAGTGTACC	21	52	61.2		Kim and Kim (2025b)	
		R	TTGGACCAGCATTAGCCATGG	21	52	61.2			
^a <i>PGK (Phosphoglycerate kinase)</i>	XM_395047	F	GACGAGGAAGGAGCAAAAATCA	22	45	60.1		Kim and Kim (2025b)	
		R	CCCATCCAGCCATCAGGTAT	20	55	60.5			
^a <i>SDH (Succinate dehydrogenase flavoprotein subunit)</i>	XM_623062	F	ATGATCAGGATA CAGTTGTGCC	22	45	60.1		Kim and Kim (2025b)	
		R	TTAGAGGGCCAATAGTTTCTCC	22	45	60.1			
^a <i>TBP (TATA-box-binding protein)</i>	XM_623085	F	CAAAATATGGTAGGCA GCTGTG	22	45	60.1		Kim and Kim (2025b)	
		R	CGAATCTAGGTTTAAACCAATACGAT	24	38	60.3			
^a <i>EF1 (Elongation factor 1-alpha F2)</i>	NM_001014993	F	GTCGTGGTTATGTTGCTGGTGAT	23	48	62.9		Jeon <i>et al.</i> (2020)	
		R	CGCAATTTCTCTTTTGATATCAGCGAA	25	40	62.5			
<i>LOC410857 (protein lethal(2)essential for life)</i>	XM_006568175.3	F	CGGTGAAGACGGTCCGGTAAT	20	55	60.5		This study	
		R	GTCGTGAGATGGAGGCAAGA	20	55	60.5			
<i>DAAM (disheveled-associated activator of morphogenesis-like protein)</i>	XM_006561849.3	F	GGCTCAGAAAGCTCTCGAA	20	55	60.5		This study	
		R	GCATGTCCTCTCGATATCC	20	55	60.5			
<i>Prx2 (Peroxi redoxin 1)</i>	XM_003249241.4	F	CCTTTTCTGATCGTGTGATGA	22	45	60.1		This study	
		R	TCAAGTACGCCATAATCACGTG	22	45	60.1			
^a <i>EcR (ecdysone receptor)</i>	NM_001098215.2	F	CGGTATACGGAGCGATCAGA	20	55	60.5		Kim and Kim (2025a)	
		R	GTTGCTCGTACTCGTTCTGG	20	55	60.5			

^aInformation on primers was obtained from previous studies.

Prx2, and *EcR* were designed using the OligoCalc software (<http://biotools.nubic.northwestern.edu/OligoCalc.html>). PCR amplification efficiency was determined from standard curves derived from 10-fold serial dilutions of cDNA, and calculated using the formula $E = 10^{-1/\text{slope}}$. All primer sets exhibited efficiencies between 92–107% with coefficients of determination (R^2) > 0.99, indicating acceptable performance for qRT-PCR analysis (Table 1).

3. RNA extraction, cDNA synthesis, qRT-PCR

For each sample, the antennae were removed to prevent contamination from damaged tissue. Total RNA was extracted from individual heads using the yesR Total Plus Kit (GenesGen, Busan, Korea) according to the manufacturer's protocol. Genomic DNA was eliminated using gDNA Eliminator Buffer and DNase I (GenesGen). The concentration and purity of RNA were measured using a SpectraMax QuickDrop spectrophotometer (Molecular Devices, Sunnyvale, CA, USA). RNA samples were stored at -80°C until use.

Complementary DNA (cDNA) was synthesized from 50 ng of total RNA using ReverTra AceTM qPCR RT Master Mix (Toyobo, Japan) following the manufacturer's protocol. The thermal cycling conditions were 37°C for 15 min, 50°C for 5 min, and 98°C for 5 min.

qRT-PCR was performed on a CFX Connect Real-Time PCR Detection System (Bio-Rad, Hercules, CA, USA) using Thunderbird SYBR qPCR Master Mix (Toyobo, Osaka, Japan). The reaction conditions were as follows: initial denaturation at 95°C for 1 min, followed by 40 cycles of 95°C for 15 s, 56°C for 15 s, and 72°C for 30 s. Each reaction was run in triplicate. Amplification specificity was confirmed by melting curve analysis. C_q values were obtained automatically using the CFX Manager software, and relative expression levels were normalized to *RAB1a* using the $2^{-\Delta\Delta C_q}$ method.

4. Statistical analysis and mathematical modeling

Normalized expression levels of each candidate gene across different PMIs were analyzed using one-way analysis of variance (ANOVA) followed by Tukey's multiple comparison test ($p < 0.05$). Expression ratios were calculated using the following formula:

$$\text{Ratio} = \frac{\text{expression level of target gene}}{\text{expression level of stable gene}}$$

Stable genes used as denominators were selected based on their expression stability rankings, determined by the smallest average standard deviation of expression levels across all PMIs (Fig. 1) and supported by a previous study (Kim and Kim, 2025a).

Cubic polynomial regression models were constructed to evaluate the correlation between gene expression (or expression ratios) and PMIs. The model with the highest coefficient of determination (R^2) was considered the most accurate for PMI estimation.

RESULTS

1. Expression patterns of candidate postmortem marker genes

The expression levels of thirteen candidate genes were quantified in nurse bee heads at seven postmortem intervals (1–7 days) using qRT-PCR (Fig. 1). Among them, five genes—*RPS5*, *RPS18*, *GAPDH*, *PPI*, and *EF1*—exhibited significant temporal changes after death ($p < 0.05$), whereas the remaining eight genes (*ARF1*, *PGK*, *SDH*, *TBP*, *LOC410857*, *DAAM*, *Prx2*, and *EcR*) showed no significant differences across PMIs ($p > 0.05$).

Expression of *RPS5* decreased markedly during the first 3 days postmortem, with statistically distinct levels at 1, 2, and 3 days ($p < 0.05$). Thereafter, expression remained stable from 3 to 7 days, suggesting that *RPS5* is a sensitive marker for distinguishing early (≤ 3 days) and late (≥ 3 days) postmortem phases (Fig. 1A). Similarly, *RPS18* expression gradually declined from 1 to 4 days postmortem, with a clear separation between early (1–2 days) and mid-to-late (3–7 days) PMIs ($p < 0.05$; Fig. 1B). In contrast, *GAPDH*, *PPI*, and *EF1* showed significant downregulation only at 1 day postmortem ($p < 0.05$), indicating limited utility for distinguishing longer PMIs (Fig. 1C–E).

Cubic regression analysis revealed strong correlations between PMI and the expression levels of *RPS5* ($R^2 = 0.997$) and *RPS18* ($R^2 = 0.982$), while other genes showed weaker associations ($R^2 < 0.90$; Table 2). These results suggest that *RPS5* and *RPS18* can serve as quantitative

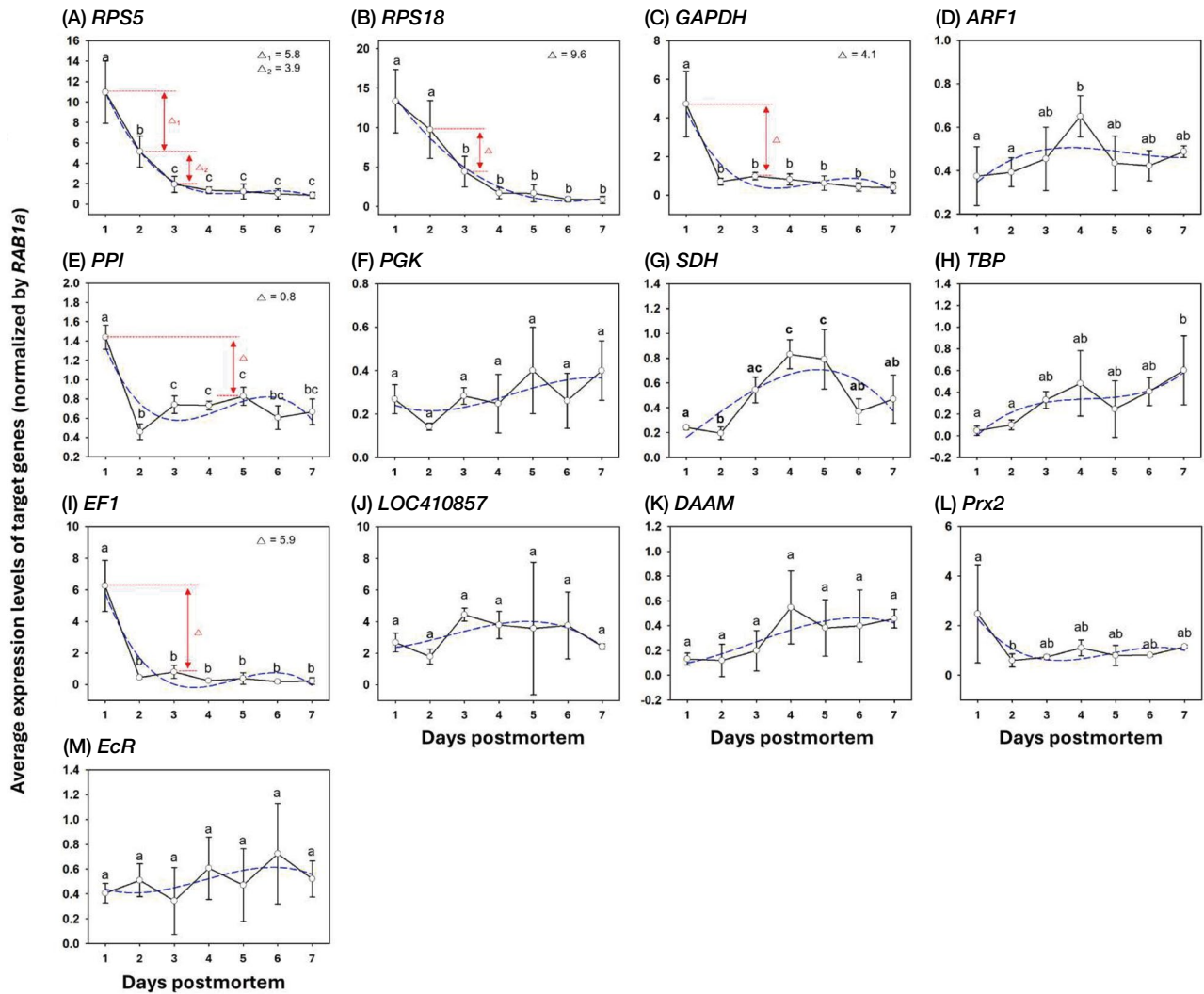


Fig. 1. Expression patterns of postmortem marker genes under postmortem condition. The expression levels (average \pm standard deviation) of target genes were measured depending on PMI. The difference value (Δ) represented the smallest gap between expression levels in PMI estimation ranges. The Different letters indicate significantly different values ($p < 0.05$) that analyzing by One-way ANOVA with Turkey's multiple comparison test.

indicators for estimating postmortem duration within 3 days after death.

2. Expression ratios of candidate genes

To improve PMI resolution, expression ratios were calculated between the five PMI-responsive genes (*RPS5*, *RPS18*, *GAPDH*, *PPI*, *EF1*) and the four stably expressed genes (*ARF1*, *PPI*, *PGK*, *SDH*). Among all combinations, ratios using *SDH* as the denominator exhibited the clearest differentiation among PMIs (Fig. 2). The *RPS5/SDH* and *RPS18/SDH* ratios significantly distinguished 1-, 2-, and 3–7-day postmortem groups ($p < 0.05$), while

GAPDH/SDH and *EF1/SDH* only separated 1-day samples from later time points. Ratios involving other denominators (e.g., *RPS5/PPI*, *RPS18/PGK*) showed partial separation between early (1–2 days) and late (3–7 days) intervals but exhibited lower model accuracy ($R^2 < 0.9$).

In contrast, the cubic regression models for *RPS5/SDH* and *RPS18/SDH* achieved high coefficients of determination ($R^2 = 0.996$ for both; Table 2), indicating strong predictive relationships between expression ratios and PMIs.

The mean difference values between PMI groups were also greatest for these two ratios: 27.3 between 1 and 2

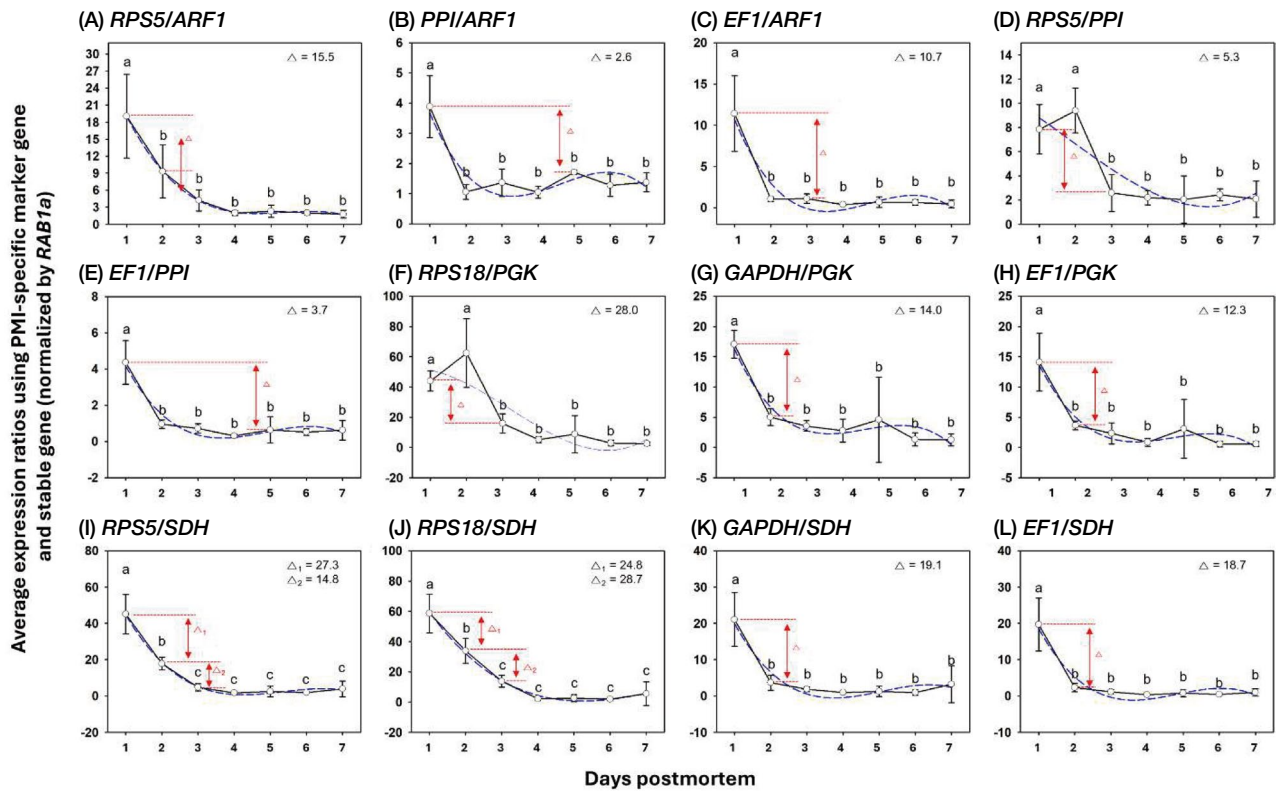


Fig. 2. The ratios of expression levels using PMI-specific marker genes and stable gene under postmortem condition. The expression ratios (average \pm standard deviation) showing distinct PMI estimated ranges were calculated using PMI-specific gene and stably expressed gene (*ARF1*, *PPI*, *PGK*, and *SDH*). The difference value (Δ) represented the smallest gap between expression levels in PMI estimation ranges. The Different letters indicate significantly different values ($p < 0.05$) that analyzing by One-way ANOVA with Turkey's multiple comparison test.

days for *RPS5/SDH* and 53.5 between 1 and 3–7 days for *RPS18/SDH*. These results demonstrate that both ratios can reliably distinguish among multiple postmortem stages and are suitable molecular markers for PMI estimation in *A. mellifera*.

3. Summary of optimal regression models

Cubic polynomial models were generated to describe the relationship between gene expression (or ratio) and PMI (Table 2). The models for *RPS5*, *RPS18*, *RPS5/SDH*, and *RPS18/SDH* exhibited $R^2 > 0.98$, demonstrating high accuracy for PMI prediction. The regression equations are summarized in Table 2. These models indicate that PMI estimation based on expression ratios provides slightly higher precision than estimation based on single-gene expression levels.

Overall, the results demonstrate that (1) *RPS5* and *RPS18* show predictable degradation patterns after death; (2) *ARF1*, *PPI*, *PGK*, and *SDH* are stable postmortem

reference genes; and (3) the ratios *RPS5/SDH* and *RPS18/SDH* provide the highest accuracy for PMI estimation in nurse bee carcasses, effectively distinguishing carcasses within 1–2 days postmortem from those after 3 days.

DISCUSSION

The present study aimed to identify candidate post-mortem marker genes that can be used to estimate post-mortem intervals (PMIs) in nurse bee (*A. mellifera*) carcasses. Estimating PMI in dead bees is crucial for understanding the causes of colony losses, since beekeepers usually detect mortality events after colony collapse has already occurred. Previous studies have primarily focused on the physiological effects of stressors on living bees (Zaobidna *et al.*, 2017; Dolezal and Toth, 2018; Butolo *et al.*, 2021; You *et al.*, 2025), but few have explored molecular indicators in carcasses.

Table 2. The regression equations for candidate PMI markers (single gene or combinations of PMI-specific gene and stable gene) showing distinct PMI estimation ranges

Candidate	The equation of regressive curve	R ²
<i>RPS5</i>	$y = -0.145x^3 + 2.271x^2 - 11.566x + 20.366$	0.997
<i>RPS18</i>	$y = -0.026x^3 + 0.862x^2 - 7.512x + 20.43$	0.982
<i>GAPDH</i>	$y = -0.103x^3 + 1.439x^2 - 6.369x + 9.415$	0.889
<i>PPI</i>	$y = -0.028x^3 + 0.37x^2 - 1.491x + 2.477$	0.686
<i>EF1</i>	$y = -0.148x^3 + 2.112x^2 - 9.413x + 13.258$	0.903
<i>RPS5/ARF1</i>	$y = -0.224x^3 + 3.601x^2 - 18.935x + 34.6$	0.999
<i>PPI/ARF1</i>	$y = -0.085x^3 + 1.177x^2 - 4.968x + 7.549$	0.862
<i>EF1/ARF1</i>	$y = -0.281x^3 + 4.001x^2 - 17.751x + 24.744$	0.928
<i>Prx2/ARF1</i>	$y = -0.199x^3 + 2.837x^2 - 12.311x + 17.396$	0.877
<i>RPS5/PPI</i>	$y = 0.048x^3 - 0.258x^2 - 1.721x + 10.729$	0.765
<i>EF1/PPI</i>	$y = -0.09x^3 + 1.31x^2 - 5.983x + 8.935$	0.952
<i>RPS18/PGK</i>	$y = 0.699x^3 - 6.752x^2 + 6.633x + 50.478$	0.771
<i>GAPDH/PGK</i>	$y = -0.367x^3 + 5.072x^2 - 22.312x + 33.992$	0.941
<i>EF1/PGK</i>	$y = -0.311x^3 + 4.373x^2 - 19.522x + 29.06$	0.948
<i>Prx2/PGK</i>	$y = -0.366x^3 + 4.889x^2 - 20.16x + 28.343$	0.875
<i>RPS5/SDH</i>	$y = -0.64x^3 + 10.251x^2 - 52.467x + 87.582$	0.996
<i>RPS18/SDH</i>	$y = -0.28x^3 + 6.494x^2 - 44.946x + 98.156$	0.996
<i>GAPDH/SDH</i>	$y = -0.403x^3 + 6.118x^2 - 28.932x + 43.167$	0.944
<i>EF1/SDH</i>	$y = -0.458x^3 + 6.641x^2 - 30.094x + 42.525$	0.944
<i>Prx2/SDH</i>	$y = -0.461x^3 + 6.553x^2 - 28.707x + 39.572$	0.927

In this study, the expression levels of thirteen candidate genes were analyzed across seven postmortem intervals. Among them, *RPS5* and *RPS18* showed clear, time-dependent degradation patterns, allowing the differentiation of carcasses within the first three days after death. The cubic regression models of these genes exhibited strong correlation with PMI ($R^2=0.997$ and 0.982 , respectively), confirming their reliability as postmortem molecular markers. In contrast, other genes (*GAPDH*, *PPI*, and *EF1*) showed significant decreases only at 1 day postmortem, indicating limited applicability for PMI estimation beyond the early phase.

Interestingly, both *RPS5* and *RPS18* have been used as reference genes under various experimental conditions, including postmortem tissues in other species (Sampaio-Silva *et al.*, 2013; Lv *et al.*, 2017; Wang *et al.*, 2022). However, the present study revealed that their expression levels were not stable after death in honey bee tissues, as confirmed by BestKeeper and geNorm analyses in our previous study (Kim and Kim, 2025a). This instability

suggests that *RPS5* and *RPS18* are unsuitable as reference genes but rather are potential postmortem markers whose degradation correlates strongly with time since death. Such discrepancies with previous studies likely reflect interspecific differences in tissue physiology and mRNA stability, as well as methodological variations in sample storage and qRT-PCR conditions.

To improve PMI estimation accuracy, this study also analyzed expression ratios between dynamic and stable genes. According to the previous study (Kim and Kim, 2025a), four genes (*ARF1*, *PPI*, *PGK*, and *SDH*) used as denominator were ranked among the top four in Ref Finder, confirming their high expression stability under postmortem conditions. In this study, these genes also showed low standard deviations of expression levels across all PMIs, supporting their suitability as denominators for calculating expression ratios in PMI estimation. Among all tested combinations, the *RPS5/SDH* and *RPS18/SDH* ratios showed the highest discrimination power among postmortem groups and exhibited strong

polynomial relationships with PMI ($R^2=0.996$ for both). These ratios reliably distinguished carcasses at 1, 2, and ≥ 3 days postmortem, extending the estimation range and improving accuracy compared with single-gene models. This finding supports previous research in which the use of gene expression ratios was shown to reduce variability and improve model robustness (Kimura *et al.*, 2011; Cianci *et al.*, 2024).

The superior performance of *RPS5/SDH* and *RPS18/SDH* may be attributed to two factors: (1) the high stability of *SDH* expression postmortem, and (2) the consistent degradation kinetics of *RPS5* and *RPS18*. Since ribosomal protein genes are highly expressed in active cells and rapidly degrade after cell death, their relative decrease provides a sensitive molecular signature of tissue decay. This suggests that ribosomal gene degradation can serve as a universal postmortem indicator across species.

Although the regression models developed in this study showed high accuracy under controlled laboratory conditions, several factors may influence their performance in the field. Temperature, humidity, and microbial activity can alter mRNA degradation rates (Bauer, 2007; Sampaio-Silva *et al.*, 2013), potentially affecting PMI estimates. Therefore, field validation studies incorporating environmental variables are required to establish practical thresholds for PMI estimation in natural conditions.

Overall, this study provides the first molecular approach to PMI estimation in honey bees using gene expression dynamics. The findings demonstrate that *RPS5* and *RPS18*, along with their ratios to *SDH*, can serve as reliable molecular markers to differentiate early and late postmortem phases in *A. mellifera* carcasses. This molecular framework may contribute to future forensic analyses of colony mortality events and improve diagnostic accuracy in apicultural investigations.

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