

Article

Uncovering Spatio-Temporal and Treatment-Derived Differences in the Molecular Physiology of a Model Coral-Dinoflagellate Mutualism with Multivariate Statistical Approaches

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Abstract: Multivariate statistical approaches (MSA), such as principal components analysis and multidimensional scaling, seek to uncover meaningful patterns within datasets by considering multiple response variables in a concerted fashion. Although these techniques are readily used by ecologists to visualize and explain differences between study sites, they could theoretically be employed to differentiate organisms within an experimental framework while simultaneously identifying response variables that drive documented experimental differences. Therefore, MSA were used herein to attempt to understand the response of the common, Indo-Pacific reef coral *Seriatopora hystrix* to temperature changes using data from laboratory-based temperature challenge studies performed in Southern Taiwan. Gene expression and physiological data partitioned experimental specimens by time of sampling, treatment temperature, and site of origin upon employing MSA, signifying that *S. hystrix* and its dinoflagellate endosymbionts display physiological and molecular signatures that are characteristic of sampling time, site of colony origin, and/or temperature regime. These findings promote the utility of MSA for documenting biologically meaningful shifts in the physiological and/or sub-cellular response of marine invertebrates exposed to environmental change.

Keywords: acclimation; coral reefs; endosymbiosis; molecular biology; multivariate statistics; temperature; upwelling

1. Introduction

In recent years, a concerted effort has been made to better understand the basic biology of reef-building corals [1–4], as well as their response to changing environments [5–7]; the latter topic is especially pertinent given the extent of the anthropogenic pressures currently facing the high biodiversity ecosystems constructed by these cnidarian-dinoflagellate (genus *Symbiodinium*) endosymbioses [8–9]. The impact of changing environments on corals is undoubtedly complex, and many species have been shown to acclimate to extreme abiotic regimes previously hypothesized to compromise the integrity of these calcium carbonate-accreting mutualisms. As an example, although most scleractinian coral-*Symbiodinium* associations readily dissociate when exposed to even small changes in their aquatic milieu (particularly with respect to temperature [10–11]), those of Southern Taiwan have proven to be markedly resilient to an array of laboratory-simulated environmental challenges (Table 1).

For instance, the common, Indo-Pacific reef-builder *Seriatopora hystrix* showed no mRNA-level molecular chaperone response when exposed for 2 d to 30 °C [12], a temperature hypothesized to ultimately elicit bleaching in this species based on observations made in Japan and elsewhere [13]. In fact, the expression of only 2 genes out of the 14 targeted (6 from *Symbiodinium* and 8 from the coral host), the cytoskeleton genes β -actin (*actb*) and α -tubulin (*tuba*), were determined by real-time PCR (qPCR) to be affected by temperature [14]. Mayfield et al. [12] hypothesized that such a lack of a molecular chaperone response, in particular, in either compartment of this holobiont (“host+

endosymbiont”) may have been due to mRNA “front-loading” (*sensu* [15]) in samples of this “*S. hystrix* short-term temperature experiment” (SHSTTE). Briefly, corals of Southern Taiwan inhabit environments characterized by episodic upwelling, whereby temperatures may change by up to 9 °C in a matter of several hours [16]. Therefore, they could be predicted to exhibit high expression levels of heat shock proteins (HSPs) and other stress genes and proteins even during ambient conditions in order to have the molecular machinery requisite for a temperature change-induced stress response at the time temperatures begin to fluctuate due to upwelling.

As an unexplored counter-hypothesis, it is plausible that concerted, biologically meaningful changes in expression of multiple gene mRNAs and other molecular physiological response variables were simply overlooked due to having used univariate statistical approaches only. Multivariate statistical approaches (MSA), such as principal components analysis (PCA) and multidimensional scaling (MDS), can uncover treatment-derived and spatio-temporal differences *not* revealed by univariate statistics-based approaches employing standard ANOVA models by instead looking at the relationships or correlations between various combinations of response variables simultaneously. Specifically, MSA can differentiate samples and treatments by integrating data across multiple parameters and so can partition samples within an experimental dataspace in a holistic manner. Another advantage of MSA, such as multivariate ANOVA (MANOVA), is that such techniques are more statistically conservative when analyzing datasets featuring a large number of response variables; by assessing all parameters (e.g., 17 in the SHSTTE) in an integrated, single-step model, the chances of making a type I error are substantially reduced.

Given these merits, MSA were used to ascertain whether the *S. hystrix-Symbiodinium* holobiont was truly unresponsive to a short-term exposure to a temperature treatment hypothesized to elicit stress (*sensu* [17]). As a comparison, the dataset of the “*S. hystrix* variable temperature study” (SHVTS), which was also conducted at Taiwan’s National Museum of Marine Biology and Aquarium (NMMBA), was re-explored, as corals of this study showed clear physiological and gene expression differences across both temperature regimes (stable vs. variable) and site of origin [18]. Regarding the latter factor, unlike the SHSTTE, in which all corals were from an upwelling site within Nanwan Bay (Taiwan’s southernmost embayment), Houbihu, half of those corals of the SHVTS were from a non-upwelling site, Houwan, which abuts NMMBA and is characterized by low coral cover and poor water quality due to coastal agricultural runoff [19]. It was predicted that MSA could be used to conclusively demonstrate the lack of a gene expression effect on high temperature samples of the SHSTTE and, similarly, further verify both site of origin and temperature treatment differences in the molecular physiology of samples of the SHVTS. MSA were also employed to define characteristic phenotypes for samples of the SHVTS by identifying molecular physiological (gene expression+physiology) parameters that best separated samples by temperature; the response variables underlying such temperature-partitioned phenotypes would be those most likely to be involved in the response of this widely distributed coral to temperature changes.

Table 1. Summary of eight environmental challenge studies performed at Taiwan's National Museum of Marine Biology and Aquarium between 2009 and 2014. In general, only corals exposed to 31.5°C for 2-4 weeks were found to bleach, die, and/or, more generally, display a significantly different phenotype that could be detected with the molecular physiological approach employed. $p\text{CO}_2$ =carbon dioxide partial pressure. ppm=parts per million. NA=not applicable. var=variable. OA=ocean acidification.

Year	Species	Sample material	Time-scale	Temp. (°C)	$p\text{CO}_2$ (ppm)	Major finding(s)	Reference(s)
2009*	<i>S. hystrix</i>	Colony	2 d	27 vs. 30	NA	No response to elevated temperature.	[12,14]
2010*	<i>S. hystrix</i>	Nubbin	7 d	26 vs. 23-29 (var.)	NA	Corals can acclimate to variable temperature, even if they have never before been exposed to such temperature regimes <i>in situ</i> .	[7,14,18, 20]
2010	<i>Pocillopora damicornis</i>	Larvae	10 d	25 vs. 29	400 vs. 630	No response to OA. Mild response to elevated temperature No interaction effect of OA and high temperature.	[22]
2010	<i>P. damicornis</i>	Nubbin	2 wk	25 vs. 29	400 vs. 850	No response to OA. Mild response to elevated temperature. No interaction effect of OA and high temperature.	
2010-2011	<i>P. damicornis</i>	Nubbin	9 mo	27 vs. 30	NA	No significant response to elevated temperature, albeit <i>Symbiodinium</i> affected more strongly than host.	[23-25]
2011	<i>P. damicornis</i>	Nubbin	4 wk	Up to 32 (var.)	NA	Corals can acclimate to high temperature if temperature decreases to ambient at night.	[26]
2011	<i>P. damicornis</i>	Nubbin	4 wk	27 vs. 31.5	NA	Exposure to 31.5°C for ~10 d elicits bleaching.	[26]
2014	<i>P. damicornis</i> <i>S. hystrix</i>	Nubbin	6 mo	25 vs. 31	400 vs. 1,000	Corals can acclimate to OA on a multi-month timescale.	

*dataset analyzed herein with MSA.

2. Materials and Methods

2.1 The experiments

The SHSTTE [12,14] and SHVTS [18,20] are described in prior works. Briefly, whole, unfragmented *S. hystrix* colonies from Houbihu were exposed to either the control (27 °C) or elevated temperature (30 °C) for 48 hr in the former, and RNAs, DNAs, and proteins were extracted from triplicate colonies housed within each of triplicate tanks at each of the two temperature treatments at each of four sampling times (6, 12, 24, and 48 hr; 18 samples/sampling time). A tank average was calculated across the three pseudo-replicates within the same tank sampled at each time, resulting in 24 data points that were analyzed by the MSA discussed below ($n=3$ biological replicates/sampling time/treatment \times 4 sampling times \times 2 temperature treatments). All coral colonies were collected under Kenting National Park permit 0992900398 issued to Dr. Tung-Yung Fan (2009). Univariate repeated measures ANOVAs were used previously [12,14] to assess the effects of time, treatment temperature, and their interaction on the 17 response variables described below.

In the SHVTS, six coral colonies were sampled from each of the two sites of origin (Houbihu [upwelling site] and Houwan [non-upwelling site]), and 48 nubbins were generated from the 12 colonies, all of which were of the same genotype [14]. Half of the nubbins from the six colonies of each site were randomly assigned to the stable temperature aquaria maintained at 26 °C, whereas the other half were placed into those aquaria of the variable temperature treatment, which fluctuated from 23–29 °C over a 6-hr period. The 12 tanks ($n=3$ for each interaction group) each contained four nubbins, two of which were sampled at time 0 (while all tanks were still at the acclimation temperature of 26 °C) and two of which were sampled after 7 d of treatment exposure; only the later 24 samples are discussed herein. In the case of the physiological response variables (discussed below), a tank average was calculated across the two pseudo-replicates sampled at the same time, resulting in a total sample size of 12 only. The molecular-scale data ($n=17$ parameters) were left unpooled for MANOVA, but not for PCA. MDS was used with the 12 pooled samples after having incorporated all 23 response variables (described below). These 23 parameters were previously assessed individually with two-way ANOVAs to determine the effects of site of origin, temperature regime, and their interaction [14,18,20].

2.2 Response variables

The same 17 molecular response variables were assessed in the samples of each experiment and included three biological composition parameters: 1) the *Symbiodinium* genome copy proportion (GCP; a proxy for cell density [21]), 2) the RNA/DNA ratio (a proxy for total transcription), and 3) the protein/DNA ratio (a proxy for total translation). Expression of 6 *Symbiodinium* of 8 host mRNAs was also quantified in each sample. The *Symbiodinium* genes spanned three cellular processes: photosynthesis, metabolism, and the stress response. The photosynthesis genes included ribulose-1,5-bisphosphate carboxylase/oxygenase (*rbcL*), photosystem I (subunit III; *psI*), and phosphoglycolate phosphatase (*pgpase*). The lone metabolism gene was nitrate transporter-2 (*nrt2*), and the two stress genes were ascorbate peroxidase (*apx1*) and *hsp70*. The host genes spanned the cellular processes of 1) cytoskeleton, 2) stress response, and 3) transport processes involved in osmoregulation. The former four were *actb*, *tuba*, tropomyosin (*trp1*), and *ezrin*. The three osmoregulation genes were transient receptor cation channel (*trcc*), organic anion transporter (*oatp*), and phospholipase- α 2 (*cplap2*). The lone stress gene was *hsp70*. The SHVTS featured six additional response variables for a total of 23 parameters assessed [14]. These included the *Symbiodinium* RBCL protein and five physiological response variables: growth, chlorophyll a concentration (chl_a; normalized two different ways [areal and per cell]), *Symbiodinium* density, and the maximum quantum yield of photosystem II (F_v/F_m).

2.3 MSA

Both experimental datasets were considered in an initial MDS analysis performed with PRIMER (ver. 5) in order to both display the composite dataset and visualize inter-experimental variation. In an MDS plot, the distance between two points is inversely proportion to their similarity; adjacent points are nearly identical, while those at opposite ends of the plot differ to the greatest extent. In this, and all other MDS analyses, Bray-Curtis similarity matrices were first created after having converted the data to Z-scores to account for the various parameters having different units of measure. Z-score transformations were used for all other MSA, and all data in the supplemental Excel spreadsheet represent Z-scores, and not raw values. After constructing the MDS plot, which featured the 17 molecular-scale response variables only, analysis of similarity (ANOSIM) was conducted with PRIMER to determine the effect of experiment on the composite molecular phenotype (i.e., gene expression+biological composition) of the coral samples. Global R distribution p -values were considered significant at an α of 0.05 based on 999 permutations. Heat maps were created with JMP (ver. 12) to portray the relative levels of the 17 molecular response variables only. Except for MDS, JMP was used for all statistical analyses.

For the SHSTTE and SHVTS datasets individually, PCA was first used to depict variation in two dimensions, and it was hypothesized that meaningful groupings of samples might be unveiled with this approach alone. PCA finds combinations of response variables that account for the greatest proportion of the variation in a dataset via their condensation into a principal component (PC). The second PC is constrained by being orthogonal to the first, and only the first two PCs were considered herein. PCA was conducted with a variety of different combinations of samples and response variables (Table 2) to find the minimum number of parameters that could visibly partition samples by temperature or time in the SHSTTE and by temperature regime or site of origin in the SHVTS.

MANOVA was next used to determine whether the multivariate means of the experimental treatment groups could be quantitatively separated within a 2-dimensional dataspace. Briefly, while univariate ANOVA tests each dependent variable in isolation from all others, MANOVA tests multiple ones simultaneously, generating a multivariate vector of univariate means (known as a "centroid") that depicts variation in multiple dimensions along canonical axes (CA). Not only can MANOVA determine differences between experimental treatments for multiple dependent variables, but it can also identify response variables that contributed most significantly to said differences. When sufficient replicates existed for the comparison of interest, Wilks' lambda values were calculated, and p values < 0.05 were considered to represent significance. When data points were missing for the time x temperature interaction groups in the SHSTTE, Roy's max root values were instead calculated.

For the SHSTTE, the effects of treatment temperature alone, time alone, and the interaction of temperature and time were tested by MANOVA ($n=17$ parameters), and for the SHVTS, MANOVA tested the effects of temperature alone, site of origin alone, and their interaction ($n=23$ parameters). MANOVA was also performed for subsets of response variables in each experiment: molecular parameters only ($n=17$ parameters), the *Symbiodinium* molecular response only ($n=6-7$ parameters), the host coral mRNAs only ($n=8$), the physiological variables only (SHVTS only; $n=4$ parameters), and photosynthesis parameters only ($n=3-6$; Table 2). For both experimental datasets, PRIMER was used to perform MDS using the Bray-Curtis similarity matrix on Z-score-transformed data, and ANOSIM was performed to determine the effects of time, temperature, and their interaction in the SHSTTE and the effects of site of origin, temperature treatment, and their interaction in the SHVTS. Finally, JMP's predictor screening function was used to rank the response variables in terms of their proportional contribution to the cumulative difference between sampling times and temperature in the SHSTTE and between site of origin and temperature regime in the SHVTS; only those parameters contributing > 10% of the cumulative difference are discussed herein.

Figure 1. MDS plot and heat maps of the SHSTTE and SHVTS datasets. The MDS analysis (a) was conducted with the 17 molecular response variables only since physiological parameters were not measured in samples of the SHSTTE. * $p < 0.01$. For the SHSTTE (b) and SHVTS (c) heat maps, the relative scale extends from dark blue (Z-score = -3) to dark red (Z-score = 5), and "x's" denote missing data. Samples marked by asterisks (*) were excluded from the MDS and most other MSA. Likewise, one sample from each of the Houbihu (HBH)-variable (var) and Houwan (HWN)-var groups was excluded from the heat maps themselves due to the respective RNA extractions having failed. Please see the main text for full names of the target genes. stab = stable.

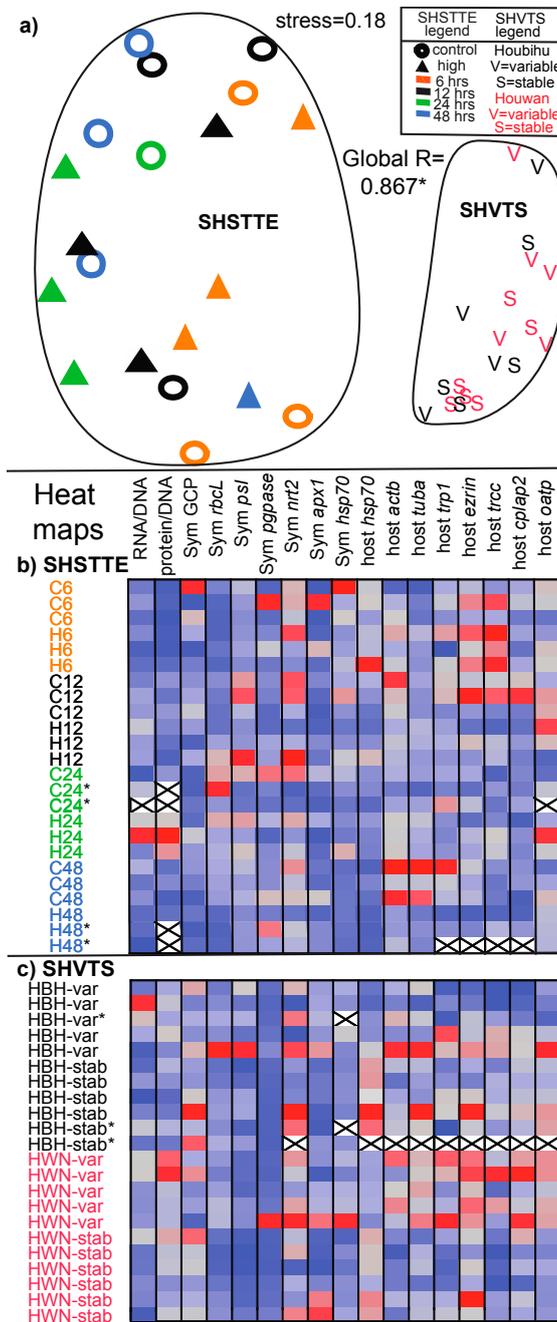


Table 2. Summary of comparisons and major findings. Dominant loading factors and canonical correlations were only included in the right-most column when the respective technique resulted in good partitioning of the data. “*Symbiodinium* molecular response” includes the GCP + 6 mRNAs for all comparisons except for PCA of the SHVTS, in which case the RBCL protein was also included. In certain cases, Wilks’ lambda or comparable MANOVA-based statistics could not be calculated due to having a large number of response variables relative to observations. Only when the *Symbiodinium* data were included could temporal partitioning (i.e., “discrimination”) be achieved in the SHSTTE; in contrast, data from the eight host coral genes were required to separate samples by temperature treatment (TT) and site of origin (SO) x TT in the SHVTS. *statistically significant observation. When *p*-values approached significance ($\alpha=0.05$), the exact values have been included. NA=not applicable. HBH=Houbihu. HWN=Houwan. stab=stable. var=variable.

Comparison-method	Figure	# parameters	# samples	Conclusion(s)	Dominant loading factors/canonical correlations
SHSTTE vs. SHVTS	1a	17	20 vs. 19	Experimental datasets are well separated*	<i>Symbiodinium hsp70</i> , host <i>oatp</i> , & RNA/DNA ^c
<i>MDS</i>					
SHSTTE (Figures 2-3)					
<i>PCA (Figure 2)</i>					
All response variables	2a	17	20	Time=6-hr samples are somewhat separated	Mix of host & <i>Symbiodinium</i> genes
<i>Symbiodinium</i> molecular response	2b	7	24	panmixia	
Host coral genes	2c	8	20	Some separation of time=6-hr samples	Mix of cytoskeleton, stress, & osmoregulation genes
All response variables (pooled)	2d	17	8	More separation by time than by TT	Mix of host & <i>Symbiodinium</i> genes
Photosynthesis parameters only		3 ^a	20	panmixia	NA
<i>MANOVA (Figure 3)</i>					
<i>Discrimination by time and TT</i>					
Host coral genes	3a	17	20	Times=6-hr & 24-hr are well separated*	Negative relationship between <i>Symbiodinium apx1</i> & <i>pgpase</i>
<i>Symbiodinium</i> molecular response		7 ^a	22	panmixia	
<i>Discrimination by TT only</i>					
Host coral genes	3b	17	20	panmixia	
<i>Symbiodinium</i> molecular response		7 ^a	22	panmixia	
<i>Discrimination by time only</i>					
Host coral genes	3c	17	20	Times=6-hr & 24-hr are well separated	Negative relationship between host <i>trcc</i> & <i>ezrin</i>
<i>Symbiodinium</i> molecular response		7 ^a	22	Time=6-hr separated from other 3 times*	Negative relationship between <i>apx1</i> & <i>psI</i>
<i>MDS (Figure 3)</i>	3d	17	20	Times=6-hr & 24-hr are well separated*	protein/DNA, <i>Symbiodinium hsp70</i> , & host <i>hsp70</i> ^c
SHVTS (Figures 5-6)					
<i>PCA (Figure 5)</i>					
All response variables	5a	23	12	Two TT are well separated	Mix of host & <i>Symbiodinium</i> genes
<i>Symbiodinium</i> molecular response	5b	8	12	Two TT are somewhat well separated	<i>Symbiodinium</i> genes & RBCL protein (PC1)
Host coral genes	5c	8	12	Two TT are well separated	Host <i>hsp70</i> (PC2)
Photosynthesis parameters only	5d	6	12	Two TT are well separated	<i>psI</i> & RBCL

MANOVA (Figure 6)					
Discrimination by SO x TT Physiological response only	6a	4	12	Two groups are well separated (HBH-var & HWN-stab)*	Fv/Fm
Discrimination by SO x TT Molecular response only	6b	17	20	Four SO x TT groups are well separated	Negative relationship between host <i>actb</i> & <i>Symbiodinium psI</i> +protein/DNA
Discrimination by SO x TT <i>Symbiodinium</i> molecular response	6c	7	19	Moderate separation by SO x TT ($p=0.059$)	
Discrimination by SO x TT Host coral genes only	6d	8	21	Two TT are well separated*	Negative relationship between <i>hsp70</i> & <i>tuba</i>
Discrimination by SO x TT All response variables	6e	23	12	Four SO x TT groups are well separated ^b	Negative relationship between <i>Symbiodinium hsp70</i> & <i>nrt2</i>
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Discrimination by SO only Physiological response only	4 ^a		12	Two SO are somewhat separated	
Discrimination by SO only Molecular response only	17 ^a		20	Two SO are well separated*	Negative relationship between <i>Symbiodinium pgpase</i> & <i>hsp70</i>
Discrimination by SO only <i>Symbiodinium</i> molecular response	7 ^a		20	panmixia	
Discrimination by SO only Host coral genes only	8 ^a		21	panmixia	
Discrimination by SO only All response variables	23 ^a		12	Two SO are well separated ^b	Negative relationship between <i>Symbiodinium psI</i> & <i>hsp70</i>
<hr/>					
Discrimination by TT only Physiological response only	4 ^a		12	Two TT are well separated ($p=0.052$)	Fv/Fm
Discrimination by TT only Molecular response only	17 ^a		20	Two TT are well separated ($p=0.058$)	Negative relationship between host <i>tuba</i> & <i>oatp</i>
Discrimination by TT only <i>Symbiodinium</i> molecular response	7 ^a		20	Two TT are well separated ($p=0.065$)	Negative relationship between <i>rbcl+hsp70</i> & <i>apx1+pgpase</i>
Discrimination by TT only Host coral genes only	8 ^a		21	Two TT are well separated*	Negative relationship between <i>hsp70</i> & <i>actb+tuba+trcc</i>
Discrimination by TT only All response variables	23 ^a		12	Two TT are well separated ^b	Negative relationship between host <i>hsp70</i> & <i>actb</i>
<hr/>					
MDS (Figure 6)	6f	23	12	Four SO x TT groups are well separated*	Physiological parameters & <i>Symbiodinium psI</i> ^c

^adata not depicted graphically. ^bcould not compute Wilks' lambda or Roy's max root. ^cdetermined by JMP's predictor screening function (Figure 4).

3. Results

3.1 Overview of the dataset

Prior to assessing variation and uncovering patterns within each of the two experiments, a collective MDS analysis was first conducted with 20 and 19 data points of the SHSTTE and SHVTS, respectively (Figure 1a); briefly 4 and 5 samples, respectively, out of the 24 total in each experiment were excluded from most of the MSA due to missing data. Some such samples are evident from heat maps of the SHSTTE (Figure 1b) and SHVTS (Figure 1c), and the associated data were generally lacking due to failed nucleic acid extractions. It is clear from the MDS plot and the corresponding ANOSIM p -value (0.002) that samples of the two experiments were well separated when looking at all 17 molecular response variables. It is also apparent that the SHSTTE demonstrated greater overall variation than did the SHVTS. Although there was some overlap, the stable and variable temperature samples of the SHVTS were somewhat separated from each other, and those samples sacrificed after 6 hr in the SHSTTE were slightly shifted to the right of those points of the other three sampling times. Both of these patterns are described in detail below using MSA specific to each dataset.

3.2 SHSTTE

3.2.1 PCA

PCA (on correlations) was first performed on all 17 molecular response variables across 20 of the 24 samples (Figure 2a), and the first two eigenvectors captured only ~40% of the variation. However, it is clear that those corals sacrificed at the 6-hr sampling time tended to partition away from the others. It was hypothesized that a reduction in the number of parameters could lead to eigenvectors collectively encompassing a greater percentage of the variation in the dataset. When looking only at the seven *Symbiodinium* parameters (GCP+ 6 mRNAs), the first and second PC encompassed ~67% of the variation (Figure 2b), and the response variable contributing the loading score with the highest positive value in PC1 was the *Symbiodinium* GCP. The second PC was dominated by the photosynthesis genes (excluding *rbcL*), meaning *Symbiodinium* density was negatively correlated with photosynthesis gene expression. PCA of the *Symbiodinium* parameters only did not partition the data by treatment temperature or time, and the data of both temperatures and all four sampling times were inter-mixed.

When looking at the host coral mRNAs only (Figure 2c), the first two PCs explained a similar percentage of the variation (~65%) as did the first two *Symbiodinium* PCs (Figure 2b); furthermore, as when looking only at the *Symbiodinium* response variables, samples did not appear well separated by temperature or time. However, the 6-hr data appear to be somewhat distinct from the others, with PC1 accounting for this separation. The dominant loading factors for PC1 were two cytoskeleton genes (*ezrin* & *trp1*) and two osmoregulation genes (*trcc* & *cplap2*). Expression of several cytoskeleton genes co-varied, as evidenced by the similar trajectory of their biplot axes, and three of the four cytoskeleton genes (excluding *ezrin*) contributed most significantly to PC2 in terms of eigenvector loading scores.

To determine whether extensive variation within treatments (i.e., a tank effect) accounted, in part, to the failure to document significant partitioning by temperature, data were pooled across triplicate tanks within each of the eight temperature x time interaction groups (Figure 2d). However, the first two PCs accounted for less than 60% of the variation, and it is clear that the four data points of each temperature were essentially mixed with those of the other. Despite this, the time 6- and 24-hr data were somewhat well separated, as was evident when all data were considered (Figure 2a).

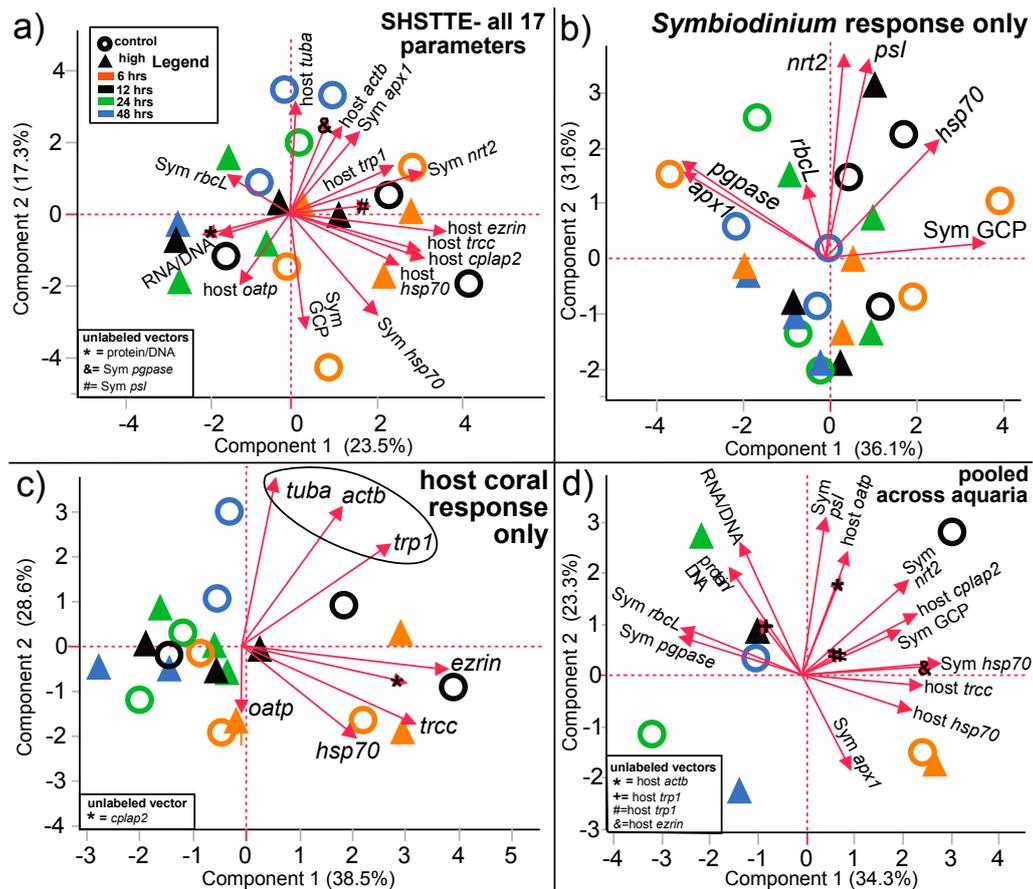


Figure 2. PCA of the SHSTTE dataset. All 17 response variables (a), including 3 biological composition parameters (RNA/DNA ratio, protein/DNA ratio, and the *Symbiodinium* GCP), expression of 6 *Symbiodinium* mRNAs, and expression of 8 coral mRNAs, were assessed across 20 of the 24 samples (four samples were omitted due to missing data points [see Figure 1.]). All 24 samples were included in the PCA of the *Symbiodinium* response variables only (GCP+6 genes; [b]). For the eight host genes (c), the same four samples as in (a) were omitted, and the ends of the vectors representing three cytoskeleton genes have been encircled to emphasize their co-variation. Data were also analyzed as pooled across aquaria for all 17 parameters for each of the eight temperature x time interaction groups (d). The legend in (a) applies to all panels.

3.2.2 MANOVA

When using MANOVA to test the interaction of temperature and time (Figure 3a) with data from all 17 response variables, Roy's max root was statistically significant, and this is likely due to the wide separation of samples of times 6 and 24 hr along CA1. This partitioning was driven by a negative relationship between *Symbiodinium pgpase* and *apx1* mRNA expression (Table 2). Within the 6- and 24-hr centroids, the high temperature samples were reasonably well separated from the controls, demonstrating the interaction of temperature and time. It should be noted, though, that only one sample comprised the control-24-hr group due to missing data. When looking at discrimination by temperature alone (Figure 3b), the control and high temperature groups were well separated along CA1; however, Wilks' lambda was not statistically significant. When looking at temporal discrimination only (Figure 3c), the 6-hr and 24-hr 95% centroids did not overlap, and were, furthermore, well separated across CA1. However, the Wilks' lambda was not statistically significant, potentially due to the substantial degree of overlap between the 12- and 48-hr centroids.

3.2.3 MDS and Predictor Screening

When looking at the MDS plot (Figure 3d), samples were significantly affected/sorted by time (ANOSIM Global R $p=0.002$), but not by temperature. Regarding the former factor, while the 12- and

48-hr samples were intermixed, the 6- and 24-hr times appear well separated, as was also evidenced by MANOVA (Figure 3a, c), and, to some extent, PCA (Figure 2a, c). PCA, MANOVA, and MDS all appear to suggest, then, that time, and not temperature, was more important in accounting for variation in the SHSTTE dataset. Therefore, the predictor screening function of JMP was used to identify the response variables that explained the greatest proportion of the differences between sampling times (Figure 4a), and these were found to be the protein/DNA ratio (21% of the cumulative difference), *Symbiodinium pgpase* mRNA expression (14%), and the RNA/DNA ratio (13%). MANOVA (Figure 3a) also found *Symbiodinium pgpase* to contribute to the separation of samples by time, specifically by distinguishing those samples of the 24-hr sampling time (Table 2).

Regarding temperature (Figure 4a), only three response variables contributed to 10% or more of the documented cumulative difference in the molecular physiology of the control and high temperature samples: host *trp1* (17%), *Symbiodinium apx1* (14%), and the RNA/DNA ratio (13%). However, the expression of neither gene, nor the RNA/DNA ratio, differed significantly between temperatures despite the fact that the expression of *trp1* was 4-fold less in high temperature samples. When looking at the interaction of time and temperature (Figure 4a), the RNA/DNA ratio and *Symbiodinium apx1* were the only factors that contributed to greater than 10% of the total difference between the four interaction groups (22 and 12%, respectively); neither showed an interaction effect when analyzed by repeated measures ANOVA ($p > 0.05$; [14]).

3.3. SHVTS

3.3.1 PCA

PCA, MANOVA, and MDS were all able to separate the samples by temperature regime in the SHVTS, and some approaches were able to separate the four interaction groups (Table 2). First, PCA was able to distinguish a variety of informative groupings (Figure 5a). When looking at all 23 response variables for data pooled across pseudo-replicates for each of the 12 aquaria, there was a clear separation of the stable and variable temperature samples along both PCs. However, the total variation encompassed by these two PCs was less than 55%. PC1 was comprised of a mix of host and *Symbiodinium* genes (Table 2), while the second PC consisted mainly of *Symbiodinium* photosynthesis genes. There was some degree of partitioning by site of colony origin within each temperature treatment, though still some overlap. Clearly, the effect of temperature on the molecular physiology of *S. hystrix* was greater than that due to site of origin upon assessment of PCA alone.

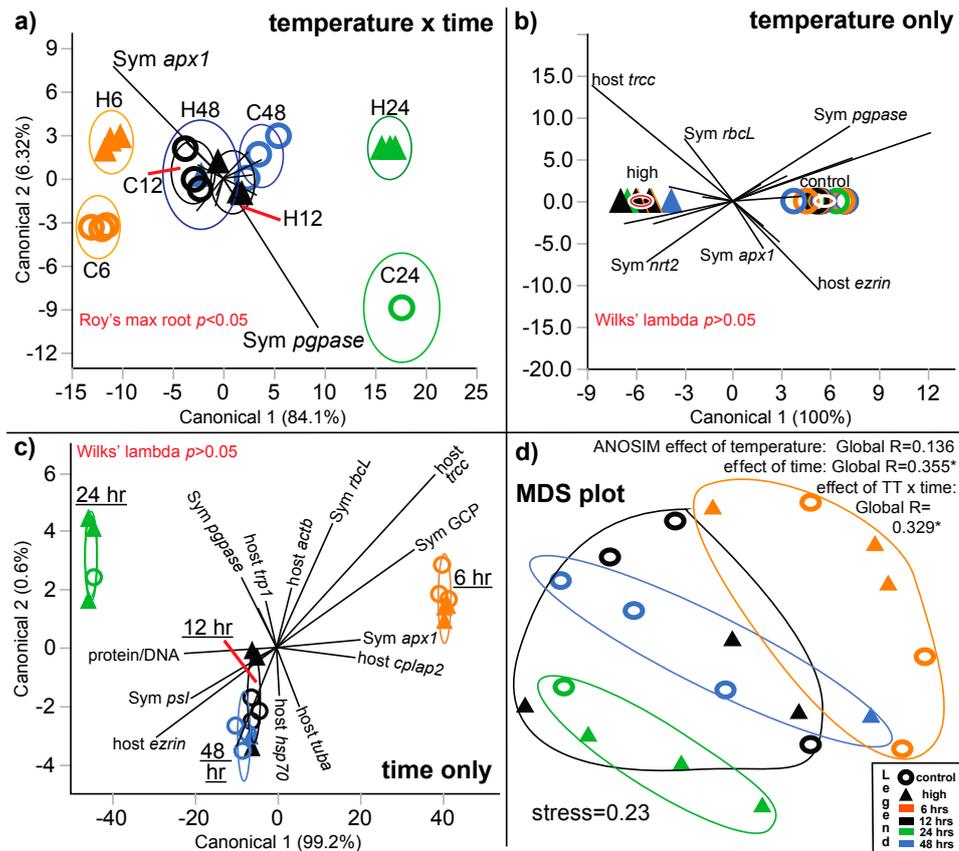


Figure 3. MANOVA and MDS analysis of the SHSTTE dataset. MANOVA was used to test for the effect of temperature \times time (a), temperature alone (b), and time alone (c), and 4 of the 24 total samples were omitted due to missing data points (see Figure 1.); therefore, Roy's max root, rather than Wilks' lambda, was calculated to test for a significant interaction effect in (a). In (a), the control-12-hr (C12) and high temperature-12-hr (H12) centroids (black circles; 95% confidence intervals for these and all centroids presented herein) lie within the high temperature-48-hr (H48; blue circle) centroid and have been labeled with red lines. In (b), the high and control temperature centroids are red with white lining and white only, respectively. In (a-c), not all axes have been presented or labeled due to spatial constraints in the panels themselves. MDS analysis of the same 20 samples with PRIMER (d). In this panel only, circles were drawn by hand and do *not* represent 95% confidence centroids. The legend for all four panels lies in the lower-right corner of (d). * $p < 0.05$.

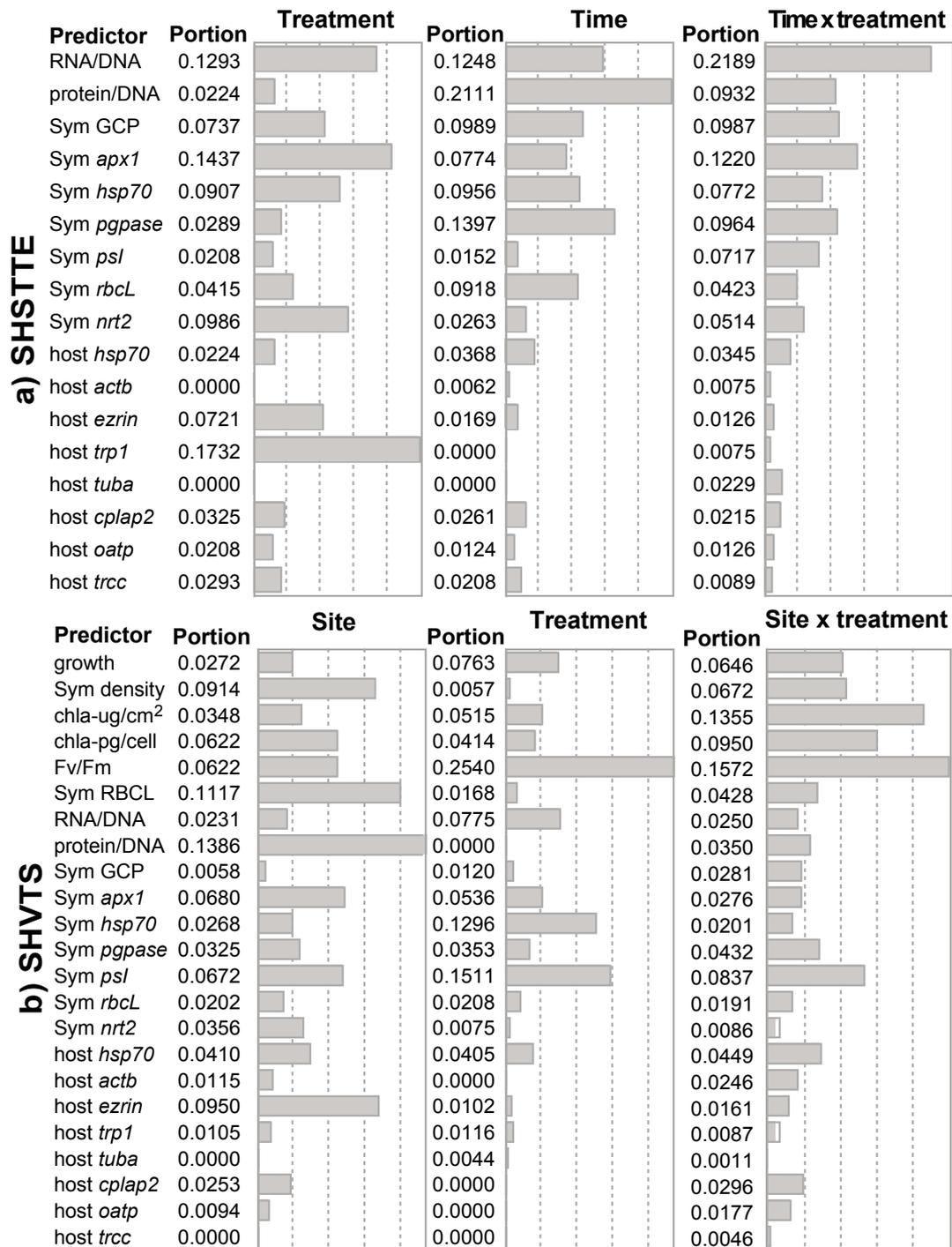


Figure 4. Predictor screening analysis of the SHSTTE and SHVTS. The predictor screening function of JMP was used to determine the response variables that accounted for the greatest proportions of the cumulative difference between temperature, time, and their interaction in the SHSTTE (17 response variables; [a]) and between temperature regime, site of origin, and their interaction in the SHVTS (23 response variables; [b]). It should be noted that the relative scales differ between the proportion plots.

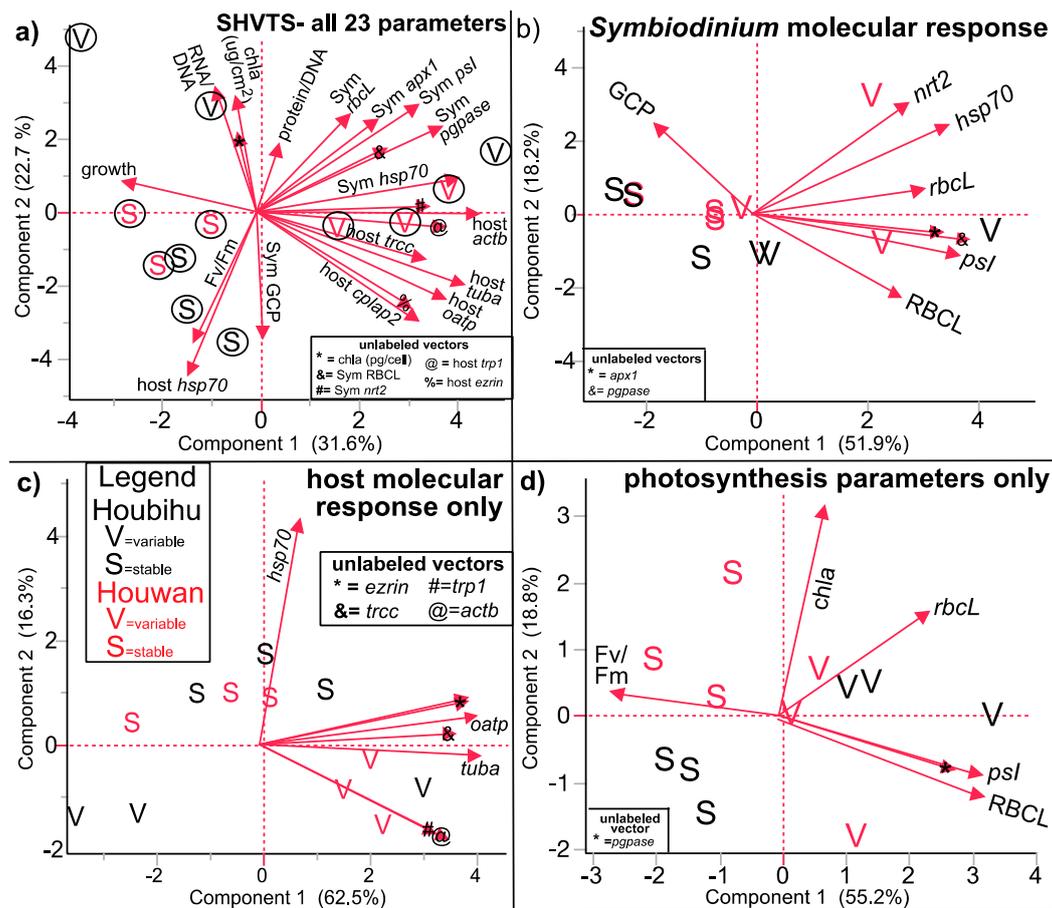


Figure 5. PCA of the SHVTS. Data from pseudo-replicate samples of the same tank were pooled, resulting in 12 data points in each plot. a) All 23 parameters. b) The *Symbiodinium* molecular response only (GCP+6 mRNAs+RBCL protein). One Houwan-stable icon is nearly masked by a Houbihu-stable one towards the left side of the plot. c) The host molecular response only (8 mRNAs). In (c), the unlabeled *cplap2* axis is below the *ezrin* one. Likewise, the *trp1* and *actb* axes are overlapping. d) The six photosynthesis parameters only. The legend for all panels is located in (c).

In order to reduce the complexity of the dataset, PCA was also conducted only with the eight *Symbiodinium* molecular response variables: the GCP, the six mRNAs, and the RBCL protein (Figure 5b). It is clear that samples of the two temperature regimes, stable and variable, could be distinguished by PC1 (51.9%), which was dominated by the photosynthesis genes in terms of highest positive eigenvector loading factor scores (Table 2). The stable temperature samples tended to cluster together, with the six variable ones showing greater variability and spread throughout the dataspace. Furthermore, the Houwan variable temperature samples showed greater dispersal than did the Houbihu ones. The *Symbiodinium* GCP was the most significant contributor to the second PC (18.7%); this indicates that *Symbiodinium* photosynthesis gene expression was negatively correlated with *Symbiodinium* density, as was also the case in the SHSTTE. When looking at individual correlations of photosynthesis gene expression vs. *Symbiodinium* GCP (data not shown), all slopes were negative; however, these slopes were not significantly different from 0 (linear regression *t*-tests, $p > 0.05$).

When looking at the host coral molecular response only (Figure 5c), it is clear that PCA was able to partition samples of the two temperature regimes. The first PC encompassed all three cytoskeleton genes and four osmoregulation genes and explained 62.5% of the variation (Table 2). The second PC consisted of *hsp70* as the only positive loading score to the eigenvector, and this PC encompassed 16.3% of the variation. It is also evident that the osmoregulation and cytoskeleton genes tended to covary. As with the PCA of the *Symbiodinium* molecular response variables only (Figure 5b), the spread of the variable temperature regime samples was greater than that of the stable temperatures ones. In contrast to the results of the PCA of the *Symbiodinium* molecular response only, though, the Houbihu

variable temperature samples showed more spread than those of Houwan exposed to this same temperature regime. Finally, when looking only at the six photosynthesis parameters (Figure 5d), it is clear that samples of the two temperature regimes were well separated along PC1 (55.2%), in which *psI* gene and RBCL protein expression contributed the highest positive loading scores (Table 2). Samples of the two sites of origin for the stable temperature treatment were separated along PC2, in which case *chl a* content (pg/cell) contributed most significantly to said partitioning.

3.3.2 MANOVA

A variety of combinations of response variables were used to see which best modeled differences between the four site of origin x temperature treatments groups with MANOVA (Table 2). First, MANOVA was performed with four of the five physiological response variables alone (areal *chl a* was excluded in place of *chl a*/cell; Figure 6a). Although Wilks' lambda for the interaction effect was significant, only two of the four groups appear well-separated: Houbihu-variable and Houwan-stable. Houbihu-stable and Houwan-variable samples appear inter-mixed. Fv/Fm was the most significant factor contributing to the separation of the former two groups (Table 2). When looking at all 17 molecular response variables with 20 of the 24 samples, all four interaction groups were well separated (Figure 6b), though Wilks' lambda could not be computed due to the large number of response variables relative to the number of biological replicates (~1:1). A negative relationship between host *actb* and *Symbiodinium psI*+protein/DNA drove the partitioning of the four interaction groups (Table 2), though such partitioning was not considered statistically significant by Roy's max root.

After performing a MANOVA of seven of the eight *Symbiodinium* molecular response variables only (RBCL protein expression was excluded since it was only assessed in 12 of the 24 samples; Figure 6c), it is clear that this sub-set was unable to differentiate the four interaction groups. In contrast, when looking at the eight host genes alone (Figure 6d), significant discrimination was achieved. However, the two sites of origin were only well separated in the variable temperature dataset and were intermixed for the stable temperature regime samples. When looking at all 23 response variables, the four site of origin x temperature treatment interaction groups were well separated (Figure 6e), though neither Wilks' lambda nor Roy's max root could be calculated due to the large number of response variables relative to observations.

3.3.3 MDS and Predictor Screening

In contrast to MANOVA, MDS can readily quantify relationships between samples even when the number of response variables is large relative to the number of observations, and MDS was able to distinguish the four interaction groups of the SHVTS (Figure 6f). ANOSIM found site of origin, temperature treatment, and interaction effects to be significant, and, within the stable temperature regime samples, it is clear that data points of the two sites of origin were well separated. However, there was some overlap between the Houbihu-variable and Houwan-variable samples due to one Houbihu-variable data point falling closer to those points of the latter group. JMP's predictor screening function (Figure 4b) found the five physiological parameters and *Symbiodinium psI* to contribute most significantly to the cumulative difference between the four interaction groups (Table 2). *psI* also contributed significantly to the cumulative difference between temperature regimes (Figure 4b), in conjunction with Fv/Fm and host *hsp70* (Table 2).

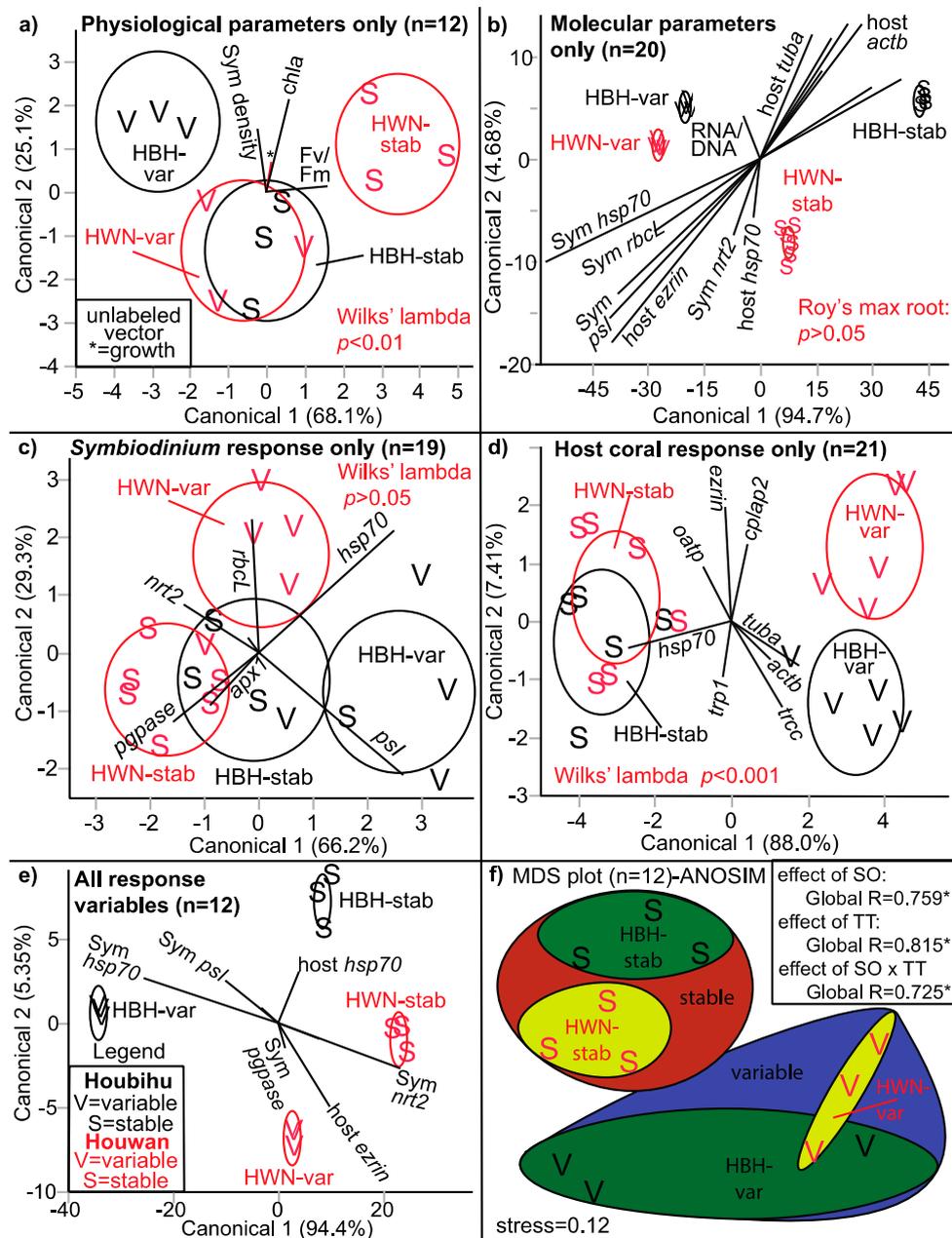


Figure 6. MANOVA and MDS analysis of the SHVTS. MANOVA was used to model differences between the four site of origin (SO) x temperature treatment (TT) interaction groups with four of the five physiological parameters (excluding areal *chl*_a; [a]), all 17 molecular response parameters (b), seven of the eight *Symbiodinium* molecular response variables (excluding the RBCL protein; [c]), all eight host coral mRNAs (d), and all 23 response variables (e). The sample sizes displayed in the individual panels reflect the number of samples and not the number of response variables. Wilks' lambda could not be computed in (b) and (e) due to the large number of response variables relative to the number of samples. In (b), not all axes have been labeled. In (c), the *Symbiodinium* GCP axis falls between those of the mRNAs *nrt2* and *rbcl*. In (e), only the dominant axes/response variables have been shown due to spatial constraints in the panel itself. In (a-d) Wilks' lambda (a, c, d) and Roy's max root (b) values test the interaction of SO and TT. In (f), PRIMER's MDS function was used to portray the SHVTS dataspace, and the circles were hand-drawn to encompass the four SO x TT groups. All other circles represent 95% confidence centroids. The legend in (e) corresponds to all panels. * $p < 0.01$.

4. Discussion

This represents the first work to exploit MSA for assessing molecular and physiological response variables spanning both compartments of an endosymbiotic organism, and the conceptual framework for doing so will ideally be of use to others interested in understanding how dual-compartmental organisms respond to changes in their environment. As mentioned above and in Table 1, corals of Southern Taiwan have proven to be resilient to a number of laboratory-induced environmental challenges, and MSA confirmed this univariate ANOVA-based observation for samples of the SHSTTE; specifically, there was no evident difference between samples of the control and high temperature treatments. This leaves at least two hypotheses remaining: 1) the corals were truly unresponsive to a short-term exposure to 30°C or 2) non-temperature-sensitive parameters were chosen. Given the well-documented photoinhibition that occurs when *Symbiodinium* are exposed to elevated temperatures (e.g., [27]), it seems likely that at least several photosynthesis genes should have undergone changes in mRNA expression. However, a recent work [28] found virtually no congruency between gene and protein expression in *S. hystrix* or its endosymbiotic dinoflagellate populations. Therefore, it could be that the respective photosynthesis- and stress-targeted *proteins* indeed underwent changes in expression upon exposure to 30°C for 48 hr, even though their respective mRNAs did not. Future work should, then, attempt to characterize the proteomes of pocilloporids, and other reef corals, exposed to theoretically stress-inducing temperatures to uncover the sub-cellular basis of the stress or acclimation response, whichever the case may be.

Despite the absence of a multivariate temperature treatment effect in the SHSTTE, there were some notable temporal differences, particularly when looking at the corals sampled after 6 hr of treatment exposure; upon using MANOVA, a mix of biological composition, host gene expression, and *Symbiodinium* gene expression data best separated the 6-hr samples from the others, and both MANOVA and JMP's predictor screening function found the protein/DNA ratio to account significantly for this temporal difference. Indeed, the protein/DNA ratio was found previously to be temporally variable in these samples [14]. Furthermore, negative correlations between two *Symbiodinium* genes (the stress gene *apx1* and the photosynthesis gene *pgpase*) and two host coral genes (the cytoskeleton gene *ezrin* and the osmoregulation gene *trcc*) genes were found to partition samples of the 6-hr time from those of the 24-hr time (i.e., the two groups that were most distinct from one another); none of these genes were found to be temporally variable in expression by univariate repeated measures ANOVA [14], demonstrating the utility of MSA in defining combinations of response variables that best explain patterns within a dataset.

From the MANOVA, MDS analysis, and, albeit less so, PCA, it is clear that corals of the four sampling times possessed unique gene expression+biological composition signatures, and this temporal change in the molecular phenotype of these samples may be related to the complex, dual-compartmental metabolism displayed by organisms, such as reef-building corals, that have acquired the capacity for photosynthesis via symbiosis [29-30]. Specifically, coral metabolism is temporally variable due to the periodic nature of light-driven photosynthesis [30-32], and metabolic hysteresis driven by dinoflagellate photosynthesis as a function of the light cycle surely contributed to the temporal variation observed in the SHSTTE. Circadian rhythm may also have accounted, in part, for the separation of the 6- and 24-hr samples in the SHSTTE [33]. The former were collected at 13:45, and, while stable, artificial light was used in this experiment, it is possible that the temporal gene expression signatures were driven by an entrained response to high light levels that would normally be experienced at such times. The 12, 24, and 48-hr sampling times corresponded to 19:30, 7:30, and 7:30, respectively, times at which light levels would be relatively low *in situ*. However, the experimental corals had been reared under non-fluctuating, artificial light (12:12hr light-dark) for nearly one month at the time of sampling (including the pre-experimental acclimation period), and circadian rhythm can be abolished within two days of changing the light regime in endosymbiotic anthozoans [30]. Therefore, other factors besides metabolic hysteresis due to photosynthesis and circadian rhythm may have accounted for the unique molecular phenotype of corals sampled after 6 hr.

All MSA documented temperature regime, and oftentimes site of origin, differences in the SHVTS, and the host coral parameters were actually more likely to partition samples by temperature in the SHVTS; this contrasts with the SHSTTE, in which the *Symbiodinium* response variables were relatively more important in the separation of samples across the dataspace. However, the predominant experimental factor leading to sample partitioning in the SHSTTE was time, rather than temperature. As *Symbiodinium* gene expression, and physiology in general, is known to be highly dynamic [30,33], this finding was not unexpected. Importantly though, the fact that the *Symbiodinium* response variables more significantly contributed to temporal variation in the SHSTTE, while host coral parameters led to a relatively greater partitioning of samples by temperature in the SHVTS, emphasizes that notion put forth by Mayfield & Gates [29] and others that it is important to consider both compartments of the coral-*Symbiodinium* endosymbiosis when attempting to gauge the molecular physiological response of the composite holobiont to environmental change.

When performing PCA on the *Symbiodinium* molecular response only, the Houwan variable temperature samples showed greater dispersal than did the Houbihu ones. This could be because the former corals had never before experienced such variable temperature profiles *in situ*; as such, the variability in their response to fluctuating temperatures could be hypothesized to be greater than that of corals of Houbihu, which *do* routinely experience upwelling. Likewise, when looking at the MDS plot of the SHVTS dataset, the spread of the variable temperature samples was greater than that of the stable ones, and a similar explanation could account for this observation. Indeed, variability in the physiological response to an environmental change has been predicted to be important in gauging marine animal health [34].

Curiously, though, the Houbihu samples exposed to variable temperatures showed a greater diversity in their molecular physiological response in the MDS plot than those of Houwan exposed to this profile, in contrast to what was observed with PCA for the *Symbiodinium* response only. This variable reaction of Houbihu samples exposed to fluctuating temperatures may be due to differential acclimation strategies between the original colonies, which may have been from different microhabitats within the Houbihu reef system. Although an effort was made to collect colonies at similar depths in a reasonably small area (~10-100 m of each other; [18]), it is possible that the light environment *in situ*, for instance, may have differed between the colonies used to make the nubbins. Nubbins from the six colonies from Houbihu were mixed in a seawater table and randomly assigned to each temperature regime. Therefore, the Houbihu-variable MDS outlier may have represented a nubbin from a colony that experienced a different abiotic environment *in situ* than the other two nubbins of that interaction group. In short, differing environmental histories of the colonies of Houbihu may have contributed to biological variation in the dataset, whereas the colonies removed from Houwan may have been characterized by more similar environmental histories and thus behaved more similarly when exposed to a foreign temperature regime. Regardless of the explanation, it is clear the molecular physiology differed significantly between corals of the two sites of origin and two temperature regimes, indicating that corals of the four interaction groups displayed distinct behavior with respect to the 23 response variables assessed herein.

Although MSA were successfully used to define time-specific phenotypes in the SHSTTE and molecular physiologies with fidelity to both site of origin and temperature regime in the SHVTS, there is not, as mentioned above, a significant degree of congruency between gene and protein expression in this reef coral [28]; therefore, although gene expression signatures may be used to partition corals from multiple sites and temperature profiles within an experimental dataspace in order to uncover intra- and inter-experimental differences, care should be taken before using such gene expression trends to enact mechanistic reconstructions of cell physiology, as has become standard in the field of coral biology [15, 35-36]. Rather, the proteins that actually conduct cellular work are better molecular targets for those interested in making statements as to how corals respond to, for instance, changes in their abiotic environments. Such proteome-scale data could be analyzed in an analogous manner as was conducted herein in order to define protein expression signatures that underlie the sub-cellular capacity for reef corals to acclimate to global climate change. MDS is an especially well-suited means of displaying a molecular phenotype that integrates a number of

different response variables and macromolecules as, unlike MANOVA, ANOSIM can still be performed when the number of response variables is large relative to the number of samples. As such, it could hypothetically be used to screen for protein biomarkers of the coral response to environmental perturbation.

5. Conclusions

The MSA utilized herein defined molecular signatures across time in the SHSTTE dataset, which was largely found to feature negative findings (i.e., no significant change) when analyzed by traditional, univariate approaches alone [12,14]. Rather than the absolute expression level of a gene mRNA characterizing a sample group, relationships between multiple response variables and genes were instead found to better distinguish corals sampled at different times. In the SHVTS, multiple groupings of response variables (e.g., gene expression, physiological, and biological composition parameters) could partition samples by temperature regime, and the molecular physiological phenotypes differed significantly between corals of these two temperature profiles. Unlike the SHSTTE, in which the *Symbiodinium* response was a greater contributor to the overall variation of the dataset, host coral response variables better partitioned data points of the two temperature regimes in the SHVTS. Furthermore, corals exposed to variable temperatures showed a greater range in their molecular physiological response relative to those exposed to stable temperature; however, depending on the MSA used, the spread in the dataspace was sometimes greater for samples of Houbihu, which experience both stable and variable temperatures *in situ*, and sometimes greater for those of Houwan, which only experience stable temperatures *in situ*. This result is perplexing and could be driven by spatial heterogeneity in the abiotic environment of Houbihu, a reef characterized by extensive temporal environmental variation due to upwelling [16].

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