Recent reconfiguration of an ancient developmental gene regulatory network in

Heliocidaris **sea urchins**

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ABSTRACT

Changes in developmental gene regulatory networks (dGRNs) underlie much of the

diversity of life[1], but the evolutionary mechanisms that operate on interactions with these

networks remain poorly understood. Closely related species with extreme phenotypic

divergence provide a valuable window into the genetic and molecular basis for changes in

dGRNs and their relationship to adaptive changes in organismal traits. Here we analyze

genomes, epigenomes, and transcriptomes during early development in two sea urchin

species in the genus *Heliocidaris* that exhibit highly divergent life histories and in an

outgroup species. Signatures of positive selection and changes in chromatin status within

putative gene regulatory elements are both enriched on the branch leading to the derived

life history, and particularly so near dGRN genes; in contrast, positive selection within

protein-coding regions have at most a modest enrichment in branch and function. Single-

cell transcriptomes reveal a dramatic delay in cell fate specification in the derived state,

which also has far fewer open chromatin regions, especially near dGRN genes with

conserved roles in cell fate specification. Experimentally perturbing the function of three

key transcription factors reveals profound evolutionary changes in the earliest events that

pattern the embryo, disrupting regulatory interactions previously conserved for ~225

million years. Together, these results demonstrate that natural selection can rapidly

reshape developmental gene expression on a broad scale when selective regimes abruptly

change and that even highly conserved dGRNs and patterning mechanisms in the early

embryo remain evolvable under appropriate ecological circumstances.

MAIN

The well-defined dGRN of early development in sea urchins[2] provides a powerful

framework for investigating the evolution of embryonic patterning mechanisms. Interactions

between genes encoding transcription factors and their target genes within this dGRN are almost

completely conserved among species that diverged ~30-40 million years (my) ago[3,4], with some

interactions conserved for ~225 my[5], ~275 my[6], or even ~480 my[7,8]. One possible explanation

for this observation is developmental constraints, such that early developmental processes are

largely immutable given their critical roles in body plan organization and tissue specification[9,10].

Under this scenario, any change in a critical interaction during early development would have

widespread effects on later processes, which would almost always be deleterious. Still, an

important confound remains untested: the species with deeply conserved developmental

mechanisms all share the same life history mode, involving low maternal provisioning and an

extended feeding larval phase. Species with derived life histories involving massive maternal

provisioning and highly-abbreviated, nonfeeding pre-metamorphic development have evolved on

multiple occasions within sea urchins[11-13], possibly in response to lower or more unpredictable

food availability[14]. These species can reveal how conserved regulatory interactions and

patterning mechanisms respond to major shifts in selective regimes.

The Australian sea urchin genus *Heliocidaris* includes two recently diverged species: *H.*

tuberculata representing the ancestral life history and *H. erythrogramma* the derived state[15]

(Fig.1a). The shift to nonfeeding development radically alters natural selection on development:

with feeding no longer necessary, high mortality rates in the plankton[16] impose strong selection

to decrease time to metamorphosis[12]. Numerous anatomical features and gene expression profiles

of early development that are broadly conserved among sea urchins differ markedly between

these closely related species[17-20](Fig.1b). We sought to learn whether these recently evolved

differences are merely superficial and mask deeply conserved developmental mechanisms, or

whether they are the product of substantive evolutionary changes in early cell fate specification

and dGRN organization. Evidence for the former would suggest developmental constraints play

an important role in limiting genetic and regulatory composition of the ancestral GRN, whereas

support for the latter would point to a flexible morphogenetic system derived from an embryonic

program conserved at least in part by stabilizing selection that is adaptable to alternative

developmental life histories.

RESULTS

Natural selection has sculpted the regulatory landscape of *H. erythrogramma*

We took advantage of the recent (~5 my) divergence between the two *Heliocidaris*

species to carry out detailed analyses of orthologous coding and noncoding regions of the

genome, focusing on the transcription factors and regulatory elements that constitute the

backbone of the dGRN and underlie cell fate specification mechanisms (see Supplementary Data

1 for list of 192 dGRN genes). Genomes of *H. erythrogramma* and *H. tuberculata* were each

sequenced, assembled into twenty-one full length chromosomes, and annotated (Fig.1c). Genome

sequences were then aligned to one another and to that of *Lytechinus variegatus*[21], an outgroup

representing the ancestral life history condition (Fig.1a).

To understand how natural selection altered the genomes of *Heliocidaris* during the

evolution of nonfeeding development, we began by testing for evidence of branch-specific

positive selection within single-copy protein coding regions[22]. At a genome-wide scale, we found

statistical support for modest enrichment of positive selection along the *H. erythrogramma*

branch, but not the *H. tuberculata* branch, when considering the full set of genes (Fisher exact

test, two-sided: $p < 1.33 \times 10^{-3}$). Of note, coding sequences of dGRN genes showed no enrichment

of positive selection on either branch (Fig.1d-e). This result provides little support for the idea

that changes in transcription factor structure and function are primarily responsible for the

extensive modifications in development and life history in *H. erythrogramma*. Scant evidence of

positive selection in the coding sequences of dGRN genes likely reflects pleiotropic constraints

imposed by the multiple functions that their encoded transcription factors execute during cell

type specification and differentiation.

Therefore, we hypothesized that functional changes in regulatory elements are instead

largely responsible for these trait changes. We carried out ATAC-seq on the two *Heliocidaris*

species and *L. variegatus* to identify open chromatin regions (OCRs) representing putative

regulatory elements in early (hatched) blastula stage embryos, by which time initial cell fates

have been specified in the ancestral dGRN. Our analyses are based on all OCRs present in at

least one species and located within a genomic region with 1:1:1 orthology among all three

species. We then tested for branch-specific positive selection within these OCRs using an

approach analogous to that described above for protein-coding regions. At a global scale, these

putative enhancer and promoter regions are enriched for evidence of positive selection on the

branch leading to *H. erythrogramma* (545) relative to *H. tuberculata* (347) (Fisher exact test,

two-sided: $p < 1.33e-11$) (Fig.1f-g). This higher incidence in signatures of positive selection

specifically within OCRs on the *H. erythrogramma* branch is indicative of positive selection on

regulatory element function that is remarkably widespread within its genome and is consistent

with our earlier finding that many expression differences between the two *Heliocidaris* species

are genetically based in *cis*[23].

Strikingly, signals of *H. erythrogramma*-specific positive selection are even more

enriched when only considering OCRs near dGRN genes (Fig.1f-g; difference in median zeta:

0.182; Fisher exact test, two-sided: p<5.33e-5). In all, 26 putative regulatory elements located

near 23 distinct dGRN genes exhibit evidence of positive selection on the *H. erythrogramma*

branch, as opposed to just four on the *H. tuberculata* branch (Fig.1c,g). These 23 genes represent

17.0% of the total within the defined dGRN with a nearby OCR, a marked enrichment compared

with the remainder of the genome, where positive selection is detected in OCRs near just 5.7% of

genes (Fisher exact test, two-sided: p<4.92e-4: Fig.2c).

Two distinct regulatory mechanisms underlie divergence in transcriptomes

While the accessibility of most OCRs and expression[19,23] of most genes are conserved

between species (see Fig. 2b and Extended Data Fig. 1 for examples), we observed a striking

decrease in chromatin accessibility of many putative regulatory elements throughout the *H.*

erythrogramma genome relative to both species representing the ancestral life history (Fig.2a).

Of 2,625 orthologous, differentially accessible OCRs between developmental modes, 1,795 sites

(68.4%) are significantly less accessible in *H. erythrogramma* (e.g. Fig.2b: *hesC*). Because

decreased chromatin accessibility can limit transcription factor access to regulatory elements and

because most regulatory interactions in the early sea urchin embryo involve activation of

transcription[24], widespread evolutionary reduction in chromatin accessibility throughout the

genome in *H. erythrogramma* embryos suggests an important role for evolutionary changes in

chromatin configuration for divergence in gene expression, in this case associated with generally

decreased or delayed zygotic transcription for many genes. This interpretation is consistent with

indications of a broad delay in embryonic cell fate specification in this species[19,25-29].

In a previous study[19] we analyzed changes in temporal gene expression profiles during

early development within *Heliocidaris* and found that the largest changes are concentrated on the

branch leading to *H. erythrogramma* and are enriched for developmental regulatory genes

generally and dGRN genes specifically. Results reported here suggest that these derived

expression profiles are the product of two distinct molecular mechanisms that alter transcription

factor binding: changes in nucleotide sequence and changes in chromatin configuration (Fig.2c-

d). The former may alter protein:DNA binding while the latter may alter protein access to

regulatory elements. Both modes of regulatory evolution are concentrated near dGRN genes

relative to the rest of the transcriptome in *H. erythrogramma* (Fig.2c). Notably, accelerated

sequence evolution or altered chromatin state (or both) is present in an OCR near a differentially

expressed dGRN gene ~3 times more frequently than the rest of the transcriptome in *H.*

erythrogramma, while no such relationship is evident in *H. tuberculata*.

The distribution of both mechanisms of regulatory evolution is highly nonrandom within

the genome (enriched near differentially expressed genes and near developmental regulatory

genes) and phylogenetically (enriched on the *H. erythrogramma* branch). These departures from

the null expectation of random distribution (i.e., resulting from genetic drift) suggest that many

of the specific changes are adaptive. Adults of the two *Heliocidaris* species occupy overlapping

habitats and ranges[30], making the suite of derived life history traits that evolved on the *H.*

erythrogramma branch the most plausible driver for many of the extensive gene regulatory

changes.

The timing of cell fate specification has changed in *H. erythrogramma*

To understand how changes in the regulation of gene expression in the *H. erythrogramma*

embryo influenced developmental mechanisms and life history traits, we leveraged information

about the ancestral dGRN to examine embryonic cell fate specification. The earliest zygotic

patterning event in the ancestral state involves specification of skeletogenic and germ cell fates

following two successive unequal cleavages of vegetal blastomeres[31]. We focus here on the well-

characterized skeletogenic cell lineage, which rapidly establishes a distinct transcriptional state[32]

and, within 24 hours after fertilization, undergoes an epithelial-to-mesenchymal transition

(EMT), fully differentiates, and begins to synthesize a complex larval endoskeleton (Fig.1b).

Specification and maintenance of the skeletogenic cell fate is regulated by interactions between

~11 transcription factors[33] (Fig.4a). These developmental events and most of the underlying

dGRN interactions are conserved across >225 my of sea urchin evolution[6].

Morphological development of the skeletogenic cell lineage in *H. erythrogramma* differs

in several regards from this ancestral state: cleavage divisions are all equal, no cells undergo

EMT before gastrulation, and the larval skeleton is delayed and reduced[20] (Fig.1b). To

understand whether underlying developmental mechanisms are conserved despite these overt

morphological differences, we carried out single-cell RNA-sequencing (scRNA-seq) of early

blastula stage *H. erythrogramma* embryos and compared the results with our published scRNA-

seq data from *L. variegatus*[34] at the same early blastula stage (prior to EMT). We chose this stage

because many major cell fates have been specified by early blastula in the ancestral condition[2,31-

33]. This result is clearly reflected in the UMAP of *L. variegatus*, which contains seven cell

clusters (Fig. 3a), each expressing a distinct suite of regulatory proteins predicted by the dGRN,

with skeletogenic cells exhibiting a particularly disparate transcriptional state (Fig.3c;

Supplementary Fig. 1). These indications of early cell fate specification and rapid divergence in

transcriptional states are also apparent in scRNA-seq data from *Strongylocentrotus purpuratus*[35],

another sea urchin representing the ancestral condition (Fig.1a), suggesting early embryonic

regulatory interactions are conserved among planktotrophic species and detectable by scRNA-

seq.

In *H. erythrogramma*, only three cell clusters are apparent at the same resolution and

developmental stage (Fig.3b). The observation of fewer distinct transcriptional states in *H.*

erythrogramma along with less localized expression of known GRN tissue marker genes

(Supplementary Fig. 1) suggests a delayed establishment of distinct transcriptional states in the

early embryo of this species – a conclusion not attributable technical factors such as analytical

parameterization (see Methods), cell number (Extended Data Fig. 2) or genes/UMIs per cell

(Supplementary Fig. 2). Given the limited number of cells in this dataset (2065 post-filtering)

and representation from a single stage, future work examining later developmental stages will be

necessary to fully resolve the timing of cell fate specification in *H. erythrogramma* and

associated expression of developmental regulatory factors.

The presence of more numerous transcriptional states in *L. variegatus* is even clearer in

an integrated projection of expression for 7,671 one-to-one orthologous genes, where *L.*

variegatus clusters remain separated but those of *H. erythrogramma* overlap broadly in UMAP

space, suggesting cells of this species have more homogenous transcriptional profiles relative to

cells in *L. variegatus* at the same developmental stage (Fig.3d, compare lower center and lower

right panels as well as insets). Three independent methods of scRNA-seq data integration are

consistent in showing fewer clusters and greater degree of overlap among cells from different

clusters in *H. erythrogramma* (Extended Data Fig. 3; see Methods). These findings are consistent

with the ATAC-seq results presented above and our earlier lineage-tracing and bulk RNA-seq

studies[19, 23, 25, 26], all of which indirectly point to a delay in fate specification in *H.*

erythrogramma.

Further, clusters in the early *H. erythrogramma* do not express similar suites of

transcription factors to those in the ancestral state, and none corresponds to the distinctive

skeletogenic cell lineage of *L. variegatus* and *S. purpuratus* that is established earlier in their

development (Fig. 3c,d; Supplementary Fig. 1,3). For instance, *delta* and *alx1*, which encode

critical early regulatory proteins, are expressed exclusively in the skeletogenic cell precursors at

the blastula stage in the ancestral state[36,37]. In *H. erythrogramma*, localized transcription of

neither *delta* nor *alx1* has commenced (Supplementary Fig. 1). Furthermore, a composite of

skeletogenic cell marker gene expression is localized to a discrete cell population in the ancestral

state but is not detected in the *H. erythrogramma* embryo (Fig. 3c). Together, these expression

differences suggest that the roles of key regulators of the skeletogenic cell fate have evolved

during the life history shift.

The ancient double negative gate within the dGRN is lost in *H. erythrogramma*

To investigate how these roles might differ in the derived developmental mode, we first

experimentally perturbed the function of Alx1, which is both necessary and sufficient for

skeletogenic cell fate specification in the ancestral state[36] (Fig.4a). Knocking down Alx1 protein

with a translation-blocking morpholino antisense oligonucleotide (MASO) in *H. erythrogramma*

eliminates both larval and adult skeleton (Fig.4b; Supplementary Table 1), phenocopying the

results of prior experiments in sea urchins representing the ancestral condition[36]. This

concordance suggests that the function of Alx1 in skeletogenic cell fate specification is

conserved. Zygotic transcription of *alx1*[19] and skeletogenesis[38] are both markedly delayed in *H.*

erythrogramma relative to the ancestral state, but this shift in timing does not by itself indicate a

substantive change to the organization of the dGRN.

We next examined HesC, a transcription factor that acts even earlier in the dGRN,

repressing transcription of *alx1* outside of the skeletogenic cell lineage (Extended Data Fig. 4).

Experimentally eliminating HesC protein in the ancestral dGRN produces a dramatic phenotype,

with most cells differentiating as skeletogenic because in the absence of *hesC* repression, *alx1* is

broadly transcribed[39]. In *H. erythrogramma*, however, we found that embryos develop normally

following HesC knock-down (Extended Data Fig. 4e-f), suggesting that it no longer acts as a

repressor of *alx1* transcription. This interpretation is consistent with restricted spatial expression

of *hesC* in *H. erythrogramma* (Extended Data Fig. 4a-d) that would seem to preclude a broad

repressive function for HesC outside of the skeletogenic lineage. However, these experiments

cannot rule out co-option of additional developmental functions by *hesC* in *H. erythrogramma* as

the assay presented in this study only measured this transcription factor's effect on skeletal

mesenchyme differentiation. Future work aimed at validating the loss-of-function phenotype will

confirm whether the function of *hesC* is completely lost or acquired novel regulatory roles during

H. erythrogramma development. Still, altered expression of *hesC* and lack of an overt knock-

down phenotype hint at a more profound evolution change within the dGRN.

We therefore turned to Pmar1, another transcriptional repressor that interacts with *hesC*

to form a double-negative logic gate within the dGRN[39]: throughout most of the embryo HesC

directly represses transcription of *alx1* and other genes encoding positive regulators of the

skeletogenic cell fate, permitting differentiation of other cell types; in the vegetal-most cells of

the embryo, however, *pmar1* is transiently expressed beginning the 16-cell stage where it

represses *hesC*, allowing *alx1* transcription and thus specification of the skeletogenic cell fate[40]

(Fig.4a).

Pmar1 is encoded by a cluster of tandem genes in sea urchins[41]. We identified 10 and 20

closely linked *pmar1* paralogues in *L. variegatus* and *H. tuberculata* respectively

(Supplementary Table 2). The homeodomain, nuclear localization signal, and two EH1

protein:protein interaction domains are typically well conserved, although a few likely

pseudogenes are present in each species (Fig.4c, Supplementary Fig. 4). In *H. erythrogramma*

we identified 11 *pmar1* paralogues (Supplementary Table 2). Surprisingly, all of these copies

contain numerous substitutions, deletions, and/or frameshifts, in many cases altering or

eliminating over half of the residues within the homeodomain (Fig.4c), and their expression is

barely detectable at the 16-cell and 32-cell stages (Supplementary Fig. 5). In contrast, likely

functional orthologs in the other two species differ by 0-3 amino acids out of 60 within the

homeodomain. Furthermore, pairwise similarity between *pmar1* orthologs within a species

averages greater than 88% for the entire peptide and 93% for the homeodomain in the ancestral

state, while *H. erythrogramma* averages just 71.0 % and 45.3%, respectively (Fig.4d). These

sequence comparisons indicate that the integrity of the *pmar1* gene family has dramatically

decayed in *H. erythrogramma,* raising the question whether these genes with a crucial role in

early embryonic patterning have maintained their function in the derived developmental mode.

Previous studies demonstrate that microinjecting *pmar1* mRNA into eggs produces a

dramatic phenotype, with the resulting widespread overexpression of Pmar1 protein converting

most of the embryo to skeletogenic cells[39,42]. Here, we utilized this assay to test the repressive

function of specific *pmar1* paralogues. We separately microinjected into *L. variegatus* embryos

mRNA encoding one *pmar1* paralog from *L. variegatus* and two from *H. tuberculata*. As

expected, these treatments replicated the published phenotype, inducing extensive conversion to

the skeletogenic cell fate, confirmed by widespread expression of the larval spicule matrix

protein MSP130 (Fig.4e; Supplementary Fig. 7; Supplementary Table 3). We then separately

tested the three most intact paralogues of *pmar1* from *H. erythrogramma*. At the same and higher

concentrations, none was able to produce the specific or any other discernible phenotype (Fig.4e,

Supplementary Fig. 6-7; Supplementary Table 3). These results indicate that the repressive role

of Pmar1 is retained in *H. tuberculata* but has been lost in *H. erythrogramma*.

270   Together, these perturbation experiments and sequence comparisons indicate that both

components of the double negative gate near the very top of the dGRN that specifies the

skeletogenic cell fate do not function in *H. erythrogramma* as they do in species with the

ancestral life history. Remarkably, this excision of a critical early regulatory interaction does not

abort either the specification or subsequent function of skeletogenic cells: the role of Alx1, the

component of the skeletogenic subcircuit immediately following the double-negative gate,

remains intact (Fig.4b), structural genes characteristic of differentiated skeletogenic cells are

transcribed[19], and a simplified larval skeleton is synthesized[38]. This finding is all the more

remarkable given that many other transcription factors appear to have conserved roles in *H.*

erythrogramma, based on similar expression profiles and in some cases, experimental

perturbation[19,23,43,44]. Taken together, these results reveal *H. erythrogramma* embryos as a mosaic

of conserved and altered features that have evolved as a product of its derived life history and

associated changes in selective regimes.

DISCUSSION

284   Prior work showed that the evolution of nonfeeding development in *Heliocidaris* was

285   accompanied by overt changes in oogenesis, cleavage geometry, morphogenesis, and larval

286   morphology, with extensive underlying changes in gene expression[17-20,23]. Whole genome

287   sequence analysis presented here demonstrates that these changes are not merely superficial

288 consequences of amplified maternal provisioning. Although we find evidence for adaptive

289 changes within some coding regions, these are dwarfed by the sheer number and widespread

290 distribution of apparently adaptive changes in the sequences of putative regulatory elements and

291 in the regulation of their chromatin states during early development (Fig.1c-g; 2). Both types of

292 molecular change are strikingly enriched on the branch where nonfeeding development evolved

293 and are over-represented among differentially expressed genes and especially among dGRN

294 genes (Fig.1f-g; 2c-d). While the potential for natural selection to influence trait evolution

295 through changes in gene regulation by altering regulatory element sequence and chromatin

296 accessibility is widely appreciated, we are aware of few cases that illustrate the influence of both

297 so extensively at a genomic scale and during such a short interval.

298  Focusing on transcriptional regulation that patterns the early embryo provides a test of

299 the idea that evolutionary conservation of early development is the product of intrinsic

300 constraints. We examined the earliest zygotic patterning event in the sea urchin embryo, where

301 three transcription factors interact to specify two distinct cell fates and simultaneously establish

302 the primary signaling center of the embryo. There is arguably no set of interactions within the

303 dGRN that is more fundamental to patterning the early sea urchin embryo, and they are

304 conserved among sea urchins that diverged ~225 my ago[5]. Remarkably, however, Pmar1 and

305 HesC, which interact to form a crucial double-negative logic gate[39], have lost their early

306 patterning roles in *H. erythrogramma* (Fig.4b-e, Extended Data Fig. 4). The case of *pmar1* is

307 particularly striking, as it is present as a tandem array of genes; uniquely in the genome of *H.*

308 *erythrogramma*, numerous deletions and point mutations alter about half of the homeodomain in

309 each of 10 the paralogues, rendering their proteins nonfunctional (Fig.4c-e, Supplementary

310 Fig.7).

The magnitude and extent of modifications to the earliest regulatory interactions within

the sea urchin dGRN in *H. erythrogramma* demonstrate that some deeply conserved embryonic

patterning mechanisms remain evolvable during substantial shifts in selective regimes. More

broadly, conservation of gene network architecture does not necessarily imply developmental

constraint, but may instead reflect long-term stabilizing selection for performance relative to a

particular environment or life history. Abrupt shifts in natural selection provide valuable natural

"perturbation experiments" that can reveal in detail how evolutionary mechanisms shape

conservation and change in gene regulation and dGRNs organization across the tree of life.

METHODS

1. Genome Sequencing and Assembly

1.1 Tissue Collection

Heliocidaris erythrogramma (*He*) and *H. tuberculata* (*Ht*) specimens were collected near

Sydney Harbor in Sydney, NSW, Australia and housed in natural sea water at the Sydney

Institute of Marine Science in Mosman, NSW, AU. The interpyramidal muscle of Aristotle's

lantern (the sea urchin's feeding apparatus), tube feet, and the ovarian tissue were dissected from

a single female individual, flash-frozen in liquid nitrogen, and stored at -80C until DNA

extraction and sequencing.

1.2 Genomic DNA Sequencing

For each species, a 3rd-generation DNA library was sequenced on a PacBio sequel II

CLR platform, generating 90.01 (*He*) and 89.47 (*Ht*) Gb of data with an N50 read length of

17.24 (*He*) and 23.70 (*Ht*) Kb. DNA from the same individual for each species was also used to

construct 10x Genomics linked-reads and Hi-C libraries, which were sequenced on a BGI-SEQ

500 platform, generating 194.11 (*He*) and 199.11 (*Ht*) Gb and 130.85 (*He*) and 229.03 Gb (*Ht*)

of data, respectively. Jellyfish v2.2.6[45] and GenomeScope v1.0.0[46] were deployed to conduct a k-

mer based survey of genome composition using linked-read sequencing data based on 17-mer

frequency distribution to estimate the genome size and heterozygosity of both *He* and *Ht* (Figure

S4 a-b).

1.3 Genome Assembly

PacBio sequencing data was employed to assemble a *de novo* contig-level genome

assembly using Canu v1.8 (minReadLength=1200; minOverlapLength=1000)[47]. Subsequently,

HaploMerger2 v3.6[48] was used to create breakpoints in the contigs where potential misjoins have

occurred by aligning allelic contigs via Lastz v1.02.00[49]. From these fragmented contigs, the

longest of each allelic pair was identified and selected using Redundans v0.14a[50] resulting in a

near-haploid level genome assembly. The output of this pipeline was polished using Pilon

v1.23[51] with 10X sequencing data to improve assembly quality and accuracy at single base

resolution. Lastly, contigs were assembled into scaffolds by mapping Hi-C read pairs to the

polished assembly with HiC-Pro[52], resulting in approximately 21.95% (*He*) and 32.10% (*Ht*)

valid Hi-C reads pairs. Juicer v1.5[53] and 3D-DNA v180419[54] were used to correct and finalize

the construction of chromosome-length scaffolds for each species. See Supplementary Fig. 8 for

HiC contact maps.

1.4 Repeat Identification and Classification

Genomic repetitive elements were identified with RepeatModeler v2.01[55] to generate

species-specific repeat element libraries for each. Repeat families were filtered via BLASTn

v2.3.0[56] for significant hits to gene models of the well-studied sea urchin *Strongylocentrotus*

purpuratus (www.echinobase.org) to prevent unintentional masking of genic regions. Repeats

were masked from the genome of each species with RepeatMasker v4.1.1 using the most

sensitive setting (-s) to identify the location of repetitive elements. Long-terminal repeats were

also secondarily identified with LTR_Finder v1.0.7. Outputs of both RepeatMasker and

LTR_Finder were then input into RepeatCraft v1.0[57] under default parameters to improve repeat

element annotation and identification, resulting in a final genome annotation of repetitive

elements. Lastly, repeats were broadly classified into functional categories described by their

mode of transposition using TEclass[58].

1.5 Gene Annotation and Prediction Strategy

Previously published paired-end RNA-seq reads from six developmental stages for each

Heliocidaris species[19] were trimmed using Trimmomatic v0.39[59] (TruSeq3-PE.fa: 2:30:10;

leading: 3; trailing: 3; slidingwindow: 4:15; minlen: 36) and properly paired reads were mapped

to their respective genomes using STAR v2.7.2. For each species, these RNA-seq alignments as

well as protein models of the *S. purpuratus* v5.0 genome[60] were input into BRAKER2[61] (--

etpmode). This program utilizes a number of additional software as a part of its pipeline

including Augustus[62], Genmark-EP+[63], Genemark-ET[64], DIAMOND[65], and Samtools[66]. Gene

models from the BRAKER output for filtered for transposable elements by aligning to a

combined database of transposable element sequences from the MAKER gene annotations

pipeline[67] and the Dfam v3.3 transposable element database[68] using BLAST-P[56]. Lastly, gene

models were improved using the PASA pipeline[69] by supplementing pre-existing gene models

with a de-novo transcriptome retrieved from ref. 19. These gene models were annotated by

aligning peptide sequences to three separate databases using BLAST-P v2.3.0[56] : 1) *S.*

purpuratus v4.2 gene models; 2) UniProt KnowledgeBase SwissProt protein models[70]; 3)

RefSeq invertebrate protein models with *S. purpuratus* excluded (e-value cutoff: 1e-5)[71]

(Supplementary Data 8). The list of sea urchin gene regulatory network (GRN) genes are

provided in Supplementary Data 1, retrieved from the Institute of Systems Biology

(www.biotapestry.org: accessed June 27, 2017).

2. Whole Genome Alignment

Prior to whole genome alignment, each genome was soft-masked for repetitive elements

using each species repeat element library. An optimal scoring matrix for whole genome

alignment between each set of species was inferred using the *lastz_D_Wrapper.pl* script of

HaploMerger2 v3.6[48]. Next whole genome alignment between each species pair was performed

in both directions following UCSC guidelines outlined in the *runLastzChain.sh* and

doBlastzChainNet.pl (https://github.com/ucscGenomeBrowser/kent) to produce .psl, .lav, .chain,

and finally liftOver files for each whole genome alignment. In addition, .maf files were

generated for *H. erythrogramma*, *H. tuberculata*, and *L. variegatus* for each chromosome using

H. erythrogramma as the reference genome using *Multiz* and *TBA*[72]. *L. variegatus* was chosen as

an outgroup species in this study because its genome was assembled and annotated using an

identical sequencing and bioinformatic strategy[21] as the two *Heliocidaris* species presented here,

thereby minimizing technical bias in this regard.

3. ATAC-seq

3.1 Sample Preparation

For each sea urchin species (*H. erythrogramma, H. tuberculata*, and *Lytechinus*

variegatus), adult animals were induced to spawn via injection of 0.5 M KCl solution into the

coelom. For each species, three unique male-female pairs were crossed to produce three

biologically-independent replicates of sea urchin embryos. Each culture was reared in large glass

dishes supplied with 20 mm filtered sea water (FSW) that was changed every six hours. Because

these species exhibit different developmental rates, a conspicuous developmental milestone,

shedding of the fertilization envelope at hatched blastula-stage, was selected to maximize

developmental synchrony within cultures and across species for comparison. Once a culture

reached the blastula stage, live embryos were collected and processed immediately for nuclei

preparation and transposase treatment as a part of the ATAC-seq protocol.

3.2 ATAC-seq Protocol and Sequencing

ATAC sample preparation was carried out according to the Omni-ATAC-seq protocol[73].

For each replicate, embryos were washed once in 1 mm FSW, lysed, and 50,000 nuclei were

isolated for the transposition reaction as described in the Omni-ATAC-seq protocol using the

Illumina TDE1 enzyme and tagmentation (TD) buffer (Cat. No. 20034197 and 20034198) (San

Diego, CA, USA). Sequencing libraries for each replicate were generated via qPCR and

sequencing libraries were purified and size selected using Ampure XP Beads at a 1.8:1 bead

volume: library volume (Beckman Coulter, Brea, CA, USA). Library quality and transposition

efficiency was accessed using a Fragment Analyzer and PROSize 2.0 (Agilent). *H.*

erythrogramma and *L. variegatus* libraries were sequenced on an Illumina HiSeq 4000

instrument using 50 bp SE sequencing at an average of 41.9 million and 37.3 million reads per

sample, respectively. *H. tuberculata* libraries were sequenced on an Illumina NovaSeq 6000

instrument using 50 bp PE sequencing (only SE were used for data analysis) at an average of

31.4 million reads per sample.

3.3 ATAC-seq Data Analysis

Raw ATAC-seq reads were trimmed for quality and sequencing adapters using cutadapt[74]

v2.3 with the following parameters: -a CTGTCTCTTATACACATCT -q 20 --trim-n -m 40.

Trimmed reads were then aligned to each species' respective genome using stampy[75] v.1.0.28

using the "--sensitive" set of parameters. ATAC-seq alignments were filtered for mitochondrial

sequences and required an alignment quality score of at least 5 using samtools v1.9[66].

In this study, we aimed to compare the evolution of orthologous non-coding sites. To

accomplish this, we performed a series of liftOvers[76] to convert ATAC-seq alignments between

genomic coordinates of each sea urchin species (see previous section for description of genome

alignments). We took an iterative, reciprocal liftOver strategy described below to minimize

possible reference bias associated with converting between genome assemblies: 1) *H.*

erythrogramma: *He → Lv → He*; 2) *H. tuberculata*: *Ht → Lv → He*; 3) *L. variegatus*: *Lv → He*

Lv → He. After filtering and coordinate conversion, all ATAC-seq alignments were referenced

to the *H. erythrogramma* genome with an average of 5.9 million alignments per sample to

orthologous genomic loci.

Following filtering and coordinate conversion, peaks were called from these alignments

using the MACS2 v2.1.2[77] callpeak function (parameters: –nomodel, –keep-dup=auto, –shift

100, –extsize 200) for each species separately. Peak coordinates were merged using the

bedtools[78] v2.25 *merge* function requiring a peak overlap of at least 200 bp to be merged into a

single peak. Lastly, for each sample, accessibility of each peak was measured with the bedtools[78]

v2.25 *multiBamCov* function.

3.4 Tests for positive selection within OCRs

In order to test for evidence of positive selection, a neutral genomic reference across all

species was assembled. To do this, the genome was first masked repetitive elements, coding

sequence, untranslated genic sequence, non-coding RNAs (including microRNAs, ribosomal

RNAs, small nuclear RNAs, and transfer RNAs) and ATAC-seq open chromatin regions (OCRs)

(see below) in the genome. The remaining, putatively neutrally evolving genome was then

divided into 300 bp windows, orthologous regions retrieved from each species' genome, and

filtered using the *filtering.py* and *pruning.py* scripts of the "adaptiphy"[79] program

(https://github.com/wodanaz/adaptiPhy). Next, branch lengths of each of these neutral sites was

estimated using phyloFit[80] (--subst-mod HKY85), highly-conserved sites removed (Figure S5),

relative branch lengths calculated, and sites falling within the middle 50% of relative branch

lengths in the *H. erythrogramma* genome were selected as the neutral reference (88,004 sites;

Supplementary Data 2, Supplementary Fig. 9).

To measure branch-specific signatures of positive selection in the non-coding genome,

the adaptiPhy[79] pipeline (https://github.com/wodanaz/adaptiPhy) for global tests of natural

selection was followed. First, orthologous sequences for non-coding sites of interest were

selected from each species' genome into FASTA format. Sequences were trimmed to include

only contiguous DNA sequence using the *prunning.py* script and filtered using the *filtering.py*

script, requiring a minimum alignment length of 75 bases. These trimmed and filtered alignments

serve as "query" sequences of tests for selection. To generate a neutral reference for

comparison, ten neutral sites were randomly selected (see above) and concatenated into a single

neutral reference sequence. In addition, for each OCR, tests for positive selection were repeated

10 times against a unique putatively neutral reference. For each query site replicate, substitution

rates of both the query and randomly concatenated neutral reference were estimated using

phyloFit[80], and the zeta score was calculated as the ratio of the query substitution rate to the

neutral reference substitution rate. In addition, p-values of likelihood ratio tests for significant

levels of branch-specific positive selection were calculated with adaptiPhy[79] pipeline using

HyPhy[81]. P-values and substitution rates for all query and neutral sites were then imported in R

v4.0.2 for analysis (Supplementary Data 3).

3.5 ATAC-seq Peak Filtering

After accessibility and rates of selection were calculated for each ATAC-seq peak, herein

referred to as an "open chromatin region" (OCR), a series of filtering and quality control metrics

were carried out to ensure only high confidence and quality peaks were compared between

species. These filtering steps are as follows: 1) each OCR is required to have at least 75bp of

contiguous, single copy sequence (see *Section 3.4*) for accurate estimations of selection; 2) for

each species, a local composition complexity (LCC)[82] value of 1.9 or more was required for the

OCR to remove repetitive or other low-complexity sequences that may generate inaccurate

estimations of selection (module: biopython.org/docs/1.75/api/Bio.SeqUtils.lcc.html); 3) a CPM

of 3 or more was required in at least 2 (of the 9) samples to remove OCRs with extremely low

accessibility; 4) the midpoint of the OCR must lie within 25kb (in either direction) of the

translational start site of a gene model; 5) the gene nearest to an OCR must be the same gene in

each of the species' genomes —in other words, for each OCR and its nearest gene in the *H.*

erythrogramma, the orthologous region in the *H. tuberculata* and *L. variegatus* genome must

also be closest to a gene model that is orthologous (determined by annotation) to the same gene

in the *H. erythrogramma* genome. Given nearly no prior knowledge is known of the cis-

regulatory landscape for these sea urchin species, these stringent filtering methods were carried

out in order to maximize confidence in comparisons of non-coding sequence evolution and

function. This method resulted in a final set of 27,322 high-confidence OCRs for cross-species

analysis (Supplementary Data 4).

3.6 ATAC-seq Statistical Analysis

Raw counts of the filtered OCRs were loaded into DESeq2[83] v1.30 to calculate

differential accessibility between sample groups. For life history strategy comparisons, *H.*

tuberculata and *L. variegatus* were treated as a single group. Differentially accessible sites were

classified as having a 2-fold accessibility difference between sample groups and supported by an

FDR of 10%. Significant levels of positive selection were classified as having a median zeta

value greater than 1.5 and supported by a median false-discovery rate less than 10% across 10

replicates for each query site. Branch-specific evidence of positive selection met these criteria for

one species, but failed to meet these criteria in the other, as evidenced by a zeta score < 1.5.

4. Coding Selection Analyses

In order to make tests for positive selection in coding sequences analogous to non-coding

sequences, only single-copy orthogroups were considered in these analyses. Single-copy

orthologs between *H. erythrogramma*, *H. tuberculata*, *L. variegatus*, and *Echinometra lucunter*

were identified using OrthoFinder[84]. Evidence of episodic positive selection was queried on both

the *H. erythrogramma* branch and *H. tuberculata* branch under default parameters using

BUSTED[22], by specifying either branch as the "foreground" branch. P-values from these

analyses are available in Supplementary Data 5. Genes with significant evidence of episodic

positive selection were supported by a p-value <= 10% by the likelihood ratio test.

5. Bulk RNA-sequencing Analysis

Raw RNA-seq reads from blastula stage embryos of *He*, *Ht*, and *Lv* were retrieved from

ref 19, trimmed and filtered for low quality bases and reads with Trimmomatic[59], and aligned to

each species respective genomes and gene models with STAR[85]. From these alignments, mRNA

expression was estimated with Salmon[86] and loaded to R for statistical analysis. Read counts for

summed to each gene's best match to the *S. purpuratus* v4.2 gene models to generate a common

reference for expression comparisons between species as described in ref 19. Differentially

expressed genes between life histories were called as having a fold-change in expression > 2 and

supported by an FDR of 10% or less between *He* and both planktotrophic species (in the same

direction), and not DE between *Ht* and *Lv* (Supplementary Data 6).

6. Single-cell RNA-Sequencing

6.1 H. erythrogramma embryo culturing

Female H. erythrogramma individuals were spawned via intracoelomic injection of 0.5ml

of 0.5M KCl. Unfertilized eggs were washed three times in 100um filtered natural sea water

(FSW). Eggs were fertilized by 2ul of concentrated sperm in .02g Para-Amino Benzoic Acid

(PABA)/100 ml FSW. Following fertilization, eggs were washed three additional times in FSW

to remove residual sperm and PABA. Fertilized embryos were then cultured at 22-23 °C. At 6

hours post fertilization (hpf) embryos were sampled for microscopy and dissociation, then fixed

for scRNA-seq.

6.2 Embryo Dissociation and Fixation

Once embryos developed to the early blastula stage (pre-skeletogenic cell ingression), a

portion of the co-culture was taken, and washed one time in Calcium-Free Artificial SeaWater

(CFASW). After washing embryos with CFASW, 3ml of embryos were added to 7ml of

dissociation buffer made (1.0M Glycine and 0.25mM EDTA, pH 8.0) at 4 °C, and gently rocked

on a rocker for 10 minutes. Following incubation, embryos were gently triturated 15-20 times to

increase disassociation, then 10ml ice cold 100% methanol was added, and cells were incubated

for 10 minutes and on the rocker. Following incubation, cells were triturated again 15-20x times,

and then another 30ml of ice cold 100% methanol was added to bring the suspension to a final

concentration of 80% methanol. This 80% methanol suspension of cells was incubated at 4 °C

for 1 hour. Following this last fixation step, cells were stored at -20 °C until library preparation.

6.3 Rehydration of Single Cells, Library Preparation, and Sequencing

Cells were centrifuged at 50xg, supernatant was discarded, and fixed cells were washed

twice and rehydrated in a Sigma 3x Saline Sodium Citrate (SSC) buffer (SKU SRE0068) before

cell count and library preparation. Cell concentration was estimated with a hemocytometer and

volume was adjusted to a final concentration of ~300 cells/ uL. Single cell libraries were

prepared using the 10x Genomics 3' v3 gene expression kit and the 10x Chromium platform to

encapsulate single cells within droplets. Library quality was verified using the Agilent 2100

Bioanalyzer. In total, ~ 3960 cells were loaded onto the 10X instrument. For the single library

preparation, 2500 cells were targeted, of which 2066 were successfully captured for sequencing.

Libraries were sequenced by the Duke Genomics and Computational Biology Core facility on

two NovaSeq6000 S1 flow cells with 28 x 8 x 91 bp sequencing performed.

6.4 FastQ Generation, Genome Indexing, and Quantification of scRNA-seq

Following sequencing, we used Cellranger v3.1.0 to convert Illumina-generated BCL

files to fastq files using the Cellranger "mkfastq" command. scRNA-seq data for early blastula

stage embryos (pre-skeletogenic cell ingression) of L*ytechinus variegatus* were retrieved from a

published scRNA-seq developmental time course of the species[34]. We then applied the "mkref"

command to index the most recent Lv3.0 genome[21] (for the *Lytechinus* data) and the *H.*

erythrogramma genome assembled in this study. The "count" command was used to demultiplex

and count reads mapping to the respective reference *Lv* (53.3% mapping rate) or *He* (94.1%

mapping rate) genome. The "mat2csv" command was used to obtain CSV RNA count matrix

files for each sample for further downstream analysis. Orthofinder v2.3.12[84] was implemented to

identify putative 1-1 orthologous gene models between *Lv* and *He* (Supplementary Data 10).

6.5 scRNA computational analyses

We employed a dual strategy for comparing scRNA-seq expression between *He* and *Lv*:

1) A non-integrated analysis in which scRNA-seq from each species was quantified against its

own gene models (Fig. 3a) and 2) an integrated analysis in which orthologous anchor genes were

used to identify cell types with overlapping expression profiles between each species (Fig. 3d).

CSV RNA count matrix files were uploaded to R and a Seurat object was generated for 1) each

species separately quantified against their own gene models and 2) orthologous genes between

H. erythrogramma and *L. variegatus* (Supplementary Data 7). Each dataset was filtered to

remove low quality cells with nFeature_RNA > 200, nFeature_RNA < 5500, and nCount_RNA

< 7500. To ensure differences in input cell number did not bias detection of different cell types

between samples, the *Lv* dataset was separately subsampled to 2065 randomly-selected cells and

clustering analyses were repeated (see Extended Data Fig. 2). Furthermore, both datasets had

comparable distributions of genes per cell and UMIs per cell numbers (Supplementary Fig. 2).

For the non-integrated analyses, "SCTransform", a regularized negative binomial

regression method that stabilizes variance across samples, was applied to perform normalization

and removal of technical variation[87], while preserving biological variation. We next performed

Principal Component Analysis on the SCTransformed Seurat object file of raw gene expression

counts, and found the nearest neighbors using 10 PC dimensions of variable gene space. UMAP

(Uniform Manifold Approximation and Projection)[88] was applied to multi-dimensional scRNA-

seq data to visualize the cells in a two-dimensional space. Finally, clustering was performed

using graph-based Louvain Clustering with resolution, res=0.5, resulting in 7 clusters in *L.*

variegatus and 3 clusters in *H. erythrogramma*. The clusters were putatively annotated using

dGRN genes and published in situ hybridization patterns as markers (Supplementary Fig. 1,

Supplementary Data 11), and ambiguous cluster identities are conservatively denoted as broad

embryological territories (e.g. endomesoderm, ectoderm A-C) or as "unknown".

To perform the integrated analyses of scRNA-seq data between species, we carried out 3

independent methods for scRNA-seq data integration to compare expression of the same

orthologous genes between *H. erythrogramma* and *L. variegatus* and identify putative

overlapping cell types: 1) Canonical Correlation Analysis (CCA)[89], 2) Reciprocal Principal

Component Analysis (RPCA) (satijalab.org/seurat/articles/integration_rpca.html), and 3)

Harmony[90] (Fig. 3d; Extended Data Fig. 3). For each scRNA-seq strategy (non-integrated and

CCA), counts for all gene models and orthologous sets of genes are included as Supplementary

Data 7, and top marker genes for each cluster are available in Supplementary Data 9.

7. *H. erythrogramma* microinjection and in-situ hybridization

7.1 MASO design and microinjection

Morpholino antisense oligos (MASOs) were constructed to target the translation start site

of *alx1* (ATCAATTCGGAGTTAAGTCTCGGCA) and *hesC*

(ATCCAGATGTGTTAAGCATGGTTGC) and synthesized by Gene Tools (Philomath, OR,

USA). Control morpholinos included a standard negative control morpholino recommended by

the manufacturer (CCTCTTACCTCAGTTACAATTTATA) and a scrambled morpholino for

HesC (ATCGACATCTGTTAACCATCGTTGC). Fertilized eggs of *H. erythrogramma* were

injected as described in ref. 91 at a concentration of 100 μM and 200 μM for *alx1* and 500 μM

for *hesC*, then reared at 22 °C in pasteurized, 0.22-micron filtered seawater + penicillin (100

unit/mL) and streptomycin sulfate (0.1 mg/mL) (Sigma P4333A). Injected embryos were

checked for developmental abnormalities and mortality every six hours until fixation.

7.2 Fixation and ISH

H. erythrogramma embryos were fixed for in-situ hybridization (ISH) for ~16 hrs

overnight at 4 °C in 4% paraformaldehyde (Sigma 158127) + 20 mM EPPS (Sigma E1894) in

FSW, washed 3x in FSW, and dehydrated step-wise into 100% MeOH and stored at -20 °C. The

full-length *HesC* coding sequence was synthesized in vitro by GenScript (Piscataway, NJ, USA)

and subcloned (NCBI insert number: MK749159), and used as template to make antisense RNA

probes for ISH. ISH of *H. erythrogramma* was performed according to previously published

methods[44]. Hybridizations were carried out at 65 °C and stringency washed at 0.1% SSC.

7.3 Imaging

Fixed *H. erythrogramma* embryos were washed with 100% EtOH, cleared and mounted

in 2:1 (v/v) benzyl benzoate: benzyl alcohol (BB:BA). DIC or polarized light (PL) micrographs

were taken on Olympus BX60 upright microscope with an Olympus DP73 camera. ISH images

were taken on a Zeiss Upright AxioImager with a Zeiss MRm camera using ZEN Pro 2012

software.

8. Pmar1 mRNA overexpression assays

8.1 mRNA Synthesis

Sequences of *pmar1* orthologs were retrieved from each species respective genome

annotations (*Lv*[21]*, He* and *Ht*, this study) (Supplementary Table 2). One *Lv* ortholog

(LVA_25833.t1) was selected for overexpression assays as it represents the ortholog tested in

previous overexpression assays[42], while two *Ht* (HTU_11636.t1 and HTU_11625.t1) and three

He (HER_770.t1, HER_761.t1, and HER_775.t1) were selected for overexpression assays as

they represent orthologs with the highest identity to the species' consensus sequence, and

therefore predicted as genes most likely to be functional. Construct templates for each ortholog

were ordered from Twist Biosciences (San Francisco, CA, USA) and mRNA was synthesized

from these constructs with a ThermoFisher MEGAshortscript T7 Transcript Kit (AM1354).

8.2 Overexpression experiments

Female *L. variegatus* individuals were spawned via intracoelomic injection of 0.5M KCl,

washed in FSW, and fertilized with 1 uL of concentrated sperm in 100 ml FSW. For each

construct, at least 4 rounds of microinjections were conducted with each round including 30-50

healthy embryos. *Lv* constructs were injected at a concentration of 250ng/µL. *Ht* constructs were

injected at a concentration of 1200ng/µL, and *He* constructs were injected at 1500ng/µL. Higher

concentrations of *Heliocidaris* constructs were used to reproducibly obtain the skeletal cell

conversion phenotype, and reduced sensitivity of these assays may be attributable to less optimal

cross-species interactions of Pmar1 in regulating the *L. variegatus* genome. Following injection,

embryos were incubated at 23 °C and imaged at 24 hpf.

8.3 Imaging and Immunostaining

At 24 hpf, live embryos from each experiment were mounted on slides and imaged using

DIC microscopy. Embryos were imaged on a Zeiss Axioplan II upright microscope controlled by

Zen software. Also at 24 hpf selected embryos were fixed in 100% ice cold methanol.

Immunostaining was carried out as described in ref. 92 to mark expression of Msp130 protein.

Blocking and incubation of the secondary antibody was increased to 1 hr. Incubation of the

primary antibody was set to 48 hours. A Zeiss 880 inverted confocal Airyscan microscope

controlled by Zen software was used to take Z-stack images of stained embryos. Pmar1

overexpression results are summarized in Supplementary Table 3.

9. Data Availability

9.1 Genomes

Sequencing reads used to assemble the *Heliocidaris* genomes and the genome assemblies

themselves are available on the Chinese National GeneBank (CNP0002233). Genome assemblies

of *Heliocidaris erythrogramma* and *Heliocidaris tuberculata* are also available on NCBI

(PRJNA827916 and PRJNA827769, respectively). Genome assembly of *Lytechinus variegatus* is

previously published[21] and available on NCBI (PRJNA657258). Genome annotations and files

associated with whole genome alignments between species are available on Dryad

(https://doi.org/10.5061/dryad.sj3tx966v).

9.2 ATAC-seq

Raw sequencing reads for the ATAC-seq dataset are available on NCBI (PRJNA828607).

Alignment files are available on Dryad (https://doi.org/10.5061/dryad.sj3tx966v). Result files

associated with the ATAC-seq analyses are available as Supplementary Data 2-4.

9.3 Bulk RNA-seq

Bulk RNA-seq data was retrieved from Israel et al, 2016[19].

9.4 Single cell RNA-seq

Raw sequencing reads for the *Heliocidaris erythrogramma* single cell ATAC-seq dataset are

available on NCBI (PRJNA833141). Sequencing reads from the *Lytechinus variegatus* single

cell RNA-seq dataset were retrieved from Massri et al, 2021[34] and available on NCBI

(PRJNA765003). Results files associated with the single cell RNA-seq analyses are available as

Supplementary Data 7,9.

AUTHOR CONTRIBUTIONS

LW, MB, GF, and GAW conceived and designed the study. LW, DK, and PC collected tissues

for genomic sequencing. YZ performed genomic DNA extraction and sequencing library

preparations. PLD, HG, and HZ assembled the genomes. PLD performed genome annotation,

alignments, and data analysis. PLD collected, prepared, and analyzed ATAC-seq libraries. AJM

collected, prepared, and analyzed single cell RNA-seq libraries. JSS and AE performed

embryonic expression and injection assays. PLD and GAW wrote the manuscript and all authors

contributed to manuscript revisions.

COMPETING INTERESTS

There are no competing interests to declare.

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MATERIALS & CORRESPONDENCE.

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FIGURE LEGENDS

Figure 1. Evolution of life history and genomes. a, Feeding larval development

(planktotrophy: green) represents the ancient and ancestral life history in sea urchins (seastar *P.*

miniata represents the outgroup)[11]. Non-feeding larval development (lecithotrophy: orange) has

evolved on multiple occasions, including recently within the genus *Heliocidaris* (arrowhead)[9]. **b,**

Evolution of non-feeding development in *H. erythrogramma* (bottom) included dramatic

modifications to otherwise broadly conserved developmental mechanisms, including changes in

cleavage geometry, cell fate specification, and morphogenesis[16,18]. **c,** Chromosome-scale genome

assemblies of *H. tuberculata* (green) and *H. erythrogramma* (orange). Outer ring: repetitive

element content; middle ring: gene content, inner ring: zeta values in OCRs from the selection

analyses. Colored points indicate statistically elevated zeta values (indicative of positive

selection) within a single OCR on the branch leading to *H. tuberculata* (green) or *H.*

erythrogramma (orange). Blue points indicate highly conserved OCRs (top 10% of phastCon

scores). Triangles denote OCRs with signature of branch-specific positive selection located near

dGRN genes. Synteny lines between chromosomes denote locations of 1-1 orthologous OCRs

between *Heliocidaris* species. **d,** Signatures of positive selection in protein-coding sequences

(CDS) of 84 dGRN and 3,832 non-dGRN single-copy orthologs. Evidence of selection is slightly

enriched on the *H. erythrogramma* branch, but dGRN genes show no difference between

branches. **e,** P-values of likelihood ratio test for positive selection in CDS on the branch leading

to each species (color indicates significant p-values; squares indicate dGRN genes). **f,** Signatures

of positive selection within single-copy OCRs near dGRN (n = 1069) and non-dGRN (n =

26,253) genes are overall much higher on the *H. erythrogramma* branch (a single gene can have

multiple OCRs associated with it). For OCRs near dGRN genes, this difference is notably

amplified: signatures of positive selection are depleted relative to non-dGRN genes on the *H.*

tuberculata branch but substantially elevated on the *H. erythrogramma* branch. **g,** P-values of

likelihood ratio test for positive selection in OCRs on the branch leading to each species. Fisher

exact test, two-sided- *: p-value < 5e-2; ***: p-value < 5e-4. OCR: open chromatin region;

dGRN: developmental gene regulatory network; CDS: coding sequence.

Figure 2. Evolution of open chromatin landscape. a, Density and volcano plots of

significantly differentially accessible orthologous OCRs between developmental life histories

(green = more open in *H. tuberculata* and *L. variegatus*, orange = more open in *H.*

erythrogramma). **b,** Examples of conservation (*foxA*) and change (*hesC*) in chromatin

accessibility landscape near dGRN genes. See Extended Data Fig. 1 for an additional example of

a conserved chromatin landscape nearby *foxN2/3* and corresponding expression domains at

blastula stage of both *Heliocidaris* species. **c,** Relationship between chromatin, positive

selection, and gene expression. Percentage of all genes (top) and differentially expressed genes

(bottom) with at least one OCR nearby that is differentially accessible (DA), has evidence of

positive selection on the *H. erythrogramma* (*H.e.*) or *H. tuberculata* (*H.t.*)-branch, or is both DA

and has evidence of positive selection on the *H.e.* or *H.t.* branch. **d,** Number of differentially

accessible OCRs and OCRs with evidence of positive selection in *H. erythrogramma* for a given

gene. Fisher exact test, two-sided-*: p-value < 5e-2; *** p-value < 5e-4. OCR: open chromatin

region; dGRN: developmental gene regulatory network.

Figure 3. Evolution of transcriptomes. Single cell RNA sequencing of pre-mesenchyme

blastula stage embryos from **a.** *L. variegatus* and **b.** *H. erythrogramma* with cells plotted into a

UMAP space prior to integration of expression data across species. Colors indicate inferred cell

lineages based on expression of marker genes (Supplementary Fig. 1). Note that *H.*

erythrogramma shows fewer distinct transcriptional states than *L. variegatus*. **c,** Composite

expression scores for four embryonic territories based on combined expression of multiple

known marker genes in the ancestral GRN (Supplementary Data 11). These composite

expression markers are localized to discrete domains in *L. variegatus* but not in *H.*

erythrogramma, suggesting that major embryonic cell lineages have yet to differentiate

transcriptionally in this species at this stage of development. **d**, Integrated UMAP of cells from

both species, color-coded by species (left) and by cluster (right), and showing both species (top),

and separated by species (middle and bottom). *L. variegatus* clusters remain well separated

(center right) but *H. erythrogramma* clusters broadly overlap with each other (lower right). As

shown in the 2X magnifications, *L. variegatus* clusters contain primarily cells of one color, while

cells from all three *H. erythrogramma* clusters are present in appreciable numbers in the same

UMAP space (transparency is set to 40% in the insets to circumvent masking; example inset

clusters are *L. variegatus* skeletogenic cells, but the same holds for other *L. variegatus* clusters).

Figure 4. Evolutionary change at the "top" of a conserved developmental dGRN. a,

Schematic of the ancestral dGRN that specifies skeletogenic cell fate: HesC suppresses this fate

in most of the embryo (blue) but Pmar1 suppresses *hesC* in the precursors of the skeletogenic

cells (magenta), where Alx1 then activates a differentiation program. **b,** Images of control and

MASO knock-down of Alx1 in *H. erythrogramma* (early larva; scale bar 100μm). Polarized light

(PL) illuminates skeletal elements (most are anlage of the adult skeleton, with a longer larval

"arm" element out of focus on the left [blue arrow]). Bar chart of *alx1* injection summary

statistics in *H. erythrogramma* (see also Supplementary Table 1). Knockdown of *alx1* expression

eliminates skeleton formation in *H. erythrogramma*, as in the ancestral dGRN[36]. **c,** Alignment of

homeodomain (DNA-binding) and RP domains (protein:protein interaction) from *pmar1*

paralogues. Green = likely functional copy, red = predicted non-functional copy, asterisks =

paralogues whose function was experimentally validated. **d,** Within-species pairwise sequence

similarity of *pmar1* paralogues. Note rapid sequence divergence among paralogues in *H.*

erythrogramma, and particularly within the homeodomain. Center line = median pairwise

similarity; dots = outliers **e,** Overexpression assays of control and *pmar1* mRNA (prism stage;

skeletogenic cells labeled with antibody that recognizes cell surface protein MSP130). DIC and

fluorescent images demonstrate that mRNA of *pmar1* paralogues from *L. variegatus* and *H.*

tuberculata convert most of the embryo to skeletogenic cells, whereas even the most intact *H.*

erythrogramma paralogues show no such phenotype, indicating loss of function (see

Supplementary Fig. 7 for additional antibody staining). dGRN: developmental gene regulatory

network; MASO: morpholino antisense oligonucleotide; DIC: differential interference contrast;

RP: repeated peptides.

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Extended Data Figure 1. Chromatin landscape and expression domain of *foxN2/3* is

conserved. a, Chromatin accessibility nearby *foxN2/3* in *H. erythrogramma* (top, orange) and *H.*

tuberculata (bottom, green), including seven open chromatin regions (OCRs). *In-situ*

hybridization of *foxN2/3* in blastula-stage **b,** *H. erythrogramma* and **c,** *H. tuberculata* embryos.

Extended Data Figure 2: Scaling down number of *L. variegatus* cells does not significantly

affect clustering. UMAP of non-integrated single cell RNA-seq data from *L. variegatus*, in

which the data has been randomly subsampled to 2065 cells so that the cell number is equivalent

to the *H. erythrogramma* dataset. This subsampling does not change the number and general

spatial relationship among clusters in *L. variegatus* (i.e. distinct cluster of skeletogenic cells

separate from remainder of cells in the embryo).

Extended Data Figure 3: Three independent methods of integrating single cell RNA-seq

data recover similar clustering relationships among *L. variegatus* and *H. erythrogramma*

cells. Canonical correlation analysis (CCA), reciprocal principal component analysis (RPCA),

and Harmony each recover more numerous, distinct transcriptional states in *L. variegatus*

relative to *H. erythrogramma* following integration of single cell expression data. All three

methods reveal a consistent number of clusters in both species. In particular, *H. erythrogramma*

contains fewer clusters there is more extensive overlap of cells among clusters. These results

suggesting that differentiation of discrete cell populations is delayed in the *H. erythrogramma*

embryo.

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Extended Data Figure 4: *hesC* appears to have lost its ancestral role of repressing larval

skeletal cell specification in *H. erythrogramma*. Derived expression patterns of *hesC* in *H.*

erythrogramma at **a,** cleavage; **b,** blastula; **and c-d,** larva stage embryos. **e-f,** Control and MASO

knock-down of HesC in *H. erythrogramma* (early larva; scale bar 100μm). Polarized light

illuminates skeletal elements. HesC knockdown appears to show no phenotype, a dramatic

change from the ancestral dGRN.