Dauer fate in a *Caenorhabditis elegans* Boolean network model

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Abstract

Cellular fates are determined by genes interacting across large, complex biological networks. A critical

question is how to identify causal relationships spanning distinct signaling pathways and underlying

organismal phenotypes. Here, we address this question by constructing a Boolean model of a well-

studied developmental network and analyzing information flows through the system. Depending on

environmental signals *Caenorhabditis elegans* develop normally to sexual maturity or enter a

reproductively delayed, developmentally quiescent 'dauer' state, progressing to maturity when the

environment changes. The developmental network that starts with environmental signal and ends in the

dauer/no dauer fate involves genes across 4 signaling pathways including cyclic GMP, Insulin/IGF-1, TGF-

β and steroid hormone synthesis. We identified three stable motifs leading to normal development,

each composed of genes interacting across the Insulin/IGF-1, TGF-Beta and steroid hormone synthesis

pathways. Three genes known to influence dauer fate, *daf-2*, *daf-7* and *hsf-1*, acted as driver nodes in

the system. Using causal logic analysis, we identified a five gene cyclic subgraph integrating the

information flow from environmental signal to dauer fate. Perturbation analysis showed that a

multifactorial insulin profile determined the stable motifs the system entered and interacted with *daf-*

12 as the switchpoint driving the dauer/no dauer fate. Our results show that complex organismal

systems can be distilled into abstract representations that permit full characterization of the causal

relationships driving developmental fates. Analyzing organismal systems from this perspective of logic

and function has important implications for studies examining the evolution and conservation of

signaling pathways.

Introduction

Cellular states, developmental pathways and organismal phenotypes are determined by interactions

between biomolecules. These biological networks are large and complex, involving genes, proteins and

regulatory molecules spanning diverse pathways and levels of biological organization [1]. Data-based

network inference often implicates large numbers of biomolecules through correlation-based methods

[2] but network modeling and analysis indicate causal relationships are the key to controlling

information flow and determining cellular fates [3]. A fundamental question is how to identify the

critical, causal relationships within complex biological systems and between distant nodes in a network

[4]. Here, we construct a Boolean model of a well-studied model system and use network analysis

techniques to identify the core interactions responsible for the phenotype.

C. elegans is a small (~1mm), free-living nematode worm that has been used as a model organism in

biological research for over 50 years [5]. Populations are androdioecious, composed of self-fertile

hermaphrodites and a small number of males, resulting in straightforward laboratory culture and rapid

propagation. *C. elegans* are transparent, facilitating microscope study, and eutelic, meaning that all

adults have the same number of somatic cells. In *C. elegans* hermaphrodites there are 959 somatic cells

while males have 1031 [6]. The entire life cycle can be completed within 3 days and these factors have

led to the establishment of *C. elegans* as an important model organism for studying development,

genetics and essential biological processes. *C. elegans* was the first metazoan to have its genome fully

sequenced [7, 8] and the long history of *C. elegans* research has provided a wealth of genetic, genomic

and developmental understanding of this small worm [9-12].

One of the most studied developmental processes in *C. elegans* is dauer [13, 14]. The small worms are

vulnerable to environmental conditions including lack of food or moisture [15] and high temperature

[16]. *C. elegans* senses its environment through chemoreceptors that perceive biomolecules including

pheromones produced by other worms [17]. In low-stress environments the worms progress through

discrete larval stages L1-L4, to adulthood and reproductive maturity [13]. Environmental triggers can

activate a developmental switch at the L1 stage and lead to dauer arrest [18]. Dauer worms can move

but are developmentally quiescent and survive without feeding for periods of up to 4 months.

C. elegans can be forced into the dauer fate with high pheromone concentrations, high temperature or

starvation in the lab but the signals can also interact quantitatively in the worm's natural environment

[17]. The dauer pheromone itself is a complex mix of ascarosides [19, 20] with natural variation in

production, composition and response between populations [21], and species [22]. *C. elegans* strains

have substantial, quantitative natural variation for dauer entry [23], [24] and exit [25] but for the

individual worm dauer is an irreversible developmental decision. Undergoing dauer has lasting effects

on worm reproduction [26], physiology [27] and locomotion [28].

Dauer is a unique, discrete developmental switch and the molecular basis has been studied extensively.

Over 100 genes across 4 signaling pathways (cyclic GMP, Insulin/IGF-1, TGF-β and steroid hormone

synthesis) have been implicated as influencing dauer [13, 29]. Despite extensive descriptive

characterization, we lack a robust understanding of the interplay between these pathways and the

causal links that drive the system to one fate or another. Here, we study a Boolean model and identify

the critical interactions determining dauer fate.

Simplified representations of complex biological networks can accurately distill systems down to

essential interactions [30]. Boolean networks are graphs where nodes represent genes and edges are

regulatory interactions defined as Boolean functions or logic gates, for example the operators 'AND' or

'OR' [31]. Boolean models of biological processes represent states as discrete 'on' or 'off' [32] and

accurately represent the switch-like nature of dauer induction. Regulatory interactions are threshold-

gated, qualitative relationships. Here, we use our Boolean network to emulate cellular behavior

observed in experimental settings and systematically perturb the system to discover causal links within

the larger complex network. We identify the critical interactions and combinations of gene states that

determine dauer fate. Evolutionary searches of pathway conservation and divergence typically examine

gene-by-gene, sequence level homology [33]. Here, we suggest that causal logic and function may be

the true phenotypes underlying the evolution of signaling pathways.

Methods

Constructing the dauer network

We first extracted 85 dauer-involved genes described in the WormBook 'Dauer' chapter [13]. For each of

these genes, we then searched on WormBase [53] for genes known to interact with these dauer genes

and genes known to act in the same pathways as these dauer genes. We focused on regulatory

interactions and included those for which we identified information regarding the direction of the

interaction, for example reported relationships between effector proteins and targets. We did not

include genetic interactions or genetic masking/enhancing/suppressing effects between genes. Some of

our gene pairs may have reported genetic interactions along with regulatory interactions but we focused

on regulatory interactions for constructing our model.

We next surveyed the literature for regulatory interactions within this gene and pathway set and

created a filtered set of dauer genes and main-effect regulatory interactions (Supplementary Table 1).

For example, of the 40 insulin signaling genes in the *C. elegans* genome few have been reported to act

independently in the Insulin/IGF-1 pathway [5]. We condensed the larger dauer-involved set down to 26

genes, each part of the four dauer-inducing pathways represented as nodes in our Boolean model, and

connected by 34 edges. We included the nodes 'pheromone' [34] and '*cmk-1*' (a food receptor [35]) as

inputs to the system. These interactions were mainly obtained from the WormBase database [53] and

verified with the scientific publications cited on WormBase (a full list of nodes, interactions, and

citations is given in Supplementary Table 1).

We used this regulatory interaction information for each gene to formulate Boolean functions

(Supplementary Table 2). We combined literature-reported regulations into a preliminary model and

used results obtained from simulations and perturbations to fine-tune the Boolean functions for each

node until we were able to mimic the expected behavior of the system. For example, multiple articles

reported that *daf-16* was activated either when *akt* was repressed [54-57] or when *daf-12* was activated

[58, 59]. Simulated conditions for normal adult development should therefore result in *daf-16*

repression and no dauer activation. To achieve this known output for *daf-16* we assigned an 'OR'

operator for these two regulators. In some cases, we were able to extract the relationship between

regulatory nodes from the literature. This approach has been similarly demonstrated in previous works

to gain qualitative insights using discrete networks [60, 61].

Studying system trajectories

The initial conditions used to generate the trajectories of the system's dynamics were based on well-

known environmental cues such as food availability (*cmk-1*) and dauer pheromone (pher) detected by *C.*

elegans [17, 35]. Favorable conditions allowed the organism to develop into an adult ("no dauer") and

unfavorable conditions into "dauer" [34]. We set the initial conditions as *cmk-1* = 1 and pher = 0 for

normal development (here, dauer = 0) and *cmk-1* = 0 and pher = 1 for dauer fate (here, dauer = 1). We

monitored the dynamics and convergence of the system based on the state of the 'dauer' output node

(trajectories for each simulation are shown in Supplementary Figure 2). Temperature is another well-

known regulator that induces dauer formation. We chose to focus here on dauer as a response to

resource availability and our model does not include temperature as an input.

We performed simulations with asynchronous updating using boolean2 software [62]. In order to

explore system dynamics, the Boolean functions for each gene needs to be updated at each timestep to

trace out the trajectory of the system. These transitions for the Boolean functions were carried out

using an asynchronous scheme that randomly assigned the states of the nodes. Biological events do not

occur at the same time and the timescale at which the events occur can be diverse. Here, we

represented this with an asynchronous update scenario. In this framework a collector class averaged the

states of nodes over multiple runs for simulating a system with a non-deterministic updating scheme.

We performed 100 simulations for each set of initial conditions and for every iteration, the collector

averaged the state of each node generated for 1000 steps. In all our simulations, the nodes reached

equilibrium after approximately 15 iterations (trajectories are shown in Supplementary Figure 2). Nodes

with states > 0.5 were considered 'ON' (as indicated by [62]) and nodes with states <0.5 were

considered 'OFF'.

Validating the model

To validate the model, we checked whether the model has the ability to reproduce known experimental

outcomes. Mutant phenotypes in *C. elegans* that lead to either adult or dauer formation are well

studied [13]. We used published experimental evidence to check whether our model could reproduce

similar results. For example, dauer-constitutive (Daf-c) mutants lead to dauer formation in conditions

that do not favor dauer while Dauer-defective (Daf-d) mutants will not develop into dauer larvae [18,

40-42]. We knocked down known Daf-c and Daf-d genes to validate that our dauer Boolean model

captured essential wild type and mutant phenotypes (Supplementary Figure 1b).

Our model simulations helped us identify the attractor space and stable motifs. A Boolean model will

eventually settle down into a static state where all the nodes are either 'on' or 'off' and remain in that

state through further iterations. This state of the model is called an attractor and in biological modeling

attractors depict specific biological phenotypes. Within this, a set of nodes within the network forms

cycles which stay in specific states, and these are called stable motifs. We can represent a network as a

reduced system using the nodes that are critical for controlling the stable motifs. We can then study the

model dynamics using the reduced form and can make predictions using simpler representations of the

model.

Attractors of Boolean networks represent biological phenotypes, here the 'dauer' and 'no dauer' fates.

We explored the set of attractors and stable motifs for our model by iterating through all possible

combinations of initial conditions and perturbations using pystablemotifs software [36, 37]. A

perturbation of a node in our experiments refers to reversing the existing state of a node and specify

that along with the initial conditions set. The network diagrams were created using Cytoscape [63].

Code, simulations and scripts are available at https://github.com/alekhyaa2/Dauer-Boolean-Model.

Results

Constructing a dauer Boolean network

Our exhaustive literature survey identified essential regulatory interactions among dauer pathway

genes. From these we constructed a Boolean network with 26 nodes and 34 interactions (Figure 1a;

Supplementary Table 1). The nodes 'pheromone' [34] and '*cmk-1*' (a food receptor [35]) were input

signals that allowed the model to converge onto the dauer/no dauer output node.

Validating the model by replicating Daf-c and Daf-d phenotypes

In *C. elegans* mutant strains can be categorized as dauer-constitutive (Daf-c), genotypes which produce

dauer arrest even in the absence of pheromone and presence of food, and Dauer-defective (Daf-d),

genotypes which do not enter dauer arrest under typical dauer-favorable conditions (i.e., pheromone

present and/or food limited) [18, 40-42]. We validated our model by knocking down known Daf-c genes

(*daf-2*, *daf-7*, *daf-11*, and *daf-9*) separately under initial conditions (*cmk-1* = 1 and pher = 0) that do not

typically favor dauer and found that these mutants followed the Daf-c phenotype and entered dauer

arrest. We also knocked down known Daf-d mutant genes (*daf-12* and *daf-16*) under conditions that

typically favor dauer (here, *cmk-1* = 0 and pher = 1) and found these replicated the Daf-d phenotype,

failing to enter dauer arrest.

Identifying stable motifs

A network attractor is a set of gene states the system can stabilize in [36]. The dauer Boolean network

had 10 different attractor states (Supplementary Figure 1) with 1 resulting in developmental progression

to adult, 8 resulting in the dauer fate and 1 resulting in either dauer or no dauer depending on system

conditions [37]. Although an attractor represents a potential stable point, not all attractors are

accessible from all states. To further analyze system stability, we divided the total network into small

subgraphs [3, 38]. We identified three Stable Motifs (SMs) in the developmental progression to adult

(Figure 1b-d). Each of these encompassed genes from at least two different pathways. Stable Motif 1

(SM1; Figure 1b) spanned 10 genes across the IGF-1/insulin signaling, TGF-β and Steroid hormone

pathways [13]. SM2 and SM3 involved 4 and 8 genes, respectively, across the IGF-1/insulin signaling and

TGF-β pathways. *daf-2*, *daf-7* and *hsf-1* (heat shock protein) were common between all 3 stable motifs

with *hsf-1* acting as a crucial node where different pathways intersect. This result has been shown by

previous experimental work [39]. Each of the SMs involved a different insulin signaling gene, *ins-1*, *ins-7*

or *ins-18*. We did not identify any SMs for dauer formation (the A3 attractor; Supplementary Figure 1).

Identifying driver nodes in the SMs

We simulated gene perturbations to identify the driver nodes for each of our identified SMs [37].

Changing the existing state of the critical nodes shared between the stable motifs (i.e., *daf-2* or *daf-7*

downregulation or *hsf-1* upregulation) prevented the system from entering into any of the SMs and

pushed it to one of the dauer attractors. Perturbation analysis identified the critical causal interactions

in the network as SM2, encompassing *daf-2*, *daf-7*, *ins-7* and *hsf-1*. Perturbation of *daf-2*, *daf-7* or *hsf-1*

genes prohibited the system from stabilizing in the SM and the attractors corresponded to each of these

nodes. We concluded that *daf-2*, *daf-7* and *hsf-1* were the driver nodes for the SMs. When we perturbed

the other nodes of these three stable motifs, the system converged onto either dauer or no dauer via

different stable motifs. (Supplementary Table 3). But in case of *daf-2*, *daf-7* and *hsf-1* perturbations, the

system did not enter into any of the stable motifs, instead settling into the attractors of those

corresponding nodes.

Perturbations of *daf-2*, *daf-7* and *hsf-1* resembled the Daf-c mutant system. When the model was

started with dauer-favorable conditions, the system did not have any stable motifs. The same lack of

stable motifs occurred with Daf-c mutant phenotypes. Under both dauer-favorable and Daf-c mutant

conditions (here, related gene perturbation and/or favorable initial conditions), the system did not have

any cyclic subgraphs facilitating control. Instead, the driver nodes decided the fate of the organism in

expressing a wildtype or a Daf-c mutant phenotype.

Analyzing causal interactions in the dauer Boolean network

We identified a reduced network representing causal information flow [43] from environmental input to

dauer fate (Figure 2a). The reduced network contained a cyclic subgraph (Figure 2b) similar to SM2,

containing *daf-2*, *daf-7*, *ins-7* and *hsf-1*. However, it included a necessary inhibitory regulation of *daf-9*

by *daf-2*. The dynamics of the entire system can be summarized as input to this subgraph. The full

reduced network (Figure 2c) included 12 nodes and 13 edges.

Insulin signaling drives the binary dauer/no dauer fate

Gene perturbations revealed an additional 7 SMs (SM4-10; Supplementary Table 3). For each of the

perturbations performed, the system could end up in one of 10 different attractors but few of these

could be reached from stable conditions with active stable motifs. Activating 'dauer' conditions including

low food and pheromone presence resulted in dauer fate and the A3 attractor (Figure 3a) and available

food with pheromone absence resulted in the system reaching the A7 attractor through SMs 1, 2 or 3

(Figure 3b).

We identified insulin-dependent dauer arrest attractors across the perturbations. Activation of *ins-1* and

ins-18 resulted in the dauer fate through SMs 4-10 but inactivation of either pushed the system to

normal development (Figure 3c). Together with *ins-7* the activation/inactivation profile of these genes

pushed the system to normal development through SMs 1, 2 or 3 (Figure 1b-d). Similarly, *daf-12*

activation resulted in dauer arrest with the insulin profile determining the SM the system entered.

When the antagonistic insulins (*ins-1* and *ins-18*) were activated the system stabilized into SM 8 and

when *ins-1* was turned off the system stabilized in either SM 2 or 3 (Figure 3d).

We found that the system must travel through SMs 1, 2 or 3 to end up in normal developmental

progression and avoid the dauer fate. To enter dauer arrest the system must travel through one of SMs

4-10 with *ins-1* and *ins-18* activated and *ins-7* inactivated. In this model insulin signaling controls the

transitions between normal development and dauer in the presence of gene perturbations. The insulin

genes act as decision makers for the convergence of these dauer pathways.

Identifying binary states across dauer pathways

For normal development we identified one strongly correlated module, one strongly anticorrelated

module and a third set of genes with quantitative relationships across the SMs (Figure 4). For dauer

arrest we identified two strongly correlated modules and two strongly anticorrelated modules. These

modules span the 4 signaling pathways involved in dauer arrest and indicate co-regulation and

correlative relationships between distant genes in the network.

Discussion

We have constructed a Boolean model of the dauer arrest network and identified the critical, causal

interactions driving the system between dauer fate and normal development. Although the network

contains 26 genes and 34 interactions, the system can be driven between the dauer/no dauer fate by

controlling a few critical nodes. Importantly, the switchpoints for this system require specific

combinations of gene states behaving synergistically or antagonistically. The results of our model

demonstrate that complex systems can be controlled through a few critical nodes and connections,

echoing both control theory [3, 44] and long-standing theoretical work on biological systems [45].

Evolutionary studies of signaling pathway divergence and conservation have focused on individual genes

and sequence-level changes [33]. Our results demonstrate that diffuse, correlative interactions have no

impact on system stability or developmental fate. We suggest that system-level function and dynamics

are, instead, the true evolutionary targets of selection.

Many of the results of our study mirror experimental dauer investigations. For example, a systematic

analysis of mutants across insulin signaling genes in *C. elegans* identified roles in lifespan, reproduction,

pathogen resistance, thermotolerance and dauer entry and exit [46]. Here, by constructing a network

that encompassed multiple signaling pathways we were able to analyze the role of these insulin

signaling genes and robustly identify the interaction between insulin signaling, heat shock response and

daf-sensitive genes that determines dauer fate. Our results demonstrate that dauer signaling pathways

find different routes to transduce the signals from the environment to developmental decision.

Over 100 genes have been described as influencing the dauer fate in *C. elegans* [13, 29]. However, many

of these have been identified through mutant analyses or total loss of function alleles [47]. Genes may

be capable of influencing the dauer state through loss-of-function or extreme phenotypes but this may

not translate to a gene's involvment in dauer fate in natural populations. We focused on the genes and

interactions determined essential to the dauer network. Distilling these further into causal interactions

revealed a core set of genes and proteins controlling dauer fate. Transcriptomic studies have identified

large, co-regulated, correlated modules associated with dauer mutants, entry and exit [46]. Our results

show that these co-regulated modules also occur in our model (Figure 4) but their expression and

function are correlated consequences of developmental decision-making occurring at a few critical
points in the system.

C. elegans is a microscopic soil-dwelling organism and extremely dependent on environmental
conditions. Some type of developmental arrest or metabolic stasis occurs in many small organisms with
similar environmental dependence. For example, tardigrades (Phylum Tardigrada) are microscopic
invertebrates closely related to nematodes and can enter both ametabolic and anhydrous states where
they extrude all water and remain in suspended animation for multiple years [48]. Despite the ubiquity
of developmental and metabolic stasis, there is little understanding of the genetics and regulatory
systems underlying these processes [49]. Evolutionary studies of developmental arrest across
nematodes have used dauer genes and pathways as a tool to construct related models. For example, the
parasitic nematode *Haemonchus contortus* transitions from both its L3 and L4 stages depending on cues
from the host's gut environment [50]. A dauer model based on *C. elegans* identified 61 gene
homologues with varying levels of conservation, each of which may be a potential target for life cycle
disruption and anti-parasitic intervention [51]. Alternatively, many of these genes may not be involved
in developmental arrest in *H. contortus*. They may have pleiotropic functions in *C. elegans* or have been
co-opted for other organismal phenotypes.

Multiple studies support this scenario of conserved dauer phenotype and divergent gene and sequence
evolution. A study of dauer-related genes and pathways within Phylum Nematoda found a high number
of *Caenorhabditis* and *C. elegans*-specific dauer genes and an evolutionarily recent co-option of TGF-β
signaling into the dauer pathway [29]. Although the insulin signaling pathway is broadly conserved,
outside of *Caenorhabditis* the insulin genes examined showed little evolutionary conservation. In the
nematode *Pristionchus pacificus* dauer pheromones produce dauer entry in individuals of other
genotypes, suggesting the worms are using dauer signaling in intraspecific competition [22]. Natural
variation in *P. pacificus* dauer is partially explained by copy number variation in *dauerless*, an 'orphan'
gene that lacks homologs in related species [52]. These examples of near-scale divergence suggest that
although dauer itself is a conserved developmental fate, the individual genes underlying dauer may not
be essential targets of selection.

Our results suggest that system-level properties responsible for developmental fate may be the true
evolutionary phenotypes under selection. The small number of genes and interactions underlying stable

motifs and the causal logic backbones may be subject to higher levels of evolutionary conservation, or

their system 'role' may be conserved. Laboratory genetic manipulations within *C. elegans* and across

nematodes, focusing on the key genes and interactions identified here, would be a tractable, exciting

avenue for future study. For example, within *C. elegans* there are at least 40 insulin signaling homologs

[5]. Divergence, loss and duplication of individual genes may not be as significant as the role the

homologs together play in dauer signaling.

This work has important future applications. The model can be used to study how different cellular

responses work in cohort with dauer pathways by extending the system to encompass genes and

interactions involved in disparate pathways including the cell cycle and immune response. Expanding

this framework can provide a tractable system to test experimental hypotheses that are currently

challenging, labor-intensive and beyond the scope of traditional laboratory investigations. Viewing the

conservation and divergence of signaling pathways through the lens of causality and system-level

function can provide a robust framework for understanding the evolution of biological networks.

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Figure 1 Models representing the four major dauer inducing pathways and the stable motifs of
this network. (a) The dauer Boolean network included 26 nodes and 34 edges. Each node
represented a molecular entity (gene, receptor, protein, etc.) involved in signal transduction
pathways. Edges indicated regulatory interactions between the nodes that activated (signified by
blue arrow heads) or inhibited (signified by red 'T' edges). The four dauer inducing pathways were
represented by four different colors. Here, *cmk-1* and pheromone were input signals which when
turned on and off, decided the dauer fate. Nodes *srbc* (pheromone receptor) and *hsf-1* (heat
shock protein) did not belong to any of the four dauer pathways. (b), (c) and (d) Subgraphs that
were part of the cyclic SMs 1, 2 and 3 respectively, into which the dauer Boolean network
stabilizes. The nodes in red text were inhibited and nodes in black text were activated.

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Figure 2 The causal logic backbone of the dauer Boolean network. (a) The edges were assigned
causal logic (blue = necessary, red = sufficient and black = sufficient and necessary). Nodes filled
grey with dashed borders were input signals where grey filled nodes were part of cyclic
subgraphs (SMs) and solid borders indicated output nodes. (b) The logic backbone derived from
identifying subgraphs from input to motif nodes and motif nodes to output signals. Activation
of pher and *ncr* was sufficient to reach the motif and *cmk-1* was necessary. The motif
stabilization was sufficient to control dauer fate. (c) The reduced dauer logic network consisted
of 12 nodes and 13 edges. *Daf-9* and *daf-12* entered a cyclic feedback loop. The insulin factors
involved in different endocrine pathways were important points of the network regulating *daf-*
2 activity and present in the reduced network.

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Figure 3 Model transitions from initial conditions to attractors. (a) Under unfavourable
conditions, the system eventually reached the "Dauer" attractor A3 without settling in any of
the cyclic motifs. (b) Under favorable conditions (*cmk-1* is on and pheromone is off), the
network stabilized in either of the three cyclic SMs which led to the attractor A7 and the "No
dauer" phenotype. (c) When a node of SM 1, 2 or 3 was perturbed the system stabilized in
different SMs and converged on either dauer arrest or normal development. (d) *daf-12*
perturbation resulted in dauer arrest through SMs 2, 3 or 8. *daf-12* presence, irrespective of
initial conditions, was sufficient for dauer formation. SMs 4-10 and the attractors of the
perturbed node simulations are given in the Supplementary Materials.
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Figure 4 Correlation plots as a function of hamming distance for every node of the attractors
converging to either "dauer" or "no dauer" when the nodes of the SMs were perturbed.
Hamming distance = 0 implied that the two nodes were in the same state and as the distance
value increased the states in which these two nodes exist in the attractors varied. From the
dauer=1 correlation plot we observed clusters of nodes that existed in same state across the
four dauer inducing pathways. *Ins-1* and *ins-18* were positively and negatively correlated when
attractors converged on dauer and no dauer phenotypes, respectively.
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538 **Supplementary Table 1** Regulatory interactions of the dauer Boolean network.

Upstream node	Downstream node	Interaction	PubMed ID
aap-1	age-1	Activating	12393910
age-1	pdk-1	Activating	10364160
akt*	daf-16	Inhibitory	11747825, 11381260, 11747821, 9716402
cmk-1	daf-28	Activating	26335407
cmk-1	daf-7	Activating	26335407
daf-1	daf-8-14*	Activating	10887089
daf-11*	tax-4	Activating	8893027
daf-12	daf-16	Activating	15383841, 12966085
daf-12	daf-16	Activating	15383841
daf-12	daf-9	Activating	15084461
daf-12	dauer	Activating	15383841
daf-16	ins-18	Activating	22683638
daf-16	ins-7	Inhibitory	12845331
daf-16	ins-7	Inhibitory	19489741
daf-16	dauer	Activating	15383841
daf-2	aap-1	Activating	12393910
daf-2	dhs-16	Activating	22505847
daf-2	hsf-1	Inhibitory	23116063
daf-28	daf-2	Activating	12654727, 27125673
daf-3	daf-5	Activating	14681186
daf-5	ins-1	Activating	21533078
daf-7	daf-1-4	Activating	10887089, 8413626, 8570636
daf-7	ins-7	Activating	19489741
daf-8-14*	daf-3	Inhibitory	20081192, 14992718
dhs-16	daf-9	Inhibitory	22505847
hsf-1	daf-7	Inhibitory	23116063
ins-1	daf-2	Inhibitory	21343369, 11274053
ins-18	daf-2	Inhibitory	22683638
ins-7	daf-2	Activating	12845331
ncr*	daf-9	Activating	15509773

pdk-1	*akt**	Activating	10364160
pher*	*srbc**	Activating	19797623
srbc*	*daf-11**	Inhibitory	19797623
tax-4	*daf-7*	Activating	16903785

*The nodes names used in literature have been modified for convenience. Here, *daf-11* =
cGMP, pher = Pheromone, *ncr = ncr-1/2, daf-8-14 = daf-8* and *daf-14, akt = akt-1/2, srbc = srbc-*
64 and *srbc-66*.

Supplementary Table 2 The Boolean rules of the dauer network. Initial conditions for dauer
were *cmk-1* = 0, *ncr* = 0 and pher = 1. Initial conditions for normal development were *cmk-1* = 1,
ncr = 1 and pher = 0. The state of each node at time 't+1' was calculated according to the states
of upstream regulator nodes at time 't'.

Downstream node *= Upstream regulators
*aap-1 *= daf-2*
*age-1 *= aap-1*
*akt *= pdk-1*
*daf-11 *=* not *srbc*
*daf-12 *=* not *daf-9* or not *daf-7*
*daf-1-4 *= daf-7*
*daf-16 *=* not *akt* or *daf-12*
*daf-2 *= (ins-7* and *daf-28)* or *(*not *ins-1* or not *ins-18)*
*daf-28 *= cmk-1*
*daf-3 *=* not *daf-8-14*
*daf-5 *= daf-3*
*daf-7 *= (tax-4* and *cmk-1)* and not *hsf-1*
*daf-8-14 *= daf-1-4*
*daf-9 *=* not *dhs-16* or *ncr* or *daf-12*
*dauer *= daf-16* and *daf-12*
*dhs-16 *= daf-2*
*hsf-1 *=* not *daf-2*
*ins-1 *= daf-5*
*ins-18 *= daf-16*
*ins-7 *= daf-7* or not *daf-16*
*pdk-1 *= age-1*
*srbc *= pher*
*tax-4 *= daf-11*

584 **Supplementary Table 3** Stable motifs in the dauer Boolean network.

Name	Stable motif	Perturbation
SM 4	{'ins-18': 1, 'pdk-1': 0, 'daf-5': 1, 'akt': 0, 'age-1': 0, 'daf-16': 1, 'aap-1': 0, 'hsf-1': 1, 'daf-3': 1, 'daf-8-14': 0, 'daf-2': 0, 'ins-1': 1, 'daf-7': 0, 'daf-1-4': 0}	*ins-7*=0
SM 5	{'ins-18': 1, 'daf-5': 1, 'daf-12': 1, 'daf-16': 1, 'hsf-1': 1, 'daf-3': 1, 'daf-8-14': 0, 'daf-2': 0, 'ins-1': 1, 'daf-7': 0, 'daf-1-4': 0}	*ins-7*=0
SM 6	{'pdk-1': 0, 'daf-5': 1, 'akt': 0, 'age-1': 0, 'daf-16': 1, 'aap-1': 0, 'hsf-1': 1, 'daf-3': 1, 'daf-8-14': 0, 'daf-2': 0, 'ins-1': 1, 'daf-7': 0, 'ins-7': 0, 'daf-1-4': 0}	*ins-18*=1
SM 7	{'daf-5': 1, 'daf-12': 1, 'daf-16': 1, 'hsf-1': 1, 'daf-3': 1, 'daf-8-14': 0, 'daf-2': 0, 'ins-1': 1, 'daf-7': 0, 'ins-7': 0, 'daf-1-4': 0}	*ins-18*=1
SM 8	{'daf-5': 1, 'daf-12': 1, 'daf-16': 1, 'hsf-1': 1, 'daf-3': 1, 'daf-8-14': 0, 'daf-2': 0, 'ins-1': 1, 'daf-7': 0, 'ins-7': 0, 'daf-1-4': 0}	*daf-12*=1, *daf-16*=1, *akt*=0, *pdk-1*=0, *age-1*=0, *aap-1*=0
SM 9	{'ins-18': 1, 'daf-12': 1, 'daf-16': 1, 'daf-2': 0, 'daf-7': 0, 'ins-7': 0, 'hsf-1': 1}	*ins-1*=1, *daf-5*=1, *daf-3*=1, *daf-8-14*=0, *daf-1-4*=0
SM 10	{'ins-18': 1, 'pdk-1': 0, 'akt': 0, 'age-1': 0, 'daf-16': 1, 'hsf-1': 1, 'daf-2': 0, 'daf-7': 0, 'ins-7': 0, 'aap-1': 0}	*ins-1*=1, *daf-5*=1, *daf-3*=1, *daf-8-14*=0, *daf-1-4*=0

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Supplementary Figure 1 Trajectories of attractors and phenotypes. (a) A heatmap that shows
the states of every node of the system in its steady state for 10 different attractors (A1 – A10)
identified for this dauer network. Attractors A3 and A7 represent the two wild type phenotypes
of this system i.e. "dauer" and "no dauer". The nodes part of the four different pathways and
those that are output/input signals are annotated. (b) A similar heatmap for different
phenotypes exhibited by the system.

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