Harnessing Complexity through Evolutionary Dimensional Reduction

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(My) standpoint in Universal Biology Life System consists of diverse components, maintains itself and can continue to produce itself Guiding Principle -- Macro-Micro Consistency: micro - diverse components (high-dimensional) Thousands of chemical species macro – unit to sustain/ reproduce as a whole (*low-dimensional description*?) molecule – cell, cell-tissue etc. Steady (growth) state An Introduction Micro-macro to Complex Systems Constraint from Biology relationship macro to micro Springer Complex-systems Universal statistical law Biology

Consistency between hierarchical levels (+collapse)



Consistency between dynamics of different levels (1)Cell reproduction vs molecule replication \rightarrow universal statistical laws in gene expression (Furusawa et al, PRL 2003,2012, Biophysics 2006,KK etal, PRX2015) (2)Adaptation \rightarrow universal adaptation laws (Kashiwagi et al Plos One2005; Furusawa, KK Phys RevE2018) (3) Differentiation: Cell vs multicellularity \rightarrow Oscillatory dynamics => pluripotency + cell-cell interaction \rightarrow differentiation, loss of pluripotency

(KK&Yomo 1997, Furusawa&KK,1998,Science 2012)

(4) Genetic vs phenotypic changes \rightarrow

Isogneic Phenotypic Variance by noise ∞ variance by genetic change Vg ∞ Evolution Speed (plasticity) **Robustness to noise** ~ to robustness to genetic change, (PNAS03,PLosOne07,Furusawa,KK,Interface2015,PRE 2018) Part I: Consistency (with robustness) between molecule and cell levels :

 \rightarrow Evolutionary Dimensional Reduction in phenotypic dynamics

 \rightarrow Law in Adaptation and Evolution

Response Theory

Part II: Evolutionary Fluctuation-Response Relationship

 \rightarrow Pheno Variance by noise \propto that by mutation

 ∞ evolution speed

Phenotypic Evolution is directed (predictable), before genetic evolution

- Basic Setup (Exp/Theory/Model)
- Phenotype=Abundances of each component (e.g., protein/mRNA) (~5000 dimensions)

Genotype- DNA seq, or rule for dynamics:



- * Model: (i)catalytic reaction network for growth
- (ii) Gene regulation net: (high-dim dynamics):
- * **Theory**: Low-dim constraint in high-dim states

Trivial(?) Law in Adaptation: Focus on steady-growth cells \rightarrow universal constraint all the components have to be roughly doubled (for cell division) : steady-growth condition **Xi** – log(concentration of component i) (i=1,,,M) \rightarrow (M-1) conditions \rightarrow 1-dimensional line M large: e.g., # of protein species $\sim (10^3 \sim 10^4)$ E: Environment; δE; added Stress $dX_{i}/dt = F_{i}(\{X_{i}\}) - \mu$ $F_i(\{X_i^*(E)\}, E\} = \mu(E).$ Stress ED Linearization, "small" δΕ, δΧ, δμ Stress E* $\frac{\delta X_j(E)}{\delta X_j(E')} = \frac{\delta \mu(E)}{\delta \mu(E')} = \frac{\delta \mu(E)}{\delta \mu(E')}$ KK, Furusawa, Yomo, PhysRevX (2015) for given type of stress E (changing strength)

Concentration xi=Ni/V: $(dV/dt)/V = \mu$ (volume V) Temporal change of concentration x (Any reaction dynamics)

$$dx_i/dt = f_i(\{x_j\}) - \mu x_i$$
 dilution

Now, the stationary state is given by a fixed point condition

 $x_i^* = f_i(\{x_j^*\})/\mu$

for all i.

As a convenience, denote X = logx, and $f_i = x_i F_i$. Then,

 $dX_i/dt = F_i(\{X_j\}) - \mu$

Response under different stress strength E

 $F_i(\{X_i^*(E)\}, E) = \mu(E).$

Trivial so far



Put E Coli under different strength of stresses; Measure gene expressions (mRNA concentrations)



Non-trivial point: Emergent "Deep Linearity"

- (1) Large Linear Regime?
- (2) Validity across different environmental condition?
- --beyond just steady-growth system
- achieved in an evolved system ?

Across Different types of stresses: $\gamma_i \equiv \frac{\partial F_i}{\partial E}$ $\gamma(a)$ depends on stress type a so correlation not $\delta X_j(E) = \delta \mu(E) \times \sum L_{ji} (1 - \gamma_i / \alpha)$ derived, but... (b) (c)2 δX_i(E_{high}) δX_i(E_{high}) δX_i(E^{heat} -1 -2 -2 2 SY (Eheat $\delta X_{i}(E_{high}^{osmo})$ $\delta X_{i}(E_{high}^{osmo})$ osmotic / heat starve/osmotic starve/heat Still highly correlated Confirmed also in protein expression changes across different environmental conditions

Fig.2b

Better(?) confirmed in protein expression changes across different environmental conditions (based on the data by Heinemann) 20 different conditions on E Coli



• High-dimensional adaptation system (diversity) is important for expanded liner regime and applicability for diverse environmental changes

* emergence of 'collective' slow variable (Image) homeostatic core (major parts) -- proportional change, self-consistent; few genes absorb specific environmental stresses

> Cove part (no direct

Relevant for robustness of a high-dimensional state

Non-trivial point: Emergent "Deep Linearity"

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Check by simulations of toy models with high-dim dynamical systems

Examine by Toy Cell Model with Catalytic Reaction Network (Cf. Furusawa,KK, PRL 2003, 2012)

k species of chemicals X_o···X_{k-1}

number ---n_o n₁ ... n_{k−1}
 random catalytic reaction network with the path rate p

for the reaction $X_i + X_j - > X_k + X_j$

Resource chemicals (<environment) are transported with the aid of a given catalyst, transporter

resource chemicals are thus
Facilitate transformed into impenetrable chemicals, transport leading to the growth.

N=**Σ** n_i exceeds N_{max} (model 1)

Genotype: Network;

Fitness: e.g., growth rate

Evolution: Mutate reaction paths, and select those with higher fitness

Model (stochastic reactions)



dX1/dt ∝ X0X4; rate equation; Stochastic model here

Evolve Network to increase the growth rate under given resource condition



resource concentrations i=1,2,..,10 e.g., (e0,e0,,,e0)

Then put different environmental conditions Env = λ (e1,e2,e3,..e10) + (1- λ) (e0,e0,..., e0) -1< e1,e2,... <1 (randomly chosen) Check the change in concentrations and growth rates against λ

Evolution shapes Global Proportionality across different environmental conditions



generation

KK, Furusawa, Ann Rev Biophys 2018

After evolution, correlation across different env cond. Increases + slope-growth-rate proportional Between same



Phenotypic constraint on a low-dimensional space



After evolution, the environmental response is constrained on a low-dimensional phenotype space.

Phenotypic change due to environmental variation, mutation, noise are constrained along a major axis



Formation of Dominant Mode Along Major Axis



Robust to perturbations – strong attraction from most directions except one direction along which evolution progresses

(Both environment- and evolution- induced) changes in high-dimensional phenotype space are constrained along low-dimensional slow-manifold

Furusawa, KK, Phys.Rev E 2018; KK, Furusawa, Ann Rev Biophys 2018

Formulation and Consequence of Hypthesis

Recall
$$\sum_{j} J_{ij} \delta X_j(E) + \gamma_i \delta E = \delta \mu(E)$$

with $\gamma_i \equiv \frac{\partial F_i}{\partial E}$. $\delta \mathbf{X} = \mathbf{L}(\delta \mu \mathbf{I} - \gamma \delta E)$

• γ(E): susceptibility to environment change

Only the smallest eigenvalue in J (or largest in L=1/J) contributes $|\lambda^{i}| >> |\lambda^{0}| \sim 0$

Most changes occur along such slow manifold

$$\delta \mathbf{X} = \lambda^0 \mathbf{w}_0 (\delta \mu (\mathbf{v}_0 \cdot \mathbf{I}) - (\mathbf{v}_0 \cdot \gamma) \delta E).$$

Projection to this manifold wo w⁰ (v⁰) right(left) eigenvector for the smallest eigenvalue, i.e., Projection to this slow manifold $\frac{\delta \mathbf{X}(\mathbf{E})}{\delta \mathbf{X}(\mathbf{E}')} = \frac{\delta \mu(E) - (\mathbf{v_0} \cdot \gamma(\mathbf{E})) \delta E / (\mathbf{v_0} \cdot \mathbf{I})}{\delta \mu(E') - (\mathbf{v_0} \cdot \gamma(\mathbf{E}') \delta E' / (\mathbf{v_0} \cdot \mathbf{I})} \operatorname{small}^{\mathsf{V} \cdot \mathsf{V_0}}$ Consequence of Slow-Manifold Hypothesis (cont'd)

→Slow manifold is roughly orthogonal to γ γ · ν_0 ~⁰

$$\hat{\mathbf{X}} = \lambda^{0} \delta \mu \mathbf{w}^{0}$$
Or, from the linear approximation
$$\delta E = \delta \mu / \alpha(E)$$

$$\frac{\delta \mathbf{X}(E)}{\delta \mathbf{X}(E')} = \frac{\delta \mu(E)}{\delta \mu(E')} \underbrace{(1 - (\mathbf{v}_{0} \cdot \mathbf{E}) / (\alpha \mathbf{v}_{0} \cdot \mathbf{I}))}_{(1 - (\mathbf{v}_{0} \cdot \gamma(E') / (\alpha \mathbf{v}_{0} \cdot \mathbf{I}))} \underbrace{(1 - (\mathbf{v}_{0} \cdot \gamma(E') / (\alpha \mathbf{v}_{0} \cdot \mathbf{I}))}_{-\Delta \mu}$$
Correction in proportion coefficient

Separation of slowest mode in catalytic reaction net model Eigenvalues of $J_{ij} = (\partial \dot{X}_i / \partial X_j)_{X_i = X_i^*}$



1st PCA

1st PCA

→ Evolution -- Recall: Phenotypic change due to environmental variation, mutation, noise are constrained along the same major axis



Phenotypic changes by evolution and environmental changes are along a common dominant mode Consequence of Hypothesis \rightarrow Correlation between Environment (E) vs Evolutionary (genetic) (G) Changes $J\delta X + \gamma(E)\delta E + \gamma(G)\delta G = \delta \mu(E)$. Again, assume that most changes occur along such slow manifold Project to this slow manifold \rightarrow

using $\gamma \cdot v_0 \sim 0$

(Genetic) evolution under the environmental condition →recover growth-- | δμ (E) | > | δμ (G) | δXi(G)/δXi(E)=δμ(G)/δμ(E)<1

→ All the expression levels tend to return the original level by evolution
Le Chatelier Principle?



Deterministic phenotypic evolution constrained in



Mutation sites are different by strains. But.. Common trends in phenotypic space (low-dim structure) PC1 is highly correlated with the growth rate Evolution of Catalytic reaction net model by switching environment (nutrient concentratyion) and check evol-env response

Mutate network and select those with higher growth –evo



Recovery of growth rate by adaptive evolution to new environment $\delta Xi(G)/\delta Xi(E) = \delta \mu(G)/\delta \mu(E) < 1$ (Ac

(Across all components)



Evolution to novel environment -- the already evolved dominant mode is adopted to adapt to new environment -> Same phenotypic path when the tape is replayed.



Different color : different strains with different genetic change

Sato, KK, PhysRevRes2020

PC1

 Evolution to novel environment -- the already evolved dominant mode is adopted to adapt to new environment -> Same phenotypic path when the tape is replayed.





Recall...



Furusawa, kk Interface 2015 Vip-Vg relationship across traits (phenotypes)

- Vg(i): Vatiance of X(i) due to genetic mutation
- Vip(i): Variance of X(i) due to noise in dynamics



Vg-Vip proportionality is explained by the slow manifold Hypothesis

Evolution occurs along this dominamt manifold \mathbf{w}

$$V_{ip}(i) = (\mathbf{w}_i^0)^2 < \delta X^2 >_{noise}$$
$$V_g(i) = (\mathbf{w}_i^0)^2 < \delta X^2 >_{mutation}$$

 \rightarrow Vg(i)/Vip(i) = independent of i

(here we do not need the growth-rate constraint, only slow-manifold constraint is needed)

Vg-Vip relationship ← Changes both by (environmental) noise and (genetic) mutations are constrained along the direction

Need further studies to establish the present theory

- (i) Further Confirmation by Experiments
- (ii) Confirmation by Models : Universality? Catalytic Reaction Net-Cell Model ☑ Gene regulation Net Model (Sato, KK in prep) ☑ Spin-glass Models (Sakata KK., PRL 2020) ☑ evolve spin Hamiltonian JijSiSj to achieve certain configuration dimensional reduction at replica symmetric phase Protein Model/Data (Tang KK., PRL2021) ☑

correlation in structure dynamics & evolutionary dim reduction (iii) Theory for dimensional reduction? –1 or few dim? outliers in eigenvalues – separation of slow modes, Renormalization Group??? Projection to Collective Modes?

Protein; Change in Native structure by noise & by evolution, highly correlated and low-dimensional



Spin-Statistical ModelSakata,KK,PRL 2020Phenotype=Spin config.SiGenotype—Interaction JijHamiltonianH=-ΣJijSiSjFitness align target spins; environment– external field

$$\psi(\mathbf{J}) = \overline{|m_{\mathcal{T}}|}, \qquad m_{\mathcal{T}} = \frac{1}{N_T} \sum_{i \in \mathcal{T}} S_i,$$

1) Robust fitted state at Replica Symmetric phase
 2) RSB → loss of robustness
 (cf Sakata Hukushima KK PRL 2000)





Correlation in Responses to ext field and to mutation to Jij







Congruence between development and evolution (cf, Haeckel, recapitulation)

For most (95%) examples, good correspondence





Messages

- (Cellular) Phenotypes are high-dimensional, but their adaptive changes are drastically restricted in a low-dimensional space
- Slow modes evolve and fascillitate evolution
- ←Result of steady-growth and evolutionary robustness (to noise and to genetic changes)
- Phenotypic evolution is rather deterministic even though genetic changes can be stochastic (replaying the tape, phenotypically same path)
- ← Phenotypic evolvability correlated by shortterm dynamics and fluctuation

Summary

Low-dimensional structure formed from highdimensional phenotypic space ← robustness (Furusawa, KK, Phys Rev E, 2018; KK, Furusawa, Ann Rev Biophys 2018;

Sato, KK, PRR 2020; Sakata, KK, PRL 2020, Tang KK PRL 2021)

