## Harnessing Complexity<br>through through Harnessing Complexity<br>through<br>Evolutionary Dimensional Reduction essing Complexity<br>through<br>Dimensional Reduction<br>Kunihiko Kaneko

Universal Biology Institute, U. Tokyo<br>  $\rightarrow$  $\rightarrow$  $\begin{aligned} &\text{Kunihiko Kaneko} \\ &\text{inverseal Biology Institute, U. Tc} \\ &\longrightarrow \\ &\text{Niels Bohr Institute} \end{aligned}$ 

(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 speciesmacro – 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 relationship and the side of macro to micro 2 Springer Complex-systems Universal statistical law<sup>®</sup> 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 **Consistency between dynamics of different levels<br>(1)Cell reproduction vs molecule replication**  $\rightarrow$ **<br>universal statistical laws in gene expression<br>(Furusawa et al, PRL 2003,2012, Biophysics 2006,KK etal, PRX2015)<br>(2)Adapta Consistency between dynamics of different levels<br>
(1)Cell reproduction vs molecule replication**  $\rightarrow$ **<br>
universal statistical laws in gene expression<br>
(Furusawa et al, PRL 2003,2012, Biophysics 2006,KK etal, PRX2015)<br>
(2)Ad Consistency between dynamics of dif<br>(1)Cell reproduction vs molecule replic<br>universal statistical laws in gene expression<br>(Furusawa et al, PRL 2003,2012, Biophysics 2006,KK eta<br>(2)Adaptation**  $\rightarrow$  **universal adaptation law** (3) Differentiation: Cell vs multicellularity  $\rightarrow$ (1) Cell reproduction vs molecule replication  $\rightarrow$ <br>universal statistical laws in gene expression<br>(Furusawa et al, PRL 2003,2012, Biophysics 2006, KK etal, PRX2015)<br>(2) Adaptation  $\rightarrow$  universal adaptation laws (Kashiwagi interaction  $\rightarrow$  differentiation, loss of pluripotency al, PRL 2003,2012, Biophysics 2006,KK etal, PRX2015)<br> **(KARA)**: Furusawa, KK Phys RevE2018)<br>
The Serve 2018)<br>
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The Serve 2012<br> **(KK&Yomo 1997, Furusawa&KK,1998,Science 2012)** (2) Adaptation  $\rightarrow$  universal adaptation laws (Kashiwagi et al<br>
Plos One2005; Furusawa, KK Phys RevE2018)<br>
(3) Differentiation: Cell vs multicellularity  $\rightarrow$ <br>
Oscillatory dynamics => pluripotency + cell-cell<br>
interaction Plos One2005; Furusawa, KK Phys RevE2018)<br>
3) Differentiation: Cell vs multicellularity →<br>
Dscillatory dynamics => pluripotency + cell-cell<br>
interaction → differentiation, loss of pluripotency<br>
(KK&Yomo 1997, Furusawa&KK (3) Differentiation: Cell vs multicellularity  $\rightarrow$ <br>Oscillatory dynamics => pluripotency + cell-cell<br>interaction  $\rightarrow$  differentiation, loss of pluripotency<br>(KK&Yomo 1997, Furusawa&KK,1998,Science 2012)<br>(4) Genetic vs phen

Robustness to noise  $\sim$  to robustness to genetic change, (PNAS03,PLosOne07,Furusawa,KK,Interface2015,PRE 2018) Part I: Consistency (with robustness) between molecule and cell levels : Part I: Consistency (with robustness) between<br>molecule and cell levels :<br>→ Evolutionary Dimensional Reduction in<br>phenotypic dynamics

phenotypic dynamics

 $\rightarrow$  Law in Adaptation and Evolution

Response Theory

Part II: Evolutionary Fluctuation-Response Relationship Phenotypic dynamics<br>
Presponse Theory<br>
Part II: Evolutionary Fluctuation-Response<br>
Pheno Variance by noise ∝that by mutation<br>  $\infty$  evolution speed

∝ evolution speed

Response Theory<br>
Notation is directed (predictable),<br>
Phenotypic Evolution is directed (predictable),<br>
before genetic evolution before genetic evolution

- 
- Basic Setup (Exp/Theory/Model)<br>• Phenotype=Abundances of each comp • Basic Setup (Exp/Theory/Model)<br>• Phenotype=Abundances of each component<br>(e.g., protein/mRNA) (~5000 dimensions) (e.g., protein/mRNA) (~5000 dimensions) • Basic Setup (Exp/Theory/Model)<br>• Phenotype=Abundances of each component<br>(e.g., protein/mRNA) (~5000 dimensions)<br>Genotype- DNA seq, or rule for dynamics:<br>Dynamics to shape Phenotype • Basic Setup (Exp/Theory/Model)<br>
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(e.g., protein/mRNA) (~5000 dimensions<br>
Genotype- DNA seq, or rule for dynamics:<br>
Dynamics to shape Ph<br>
Geno-Pheno Mapping?



- \* 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<br>steady-growth cells  $\rightarrow$  universal constraint<br>all the components have to be roughly doubled<br>(for cell division) : steady-growth condition<br>Xi – log(concentration of component i) (i=1, Trivial(?) Law in Adaptation: Focus on steady-growth cells  $\rightarrow$  universal constraint all the components have to be roughly doubled e to be roughly doubled<br>y-growth condition<br>f component i) (i=1,,,M)<br>dimensional line<br>in species  $\sim (0^2 \sim 10^4)$ <br>E: Environment; δE; added Stress<br> $F_i(\lbrace X_i^*(E) \rbrace, E) = \mu(E)$ . (for cell division) : steady-growth condition  $\rightarrow$ (M-1) conditions  $\rightarrow$  1-dimensional line M large: e.g., # of protein species  $\sim$   $(10^3 \sim 10^4)$  $d_i X_i/dt = F_i({X_i}) - \mu$ Stress E<sup>b</sup> Linearization , "small" δE, δX、δμ Stress E<sup>n</sup>  $\frac{\delta X_j(E)}{\delta X_j(E')} = \frac{\delta \mu(E)}{\delta \mu(E')}$  = indep't of j 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) ☑

$$
dx_i/dt = f_i(\{x_j\}) - \omega \text{dilution}
$$

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

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

for all  $i$ .

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

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

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: Across Different types of stresses:<br>
γi(a) depends on stress type a so correlation not<br>
despived, but...<br>
a depends on stress type a so correlation not<br>
(c)  $\frac{dS_{X_3(E)=\delta\mu(E)\times\sum E_{j}(1-\gamma_i)}{E_{X_3(E)=\delta\mu(E)\times\sum E_{j}(1-\gamma_i)}}$ derived, but…  $\sum_{\substack{\text{S.} \text{S.} \text{S.} \text{S.} \text{C.} \text{C$ Still highly correlated  $\frac{z}{2}$ <br>  $\frac{1}{2}$ <br>  $\frac{z}{2}$ <br>  $\frac{1}{2}$ <br>  $\frac{5x_i(e^{osmo})}{6x_i(e^{osmo})}$ <br>  $\frac{1}{2}$ expression changes across different environmental conditions

Fig. 20

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



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

\*emergence of 'collective' slow variable (Image) • High-dimensional adaptation system (diversity) is<br>important for expanded liner regime and<br>applicability for diverse environmental changes<br>\* emergence of 'collective' slow variable (Image)<br>homeostatic core (major parts) change, self-consistent ; few genes absorb specific environmental stresses

 $env2$ 

Relevant for robustness of a high-dimensional state

#### Non-trivial point: Emergent "Deep Linearity"

- (1) Large Linear Regime?
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--beyond just steady-growth system

achieved in an evolved system ?

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  $\mathbf{x}_o \cdots \mathbf{x}_{k-1}$

number --- $n_0$   $n_1$  ...  $n_{k-1}$ **n** random catalytic reaction network with the path rate p

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

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

**resource chemicals are thus** transformed into impenetrable chemicals, transport /  $\cdot$ , Ma leading to the growth.

 $\blacksquare$  N=Σn<sub>i</sub> exceeds N<sub>max</sub> (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 iven resource condition



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

Env =  $\lambda$  (e1,e2,e3,..e10) + (1- $\lambda$ ) (e0,e0,..., e0)<br>Check the change in concentrations and growth rates against.  $-1 < e1, e2, \ldots < 1$  (randomly chosen) Check the change in concentrations and growth rates against λ

# Evolution shapes Global Proportionality across different environmental conditions



KK, Furusawa, Ann Rev Biophys 2018

#### After evolution, correlation across different env cond. Ø 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



Along Major Axis<br>Robust to<br>perturbations – strong<br>attractions attraction from most directions …… except one direction along which evolution progresses

changes in high-dimensional phenotype space are constrained along low-dimensional slow-manifold  $X_1$ <br>Both environment- and evolution- induced)<br>thanges in high-dimensional phenotype space are<br>constrained along low-dimensional slow-manifold<br>Furusawa, KK, Phys.Rev E 2018; KK, Furusawa, Ann Rev Biophys 2018

Formulation and Consequence of Hypthesis

Recall 
$$
\sum_{j} J_{ij} \delta X_{j}(E) + \gamma_{i} \delta E = \delta \mu(E)
$$

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

• γ(E): susceptibility to environment change

Only the smallest eigenvalue in J (or largest in L=1/J) contributes | λ' | >> | λ ο | ∼0|

Most changes occur along such slow manifold

**Projection to this manifold wo**  $w^{\prime\prime}$  (v<sup>0</sup>) right(left) eigenvector for the smallest contributes  $|\lambda^+| \gg |\lambda^0| \sim 0$ <br>
Most changes occur along such slow manifold<br>  $\delta X = \lambda^0 w_0 (\delta \mu (v_0 \cdot I) - (v_0 \cdot \gamma) \delta E).$ <br> **Projection to this manifold we**<br> **w** (v<sup>0</sup>) right(left) eigenvector for the smallest<br>
eigenvalue, i.e. γ・v small  $\overline{0}$ ,  $\overline{0}$ ,  $\overline{0}$ ,  $\overline{0}$  $0 \rightarrow$ riaht/loft)  $\overline{\mathbf{0}}$  and  $\overline{\mathbf{0}}$  a  $\rightarrow$  Slow manifold is roughly orthogonal to **γ** Consequence of Slow-Manifold Hypothesis (cont'd)  $\gamma \cdot v_0 \sim 0$ 

 $0$  and  $0$  a



Separation of slowest mode in catalytic reaction net model eparation of slowest mode in catalytic<br>Eigenvalues of  $J_{ij} = (\partial \dot{X_i}/\partial X_j)_{\mathbf{X_i} = \mathbf{X_i^*}}$ 



 $1<sup>st</sup> PCA$ 

1<sup>st</sup> PCA

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



Phenotypic changes by evolution and environmental changes are along a common dominant mode

Again, assume that most changes occur along such slow manifold Consequence of Hypothesis  $\rightarrow$  Correlation<br>
between Environment (E) vs Evolutionary<br>
(genetic) (G) Changes<br>  $J\delta X + \gamma(E)\delta E + \gamma(G)\delta G = \delta \mu(E)$ .<br>
Again, assume that<br>
most changes occur along such slow manifold<br>
Project to this sl Consequence of Hypothesis  $\rightarrow$  Correlation between Environment (E) vs Evolutionary (genetic) (G) Changes<br> $J\delta X + \gamma(E)\delta E + \gamma(G)\delta G = \delta \mu(E)$ .

$$
\delta Xi(G)/\delta Xi(E)=\delta\mu(G)/\delta\mu(E)
$$

using γ·νο ~0

(Genetic) evolution under the environmental condition  $\rightarrow$  recover growth--  $|\delta \mu(E)| > |\delta \mu(G)|$ δXi(G)/δXi(E)=δμ(G)/δμ(E)<1  $\begin{array}{ll} \hline \delta \mathrm{Xi}(G)/\delta \mathrm{Xi}(E)=\delta \mu(G)/\delta \mu(E) & \textrm{using } v \cdot v_0 \leq 0 \ \end{array}$ <br>
(Genetic) evolution under the environmental condition<br>  $\rightarrow$  recover growth--  $\begin{array}{ll} \hline \delta \mu(E) > | \delta \mu(G) | \ \end{array}$ <br>  $\rightarrow$  All the expression levels tend to

Le Chatelier Principle? level by evolution



### 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

 $(Across all complete)$  $\delta X$ i(G)/ $\delta X$ i(E)=δμ(G)/δμ(E)<1



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



### with different genetic change<br>
Sato, KK, PhysRevRes2020

 $PC1$ 

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





#### Recall…



<sup>Furusawa, kk Interfac</sup><br>Vip-Vg relationship across traits (phenotypes)<br>Vg(i): Vatiance of X(i) due to genetic mutation Furusawa, kk Interface 2015<br>Vip-Vg relationship across traits (phenotypes)

- 
- *V*g relationship across traits (phenotypes)<br>Vg(i):Vatiance of X(i) <mark>due to genetic mutation</mark><br>Vip(i):Variance of X(i) <mark>due to noise </mark>in dynamics Vip(i): Variance of X(i) due to noise in dynamics



## Vg-Vip proportionality is explained by the slow manifold<br>Hypothesis<br>Evelution ecoure along this deminemt menifold **w Hypothesis** Vg-Vip proportionality is explained by the slow manifold<br>Hypothesis<br>Evolution occurs along this dominamt manifold **w**<br> $V_{ip}(i) = (\mathbf{w}_i^0)^2 < \delta X^2 >_{noise}$

Evolution occurs along this dominamt manifold

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

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

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

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

 $V_g(i) = (\mathbf{w}_i^0)^2 < \delta X^2 >_{mutation}$ <br>  $V_g(i)/Vip(i) = \text{independent of } i$ <br>
here we do not need the growth-rate constraint<br>
bw-manifold constraint is needed)<br>  $V_g-Vip$  relationship  $\leftarrow$  Changes both by<br>
(environmental) noise and (genetic) mutation (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) Ø 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) Ø 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 Model Phenotype=Spin config.Si Genotype-Interaction Jij Hamiltonian H=-ΣJijSiSj Spin-Statistical Model<br>
Phenotype=Spin config.Si Genotype-Interaction Jij<br>
Hamiltonian H=-ΣJijSiSj<br>
Fitness align target spins; environment- external field<br>  $\psi(J) = \overline{m_T}$ ,  $m_T = \frac{1}{N_T} \sum_{i \in T} S_i$ , Sakata,KK,PRL 2020 Spin-Statistical Model<br>
Phenotype=Spin config.Si Genotype—Interaction Jij<br>
Hamiltonian H=-ΣJijSiSj<br>
Fitness align target spins; environment— external field<br>  $\psi(J) = \overline{m_T}$ ,  $m_T = \frac{1}{N_T} \sum_{i \in T} S_i$ ,<br>
1) Robust fitted state Moder<br>
in config.Si Genotype—Interaction Jij<br>  $|=-\sum$ JijSiSj<br>
rget spins; environment— external field<br>  $m_T = \frac{1}{N_T} \sum_{i \in T} S_i$ ,<br>
d state at Replica Symmetric phase<br>
of robustness<br>
(cf Sakata,Hukushima,KK PRL 2009)<br>  $\begin{bmatrix}$ 

2)  $RSB \rightarrow loss$  of robustness



**Correlation in Responses to ext** field and to mutation to Jij







60

30

 $\mathbf 0$ 

 $\mathbf{0}$ 

1000

generations

2000

2000

60

30

 $\mathbf 0$ 

 $\Omega$ 

1000

developmental time

D

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

For most (95%) examples, good correspondence



 $0.8$  $0.6$ 

 $0.4$ 

 $0.2$ 

#### **Messages**

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

#### **Summary**

#### Low-dimensional structure formed from highdimensional phenotypic space  $\leftarrow$  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)

