

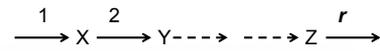


DYNAMICS AND DESIGN OF CELLULAR NETWORKS

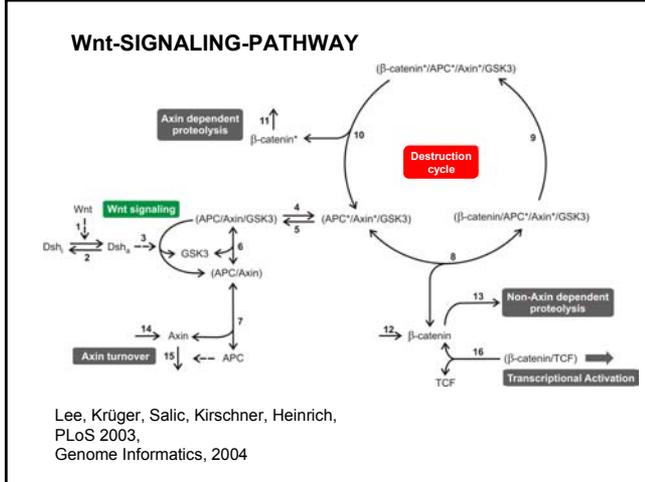
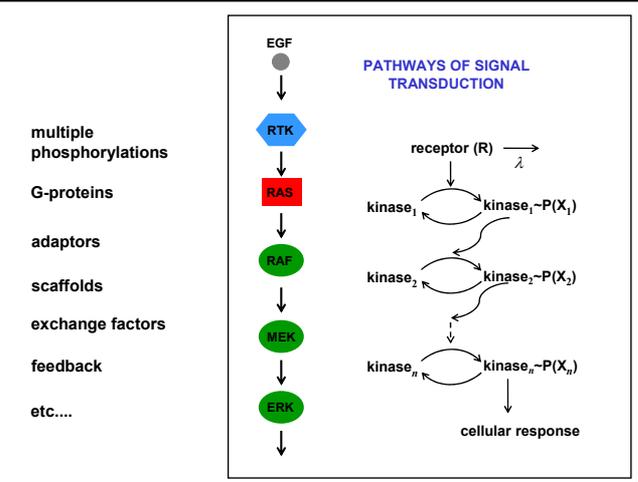
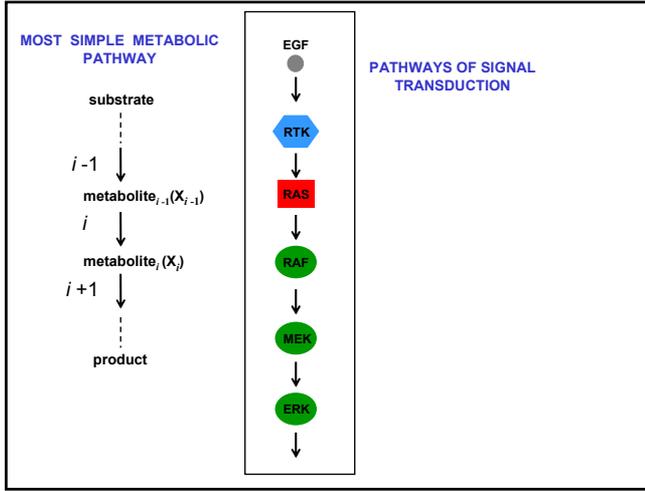
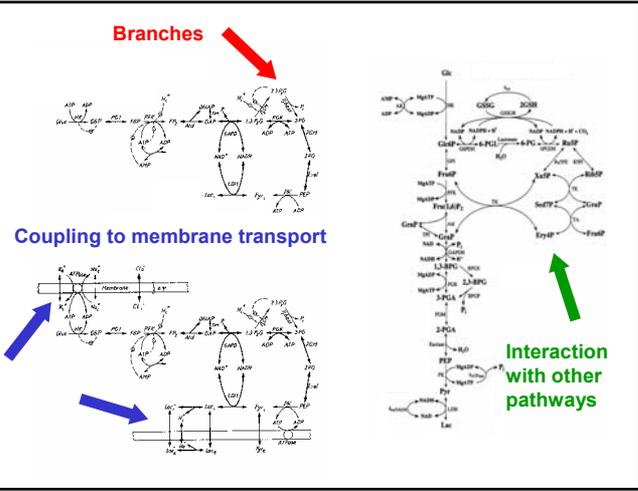
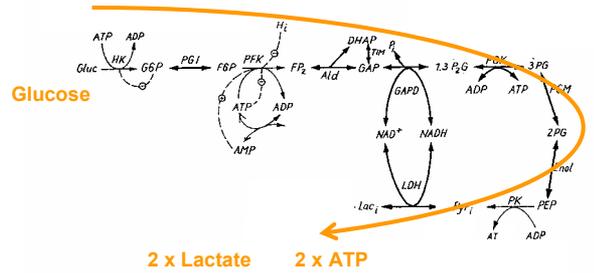
Reinhart Heinrich

Institute of Biology/Theoretical Biophysics
Humboldt-University, Berlin

FEBS Advanced Course, Systems Biology
Gosau, March 12, 2005

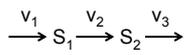


Glycolysis



Lee, Krüger, Salic, Kirschner, Heinrich, PLoS 2003, Genome Informatics, 2004

high number of variables nonlinearity regulatory loops
 separation of time constants kinetic parameters often unknown



$$\frac{dS_1}{dt} = v_1 - v_2$$

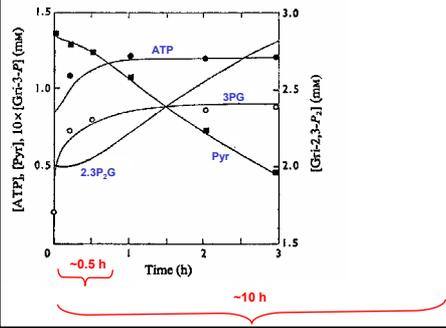
$$\frac{dS_2}{dt} = v_2 - v_3$$

metabolite concentrations stoichiometric coefficients reaction rates

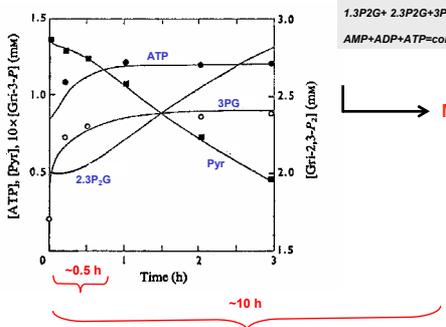
$$\frac{dS_i}{dt} = \sum_{j=1}^r n_{ij} v_j \quad \frac{dS}{dt} = N \cdot v$$

$$v = (v_1, \dots, v_r)^T \quad v = v(S, p)$$

TIME DEPENDENCIES OF GLYCOLYTIC METABOLITES



TIME DEPENDENCIES OF GLYCOLYTIC METABOLITES



separation of time scales
 quasi-steady states

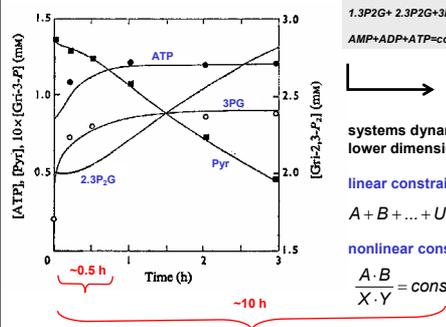
+

conservation quantities

$1.3P2G + 2.3P2G + 3PG + 2PG + PEP + Pyr + NADH = const$
 $AMP + ADP + ATP = const$

MODEL REDUCTION

REDUCING COMPLEXITY



separation of time scales
 (time hierarchy) quasi-steady states

+

conservation quantities

$1.3P2G + 2.3P2G + 3PG + 2PG + PEP + Pyr + NADH = const$
 $AMP + ADP + ATP = const$

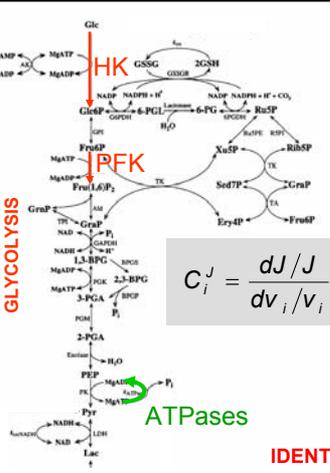
MODEL REDUCTION

systems dynamics in a subspace of lower dimension

linear constraints
 $A + B + \dots + U + V = const$

nonlinear constraints
 $\frac{A \cdot B}{X \cdot Y} = const$

$A \rightleftharpoons B$ $X \rightleftharpoons Y$



REDUCING COMPLEXITY

$$C_{HK}^J = 0.34$$

$$C_{PFK}^J = 0.18$$

$$C_{ATPase}^J = 0.48$$

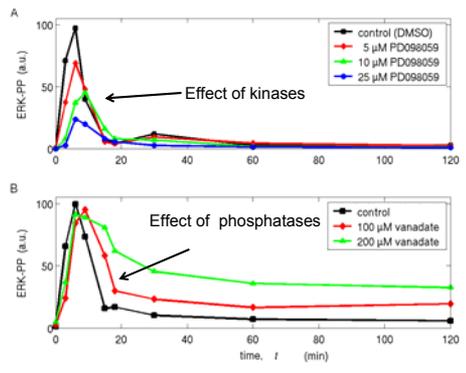
$$C_i^J = \frac{dJ/J}{dv_i/v_i}$$

$$\sum_{k=1}^r C_k^J = 1 \quad \sum_{k=1}^r C_k^S = 0$$

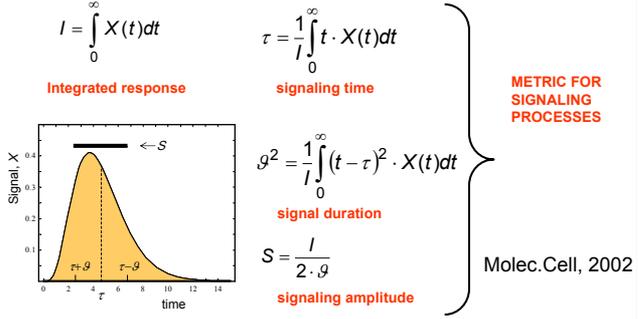
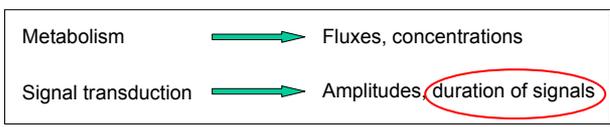
IDENTIFICATION OF KEY STEPS



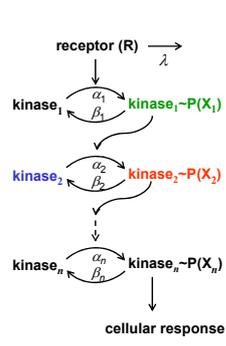
SIGNAL TRANSDUCTION PATHWAYS



J.J.Hornberg et al. FEBS Journal, (2005)



MATHEMATICAL DESCRIPTION



$\frac{dX_i}{dt} = \alpha_i^* X_{i-1} \cdot \tilde{X}_i - \beta_i X_i$

$X_i + \tilde{X}_i = K_i$

$\frac{dX_i}{dt} = \alpha_i X_{i-1} \left(1 - \frac{X_i}{K_i} \right) - \beta_i X_i$

$X_i \ll K_i$: weakly activated pathway

Labels: upstream kinase active form, downstream kinase inactive form, downstream kinase active form.

WEAKLY ACTIVATED PATHWAYS

$\frac{dX_i}{dt} = \alpha_i X_{i-1} - \beta_i X_i$

$\tau = \tau_R + \sum_{i=1}^n \frac{1}{\beta_i}$

$g = \sqrt{\tau_R^2 + \sum_{i=1}^n \frac{1}{\beta_i^2}}$

Labels: "life time" of the receptor, rate constants of phosphatases

Signaling time and signal duration completely independent of kinases

THESE AND OTHER RESULTS: PHOSPHATASES CAN BE MORE IMPORTANT FOR REGULATION THAN KINASES

STRONGLY ACTIVATED PATHWAYS

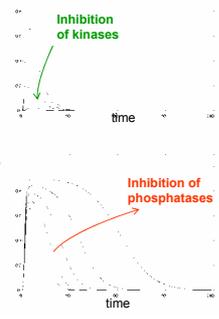
$\frac{dX_i}{dt} = \alpha_i X_{i-1} \left(1 - \frac{X_i}{K_i} \right) - \beta_i X_i$

strong inhibition of phosphatases

$g \cong \frac{1}{\beta_j}$ **LINEAR EFFECT**

strong activation of kinases

$g \cong \ln \alpha_j$ **LOGARITHMIC EFFECT**



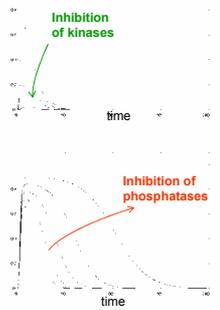
STRONGLY ACTIVATED PATHWAYS

$\frac{dX_i}{dt} = \alpha_i X_{i-1} \left(1 - \frac{X_i}{K_i} \right) - \beta_i X_i$

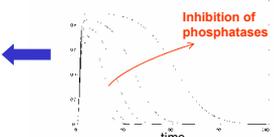
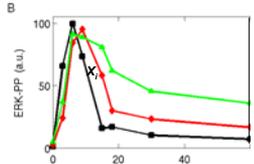
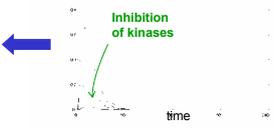
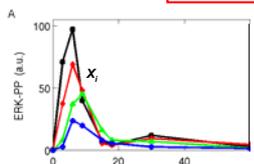
CONTROL COEFFICIENTS

$\sum_{i=1}^n C_{\alpha_i}^A + \sum_{i=1}^n C_{\beta_i}^A = 0$

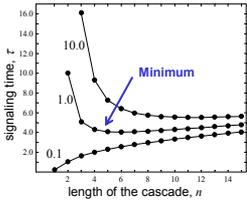
$\sum_{i=1}^n C_{\alpha_i}^r + \sum_{i=1}^n C_{\beta_i}^r = -1$



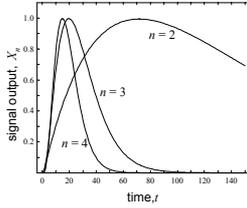
PREDICTIONS



SIGNALS CAN BE FASTER TRANSMITTED THROUGH LONGER CHAINS



$$\tau = \tau_R + \sum_{i=1}^n \frac{1}{\beta_i}$$

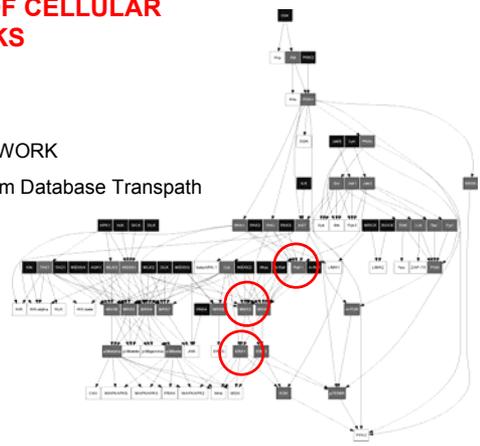


$$S \propto \frac{\alpha_1 \alpha_2 \dots \alpha_n}{\beta_1 \beta_2 \dots \beta_n}$$

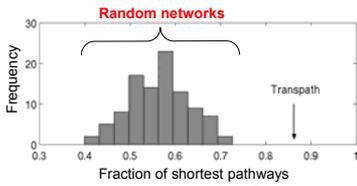
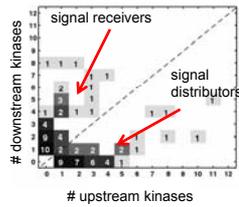
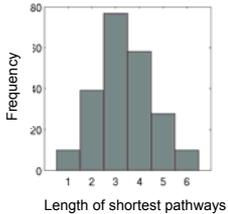
ONE REASON WHY MAPK – PATHWAY 3-4 KINASES ??

DESIGN OF CELLULAR NETWORKS

KINASE NETWORK
Retrieved from Database Transpath

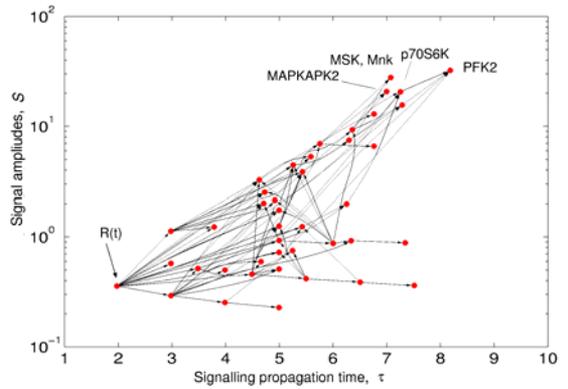


DESIGN OF KINASE NETWORKS

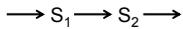


Binder & Heinrich (2004) Genome Informatics. Vol. 15

DYNAMICS OF LARGE SCALE SIGNALING NETWORKS



DESIGN OF CELLULAR NETWORKS

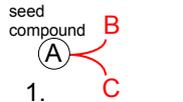


MAIN FEATURE OF ANY METABOLIC NETWORK:

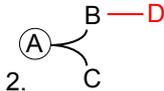
Occurrence of one reaction depends generally on the occurrence of other reactions



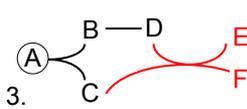
STRUCTURAL ANALYSIS OF METABOLISM BY "NETWORK EXPANSION"



First three generations of an expanding network



STARTING POINT: set of initial compounds (seed) plus set of possible reactions



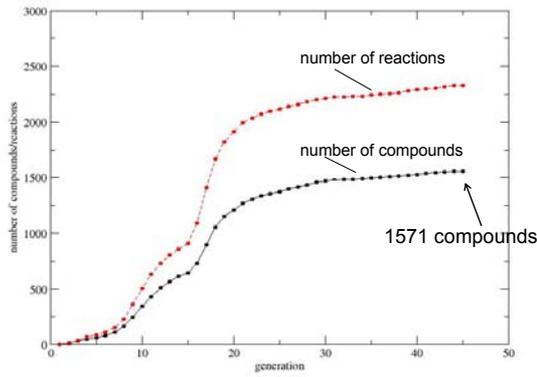
LARGE SCALE ANALYSIS

KEGG Database: 5311 reactions, 4587 compounds

etc.

EXPANSION OF ATP

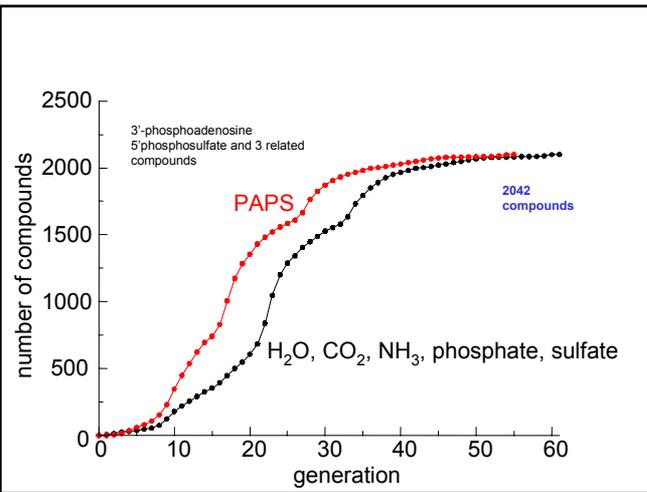
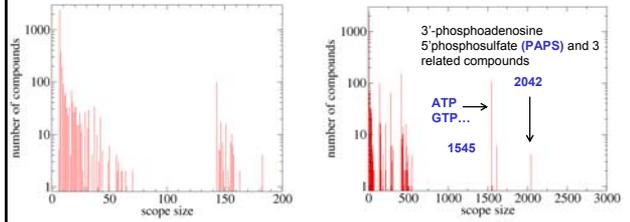
Handorf, Ebenhöh, Heinrich, 2004



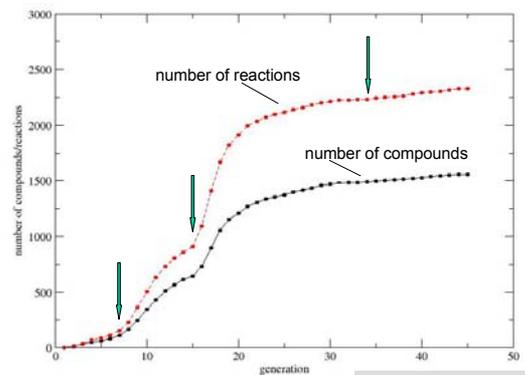
Scope Σ_A of compounds

The scope of a compound A is the set of all compounds that can be reached by an expansion starting from A

Scopes sizes of all metabolic compounds



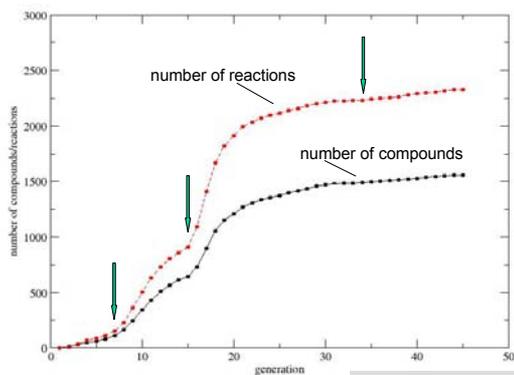
KEGG-Database: 5161 reactions and 4450 compounds



Seed compound: ATP

T. Handorf, O.Ebenhöh, R. Heinrich A. (2004), Genome Informatics Series, Vol. 15

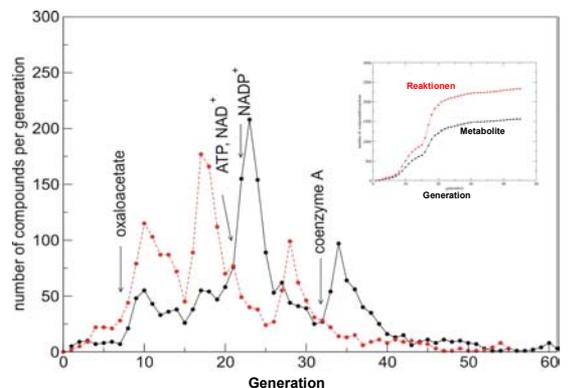
KEGG-Database: 5161 reactions and 4450 compounds

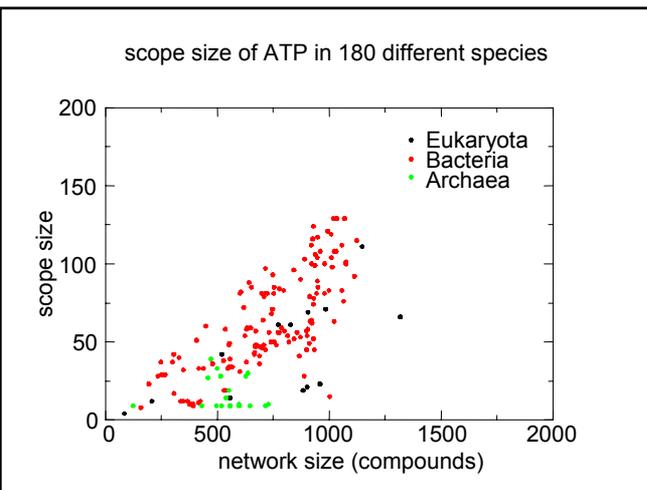
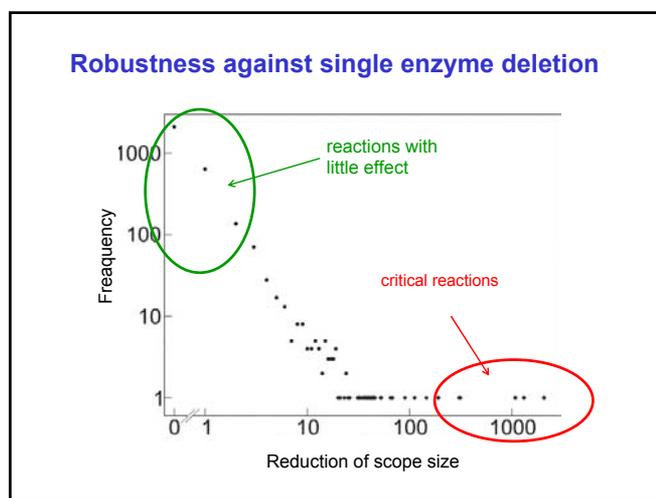
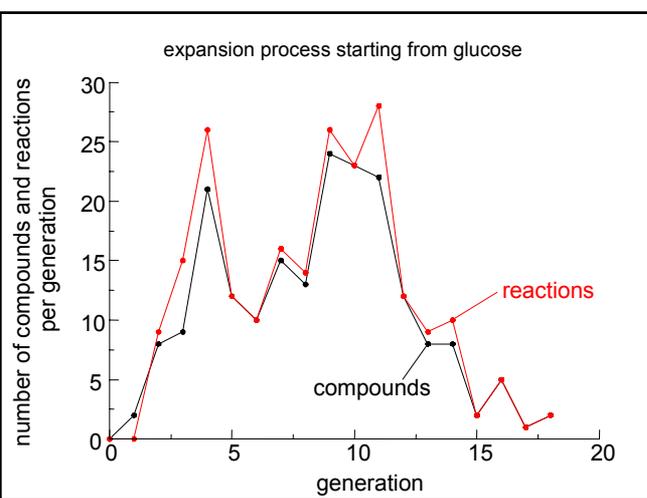


Seed compound: ATP

T. Handorf, O.Ebenhöh, R. Heinrich A. (2004), Genome Informatics Series, Vol. 15

Expansion of $\{H_2O, CO_2, NH_3, H_3PO_4, H_2SO_4\}$





Two posters on network expansion:

P-P12: Scopes: A new concept for the structural analysis of metabolic networks
(T.Handorf, O.Ebenhöh, R. Heinrich)

P-P09: Phylogenetic analysis based on structural information of metabolic networks
(O.Ebenhöh, T. Handorf, R. Heinrich)

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Roland Krüger
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Oliver Ebenhöh
Edda Klipp (MPI)
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