Self-contained RNA inhibition with trans-cleaving ribozymes

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Overview

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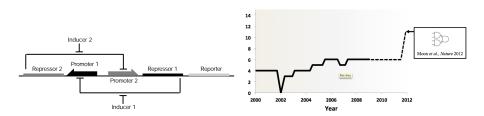
Promise of Synthetic Biology

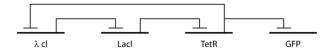
Complexity of eukaryotes $\not\approx \#$ of protein coding genes		
Oryza sativa (rice)	470 million	51,00
Gallus gallus (chicken)	1 billion	20,000-23,00
Canis familiaris (dog)	2.4 billion	19,00
Mus musculus (mouse)	2.5 billion	30,00
Homo sapiens	2.9 billion	20,000-25,000

The root of biological complexity is believed to be genetic regulation. However, the creation of novel protein regulatory elements is too difficult. Re-writers RNA-world may be the key to getting a handle on regulation.

Motivation

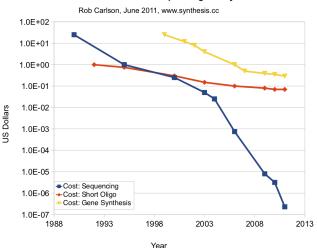
Toggle switch and repressilator in 2000, to now.





Exponential decrease in the cost of enabling technologies should result in exponential growth of circuit complexity.

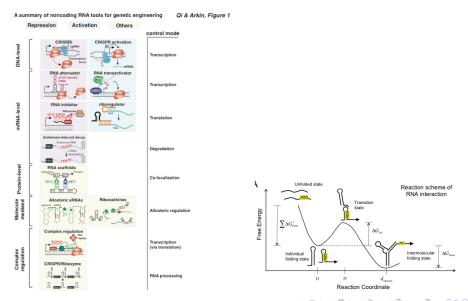
Cost Per Base of DNA Sequencing and Synthesis



Pitfalls in current promoter-repressor pair design

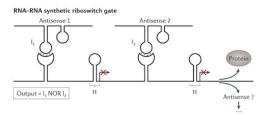
- Orthogonal Limited number of repressors (until very recently)
- Predictable Gene circuit evolves away
- Safe shRNA toxicity in gene therapy
- Reliable 40 hour toggle switch breakdown
- Designable Protein structure prediction too difficult
- Composable Unpredictable behavior when juxtaposed

Signal Transduction with RNA, possible tools



Choice of trans-cleaving hammerhead ribozymes

Self-contained mechanism of RNA degradation Composable and functionally complete: NOR gate



Base-pairing rather than aptamer coupling for ease of rational design Watson Crick base pairing dominates free energy minimization

General riboregulation model

Stochastic model - Gillespie algorithm

Algorithm 1 Gillespie

```
Inputs:
Set of M reactions; R_i = (c_i)i \in 1, ..., M
Initial population sizes, endtime
Output: Catalog of Molecular events
while \tau < endtime do
  a_0 = 0
  for i = 1 to M do
     a_i = h_i c_i, where h_i is the amount of reactant
     a_0+=a_i
  end for
  (\tau,\mu) \leftarrow P(\tau,\mu)
  updatePopulation(R_{\mu})
end while
```

trans-cleaving ribozyme model and kinetics

From Samarsky et al.

$$A_{DNA} \xrightarrow{k_{tc_o}} A_{RNA} , B_{DNA} \xrightarrow{k_{tc_b}} B_{RNA}$$

$$A_{RNA} + B_{RNA} \xrightarrow{k_1} A_{RNA} \cdot B_{RNA} \xrightarrow{k_2} A_{RNA} \cdot F_1 \cdot F_2 \xrightarrow{k_3} A_{RNA} + F_1'$$

All species undergo degradation at some rate k_{deg} .

Approximated by MM kinetics

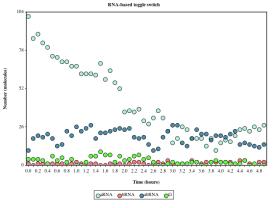
Fraction of substrate/time: $k_{obs} = k_2 \times [A_{RNA} \cdot B_{RNA}]/[B_{RNA_0}]$

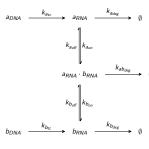
 $k_1 >> k_{-1}$, and tunable.

 $k_3 >> k_2 >> k_{deg}$ by single orders of magnitude.

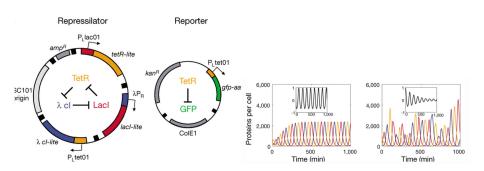
Toggle switch model riboregulation

No possible bistable point. First limitation.

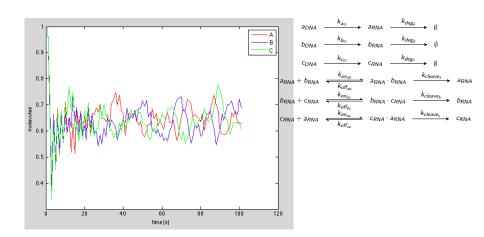




Classic Repressilator



Riboregulatory Repressilator



Repressilator Comparison

Rate of oscillation

Tunable:

- promoter strength well characterized process
- strength of binding determined by W-C Base-pairing
- cleavage rate fairly fixed
- degradation rate introduction of cleavage sites

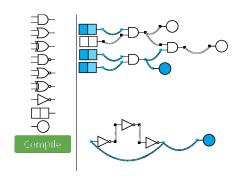
Conclusions

Character of trans-cleaving ribozymes

- Orthogonal determined by WC Base-pairing
- Predictable less stress on cells, cleaving reaction well characterized
- Safe mRNA is less persistent, and less chemically reactive
- Reliable still possible for cell to mutate
- Designable minimization of free energy based on WC Base-pairing
- Composable Unpredictable behavior when juxtaposed

Further work

Biocircuit-Design Automation, Biocompiler



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Sources

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