Lecture 5: Entropy Rules!

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DEFINITION OF A MICROSTATE

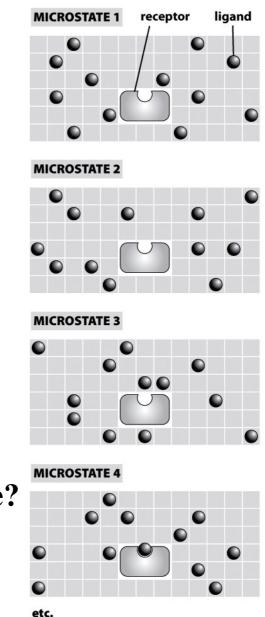
Example: Binding of RNA polymerase to a DNA target site: a simple ligandreceptor binding.

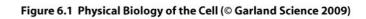
- → a lattice model of the solution
 → Ω ... number of lattice sites
- → L ... number of ligands
- → a single receptor

What is a microstate? [see Fig.]

How many possible microstates is there? (1) receptor is unoccupied: $\Omega!/[L! (\Omega - L)!]$

(2) receptor occupied: $L \rightarrow L - 1$





Another example of microstates: DNA in solution

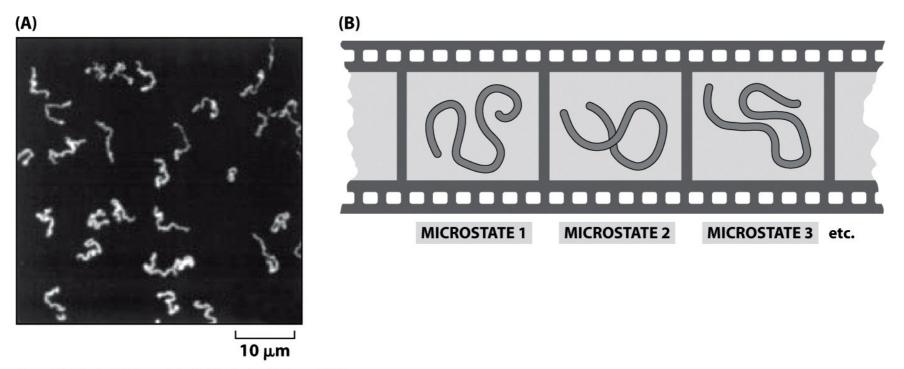
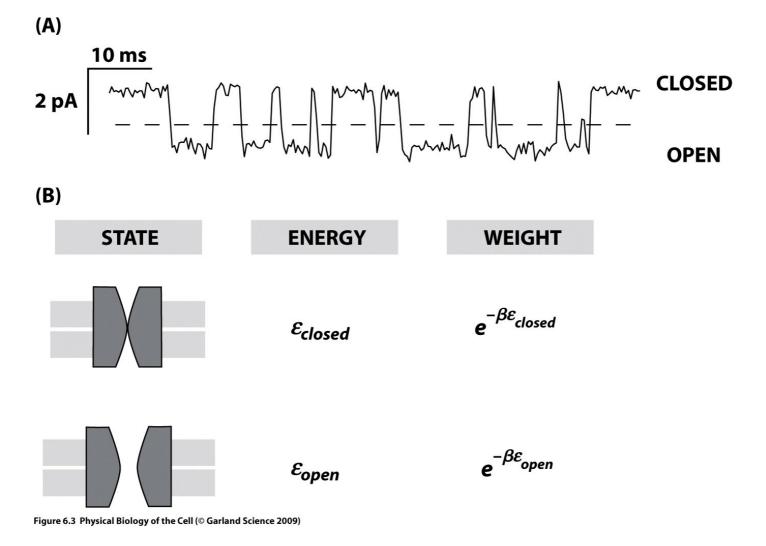


Figure 6.2 Physical Biology of the Cell (© Garland Science 2009)

(A) fluorescence microscopy image of DNA(B) individual microstates of a single DNA molecule

What is the occurrence probability of each microstate? Example: A *two-state system* (ion channel: open versus closed)



Two state system: only two microstates exist

- → the time the ion channel is open versus the time the ion channel is closed can be used to calculate the occurrence probabilities, p_{qpn} and p_{doed}
- → what determines these probabilities? energies of individual microstates, E and E and E

The probability of finding a microstate with an energy E_i is

$$\mathbf{p}(\mathbf{E}_{i}) = \exp(-\mathbf{E}_{i}/\mathbf{k}_{B}T) / \mathbf{Z}$$

The role and identity of Z:

- probabilities need to be normalized: $\sum_{i} p(E_{i}) = 1$
- Z is known as the partition function

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$$Z = \sum_{i} \exp(-E_{i}/k_{B}T) \quad \dots \text{ sum over all microstates}$$

Why do we need the probabilities and the partition function?
$$\langle E \rangle = \sum_{i} E_{i} p(E_{i}) = Z^{-1} \sum_{i} E_{i} \exp(-E_{i}/k_{B}T) \quad \dots \text{ calculate the}$$

average quantities (e.g. average energy)

Useful expressions in terms of $\beta = (k_B T)^{-1}$

$$\langle \mathbf{E} \rangle = - \mathbf{Z}^{-1} \partial \mathbf{Z} / \partial \boldsymbol{\beta} = - \partial (\ln \mathbf{Z}) / \partial \boldsymbol{\beta}$$

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Ligand-Receptor Binding:

binding of oxygen to hemoglobin binding of transcription factors to DNA

How do we calculate the probability of receptor binding?

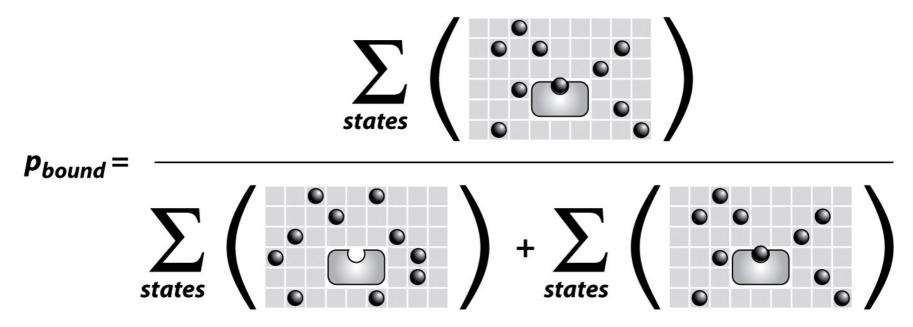


Figure 6.5 Physical Biology of the Cell (© Garland Science 2009)

There are many microstates in which the receptor is bound and many microstates in which no binding takes place: *multiplicities of the two states*

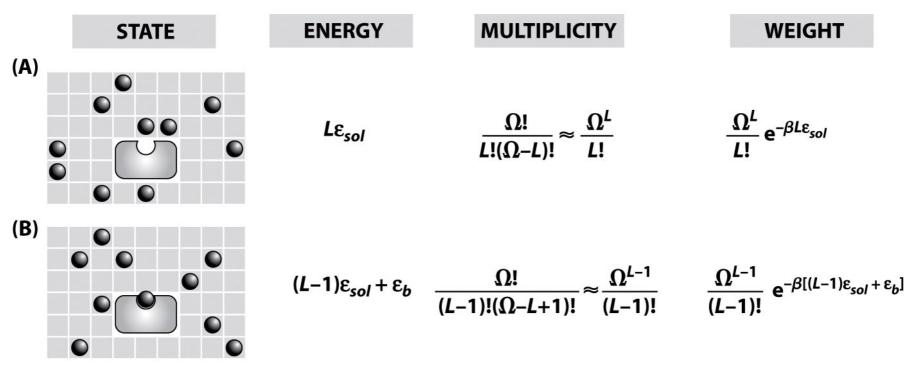


Figure 6.4 Physical Biology of the Cell (© Garland Science 2009)

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Weight for a situation in which a receptor is bound:

weight (bound) =
$$\exp(-\beta \varepsilon_{b}) \times \sum_{sd} \exp[-\beta(L-1)\varepsilon_{sd}]$$

 $\sum_{s_1} \exp[-\beta(L-1)\varepsilon_{s_1}] = \text{multiplicity } \times \exp[-\beta(L-1)\varepsilon_{s_1}]$

 $= \Omega! / [(L-1)! (\Omega - L + 1)!] \exp[-\beta(L-1)\varepsilon_{st}]$

weight (bound) = $\Omega!/[(L-1)!(\Omega-L+1)!] \exp[-\beta \varepsilon_{\beta} - \beta (L-1)\varepsilon_{st}]$

weight (unbound) = $\Omega!/[L! (\Omega - L)!] \exp[-\beta L\epsilon_{st}]$

Partition function:

 $Z(L,\Omega) =$ weight (bound) + weight (unbound)

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Useful approximation for the case $L \ll \Omega$: $\Omega!/(\Omega - L)! = \Omega^{L}$

Can be derived using Stirling's approximation: ln(N!) = N lnN - N

$$\begin{split} \ln[\Omega!/(\Omega - L)!] &= \ln\Omega! - \ln(\Omega - L)! \approx \Omega \quad \ln\Omega \quad -\Omega - (\Omega - L) \\ \ln(\Omega - L) + (\Omega - L) \approx \Omega \quad \ln\Omega - (\Omega - L) \ln(\Omega - L) = \\ \ln[\Omega^{\Omega} (\Omega - L)^{L} / (\Omega - L)^{(\Omega - L)}] \approx \ln[\Omega^{\Omega} \Omega^{L} (\Omega - L)^{L} / \Omega - L)^{\Omega}] \approx \\ \ln \Omega^{L} \end{split}$$

Thus, we can calculate p_{hand} as:

$$p_{\text{bund}} = (L/\Omega) \exp(-\beta\Delta\epsilon) / [1 + (L/\Omega) \exp(-\beta\Delta\epsilon)]$$

$$p_{\text{bund}} = (c/c_0) \exp(-\beta\Delta\epsilon) / [1 + (c/c_0) \exp(-\beta\Delta\epsilon)]$$

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Classical result: a competition between energetic and entropic contributions to the free energy: $c/c_0 = \frac{1}{2}$... half occupancy

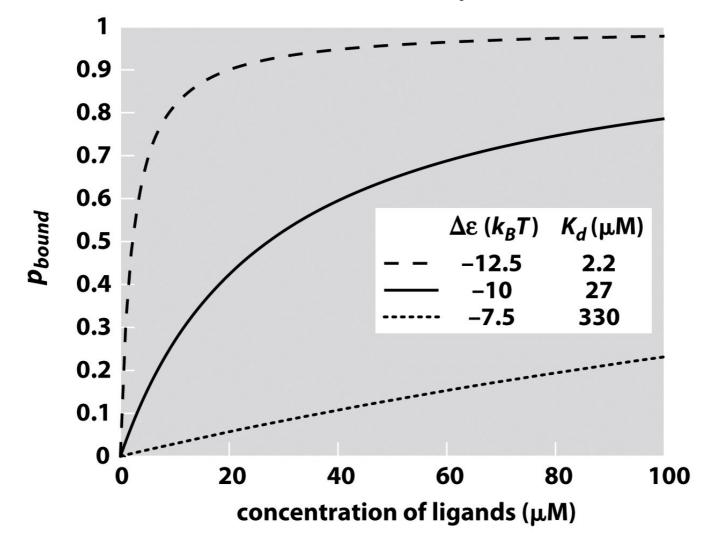
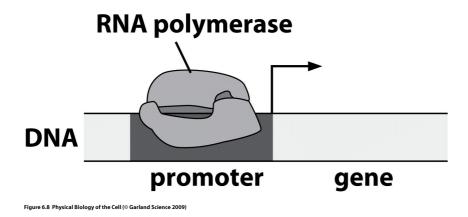


Figure 6.6 Physical Biology of the Cell (© Garland Science 2009)

Statistical Mechanics of Gene Expression: RNA polymerase binding at promotor sites

Cells can control transcription and translation: revised central dogma



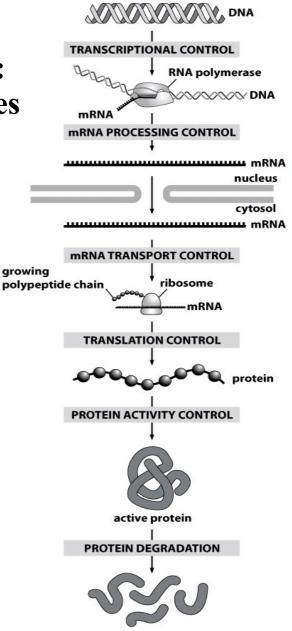


Figure 6.7 Physical Biology of the Cell (© Garland Science 2009)

Transcription: a process that begins once the polymerase Escaped the promotor and moves along the gene (part of DNA) And results in creation of mRNA molecule (*a transcript*).

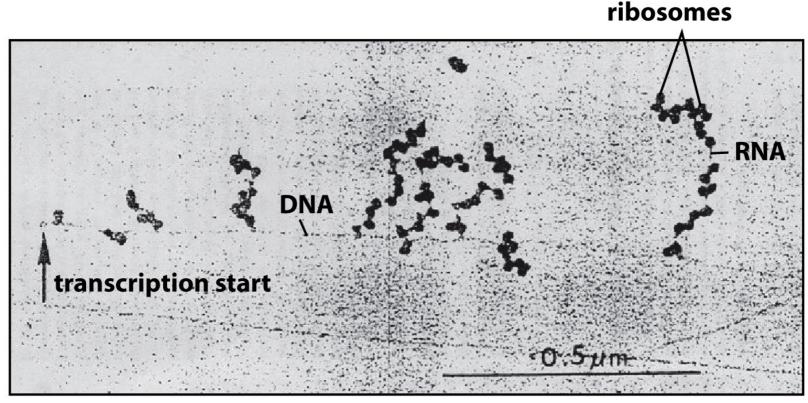


Figure 3.9 Physical Biology of the Cell (© Garland Science 2009)

a microscopy image of transcription

Experimental evidence: thousands of RNA polymerase molecules in *E. coli* bound to the DNA promoter sites

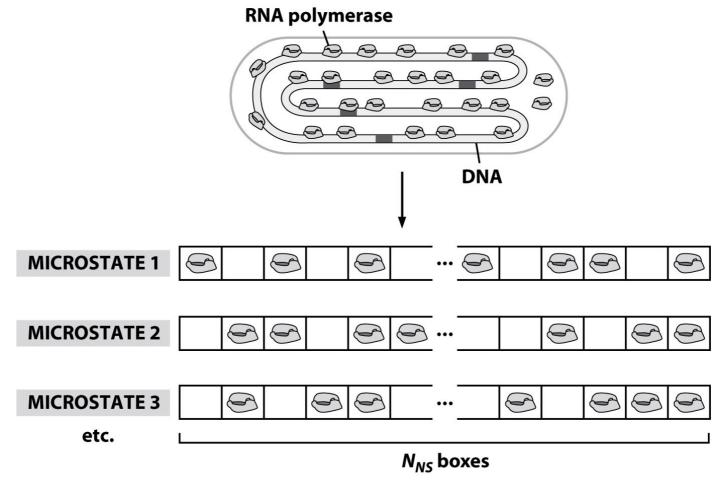


Figure 6.9 Physical Biology of the Cell (© Garland Science 2009)

Simplest model of RNA polymerase binding to DNA:

DNA modeled as: -N_N distinct boxes (NS ... non-specific sites) -P number of RNA polymerase molecules (only one molecule per non-specific DNA site)

Partial partition function for non-specific binding:

$$Z_{N}(P, N_{N}) = N_{N}!/[P!(N_{N}-P)!] \exp(-\beta P \varepsilon^{N}_{pl})$$

$$= \int_{e_{pd}} \int$$

The total partition function is a sum of two parts:

[no RNAP on promoter] + [one RNAP on promoter]

$$Z(P, N_{NS}) = Z_{NS}(P, N_{NS}) + Z_{NS}(P-1, N_{NS}) \exp(-\beta \varepsilon_{\mu l}^{S})$$

Probability of one RNAP bound to the promoter site is:

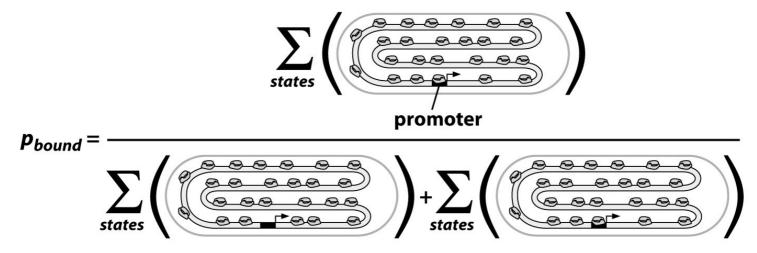


Figure 6.12 Physical Biology of the Cell (© Garland Science 2009)

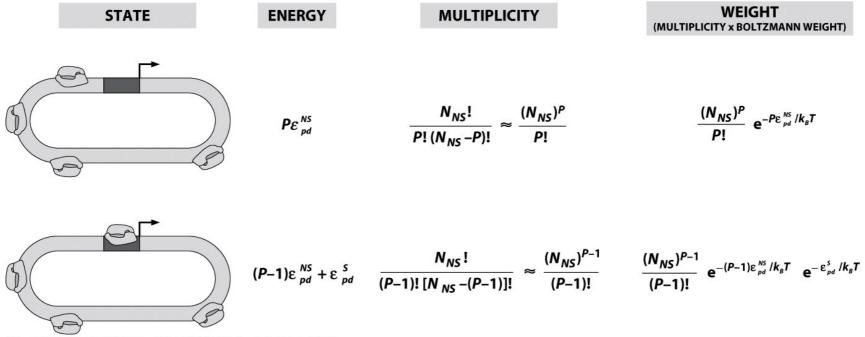


Figure 6.11 Physical Biology of the Cell (© Garland Science 2009)

$$p_{hnd} = Z_{NS}(P-1,N_{NS}) \exp(-\beta\epsilon_{\mu}^{S}) / [Z_{NS}(P,N_{NS}) + Z_{NS}(P-1,N_{NS})\exp(-\beta\epsilon_{\mu}^{S})] = [1 + N_{NS}/P \exp(\beta\Delta\epsilon_{\mu})]^{-1} \Delta\epsilon_{\mu} = \epsilon_{\mu}^{S} - \epsilon_{\mu}^{NS}$$

The more negative the difference $\Delta \varepsilon_{\mu\nu}$, the higher the probability of binding (lac P1: -2.9 k_BT; T7 A1: -8.1 k_BT).

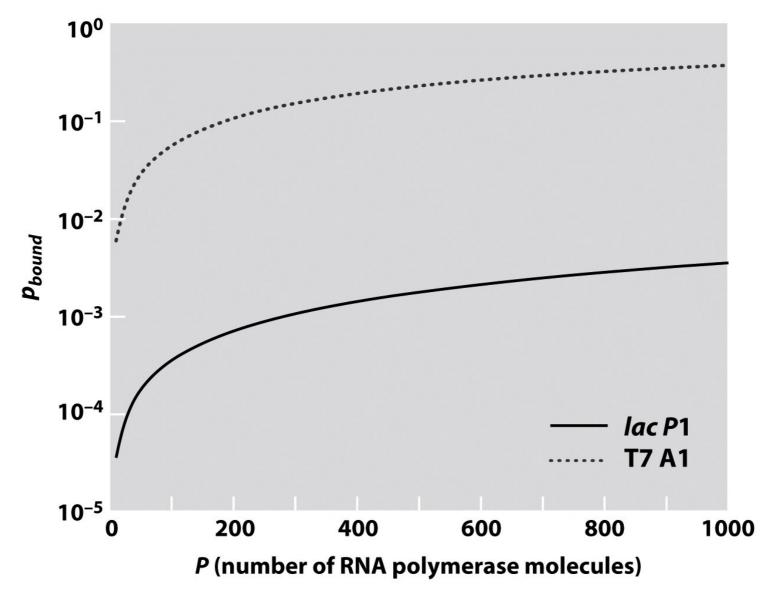


Figure 6.13 Physical Biology of the Cell (© Garland Science 2009)

Classical derivation of the Boltzmann distribution:

system + reservoir = isolated system maximal entropy principle

Fundamental idea:

probability of finding *a microstate* of the system is proportional to the number of states available to the reservoir when the system is in its specific microstate:

$$p(E_{s}^{I})/p(E_{s}^{II}) = W_{r}(E_{tt} - E_{s}^{I})/W_{r}(E_{tt} - E_{s}^{II})$$

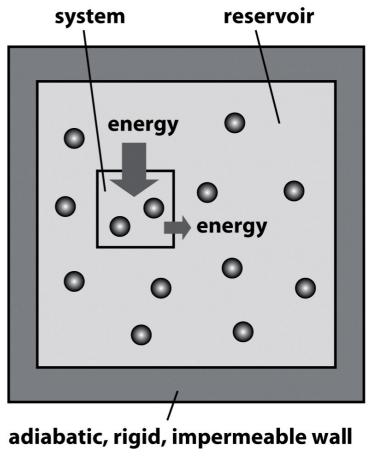


Figure 6.14 Physical Biology of the Cell (© Garland Science 2009)

$$\begin{split} & W_{ut} \left(\mathbf{E}_{tt} - \mathbf{E}_{s}^{\mathrm{I}} \right) = 1 \ \textbf{X} \ W_{r} \left(\mathbf{E}_{tt} - \mathbf{E}_{s}^{\mathrm{I}} \right) & \textbf{(A)} \quad \textbf{STATE} \quad \textbf{WEIGHT} \\ & \textbf{One state of the system X all possible} \\ & \text{States of the reservoir} \\ & S = k_{B} \ln W \\ & S_{r} \left(\mathbf{E}_{tt} - \mathbf{E}_{s} \right) = S_{r} \left(\mathbf{E}_{tt} \right) - \left(\partial S_{r} / \partial \mathbf{E} \right) \mathbf{E}_{s} \\ & \left(\partial S_{r} / \partial \mathbf{E} \right) = 1 / T \\ & \mathbf{E}_{s}^{\mathrm{I}} \left(\mathbf{E}_{tt} - \mathbf{E}_{s} \right) = S_{r} \left(\mathbf{E}_{tt} \right) - \left(\partial S_{r} / \partial \mathbf{E} \right) \mathbf{E}_{s} \\ & \left(\partial S_{r} / \partial \mathbf{E} \right) = 1 / T \\ & \mathbf{E}_{s}^{\mathrm{I}} \left(\mathbf{E}_{tt} - \mathbf{E}_{s} \right) = \exp(-\beta \mathbf{E}_{s}^{\mathrm{I}}) / \exp(-\beta \mathbf{E}_{s}^{\mathrm{II}}) & \textbf{(B)} \\ & \textbf{STATE} \quad \textbf{WEIGHT} \\ & \mathbf{E}_{s}^{\mathrm{I}} \left(\mathbf{E}_{s} - \mathbf{E}_{s} \right) = \frac{1}{2} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,1} \right) \\ & \left(\partial S_{r} / \partial \mathbf{E} \right) = 1 / T \\ & \left(\partial S_{r} / \partial \mathbf{E} \right) = 1 / T \\ & \left(\partial S_{r} / \partial \mathbf{E} \right) = \frac{1}{2} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,1} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_{n,2} \right) \\ & \mathbf{E}_{s} \left(\mathbf{E}_{s} - \beta \mathbf{E}_$$