

Branching Random Walks applied to Antibody Affinity Maturation

Irene Balelli

Université Paris 13, Laboratoire Analyse Géométrie et Applications

balelli@math.univ-paris13.fr



Abstract

Antibody Affinity Maturation is a key process in adaptive immunity. During an **Immune Response**, activated B-cells give rise to Germinal Centers in follicles, special micro-environments where they **proliferate, mutate and differentiate**. Moreover, they are submitted to powerful **selection** mechanisms to produce **high affinity** antibodies against the presented antigen. The purpose of the study is to build and analyse a mathematical model of the mutation-division-selection process of B-cells, aiming to understand how the different **biological parameters** affect the system's **functionality**. We are particularly interested in estimating via probabilistic methods **typical time-scales** to reach a specific configuration (or a set) of the traits of B-cells, as a function of the introduced mutational rule, as well as in **quantifying GCs' efficiency**.

Modeling assumptions and objectives

Hypotheses:

- **Two amino acid classes:** 0 or 1.
- B-cells and antigen's **traits:** N -length **binary strings**. $\mathcal{H}_N := \{0, 1\}^N$ is the state-space.
- **Mutation:** define the **transition probability matrix** over \mathcal{H}_N .
- **Affinity:** N -**Hamming distance** between B-cell and antigen.

Objectives:

- Evaluating the **typical time-scales of the exploration** of \mathcal{H}_N .
- Identify and study the **parameters which mostly influence** the system functionality.

1 Pure mutational models

Simple point mutations: Class switch of a randomly chosen amino acid. The resulting RW is a Simple Random Walk on the N -dimensional hypercube.

We denote by \mathcal{P} the corresponding transition probability matrix.

Multiple point mutations: i independent simple point mutations with probability $1/k$, for all $1 \leq i \leq k$, k being fixed between 1 and N .

We denote by $\mathcal{P}^{(k)}$ the corresponding transition probability matrix: $\mathcal{P}^{(k)} = \frac{1}{k} \sum_{i=1}^k \mathcal{P}^i$

1.1 Comparison of hitting times

Let \bar{d} be the **initial Hamming distance** between B-cell and antigen. We determined **explicit formulas to estimate the hitting time** to cover this distance, as a function of the mutational rule.

Mutational model	Hitting time
Simple point mutations	$H(\bar{d}) = \sum_{d=0}^{\bar{d}-1} \frac{\sum_{j=1}^{N-1-d} C_N^{d+j} + 1}{C_N^d} \sim 2^N$ for N big enough
Multiple point mutations	$\bar{T}_N^{(k)}(\bar{d}) = \sum_{l=2}^{2^N} \mu_l^{(k)} - \frac{1}{2^N C_N^{\bar{d}}} \sum_{l=2}^{2^N} \mu_l^{(k)} R_N(l, \bar{d})$

Table 1: Estimation of hitting times

2 Mutation and division

At $t = 0$ a **randomly chosen node** is labelled as **active**. At each step t , each active node chooses two of its neighbors, independently and with replacement, to become active at time $t + 1$. We are not interested here in counting how many times a node is chosen to become active. We obtain a **Branching Random Walk with coalescence** on \mathcal{H}_N .

2.1 Portion of \mathcal{H}_N covered in $\mathcal{O}(N)$

Let S_t be the active node set at time t . We estimate the size of S_T , $|S_T|$, for $T = \mathcal{O}(N)$, and depending on the allowed mutational rule.

Model	$ S_T $ in $T = \mathcal{O}(N)$
BRW- \mathcal{P}	$ S_T \geq 2^{N-r}, r > \frac{N^2 e^{-2} + N - 2}{N e^{-2} + N - 2}$
BRW- $\mathcal{P}^{(k)}$	$ S_T \geq \delta 2^N, \delta \leq 1/2$

Table 2: Portion of \mathcal{H}_N covered in $T = \mathcal{O}(N)$

3 Mutation, division and selection with multiplicity

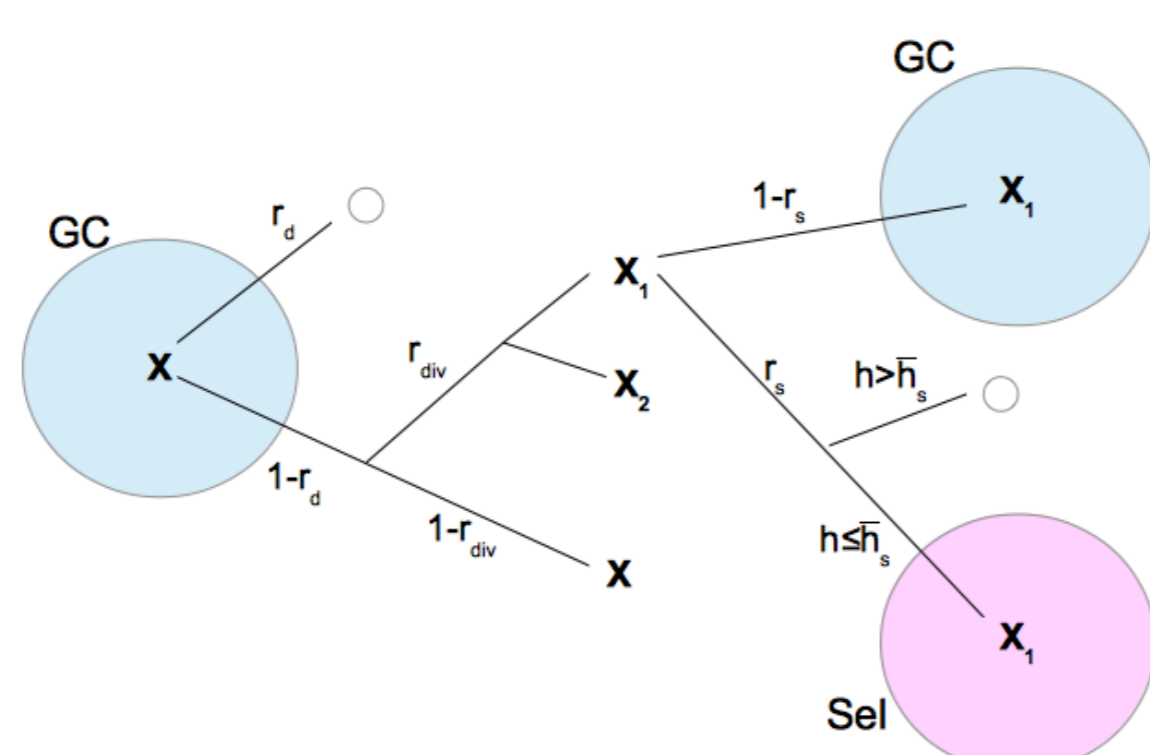


Figure 1: Schematic representation of the process. r_d = death rate; r_{div} = division rate; r_s = selection rate.

3.1 Extinction probability in the GC

We denote by η_{z_0} the **extinction probability** of the GC population, starting from z_0 naive B-cells.

Theorem:

- if $r_s \geq 1 - \frac{1}{(1-r_d)(1+r_{div})}$, then $\eta_{z_0} = 1$
- otherwise $\eta_{z_0} = \eta^{z_0} < 1$ where $\eta < 1$ is the smaller fixed point of $F(s) := p_0 + p_1 s + p_2 s^2$, with:
 - $p_0 := r_d + r_s(1-r_d)(1-r_{div}+r_{div}r_s)$
 - $p_1 := (1-r_d)(1-r_s)(1-r_{div}+2r_{div}r_s)$
 - $p_2 := r_{div}(1-r_d)(1-r_s)^2$

3.2 r_s maximizing the expectation of selected B-cells at time t

Numerical simulations evidence the existence of an optimal r_s which maximizes the expected number of selected B-cells at a given time t .

Theorem:

For any time $t \in \mathbb{N}^*$, the optimal value of r_s wrt the expectation of selected B-cells is: $r_s(t) = 1/t$.

4 Numerical Simulations

Hitting times to cover a fixed Hamming distance, referring to $\mathcal{P}^{(k)}$ (Section 1)

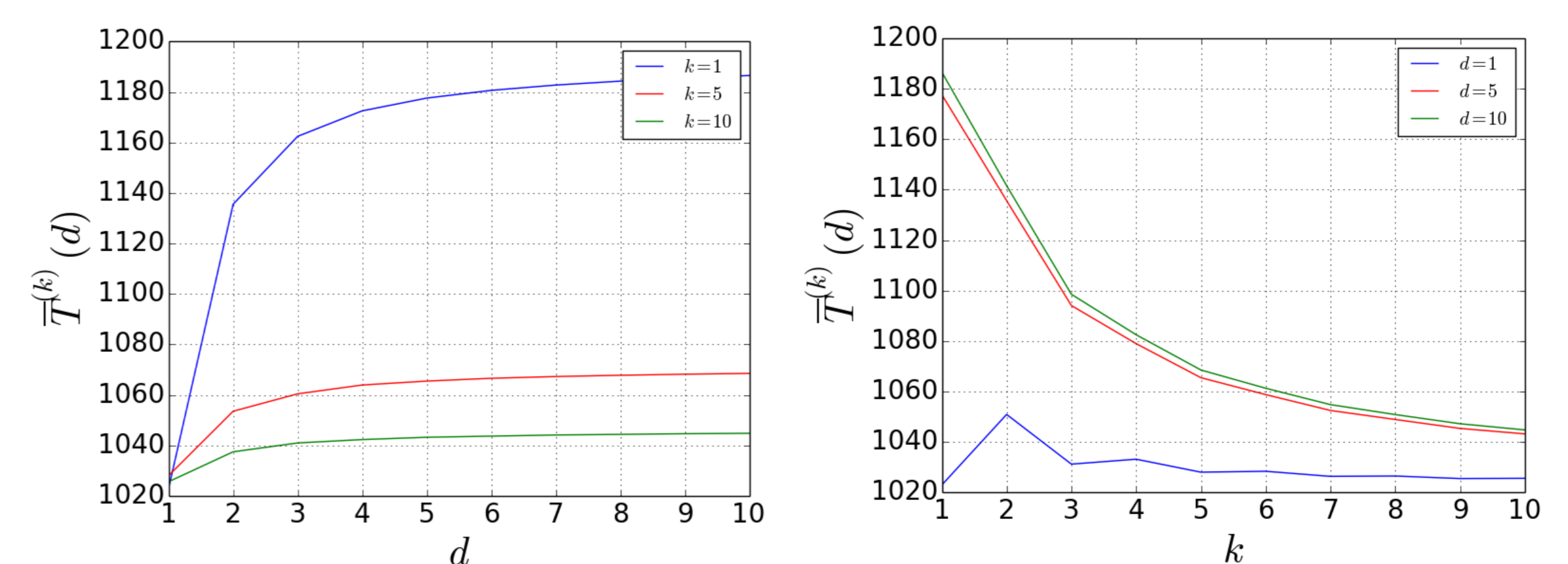


Figure 2: Dependence of $\bar{T}_N^{(k)}(d)$ on d (left) and k (right).

Evolution of the active set size for the BRW- \mathcal{P} and BRW- $\mathcal{P}^{(k)}$ (Section 2)

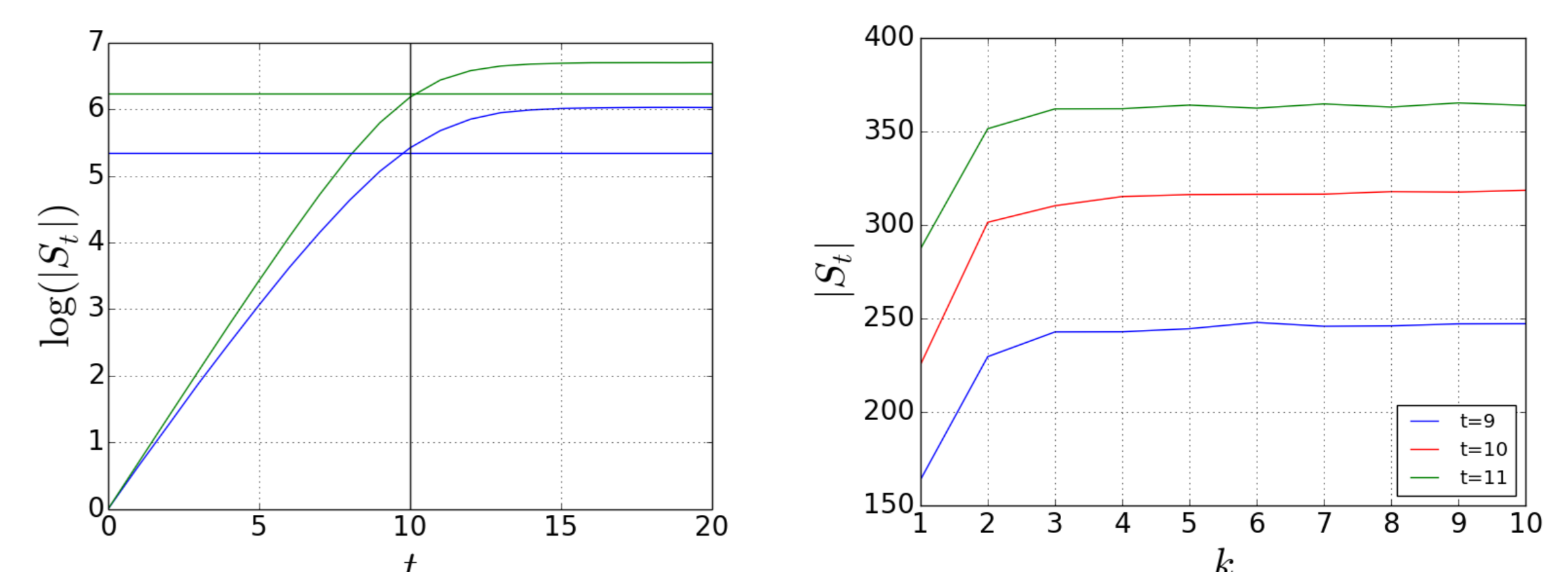


Figure 3: **Left:** Evolution of the active set size comparing the BRW- \mathcal{P} (blue) and the BRW- $\mathcal{P}^{(7)}$ (green) for $N = 10$. **Right:** Average size of $|S_t|$ after $t = N - 1, N, N + 1$ time steps, comparing the 2-BRW- $\mathcal{P}^{(k)}$ with $k \in \{1, \dots, N\}$.

Evolution of the selected pool and optimal r_s (Section 3)

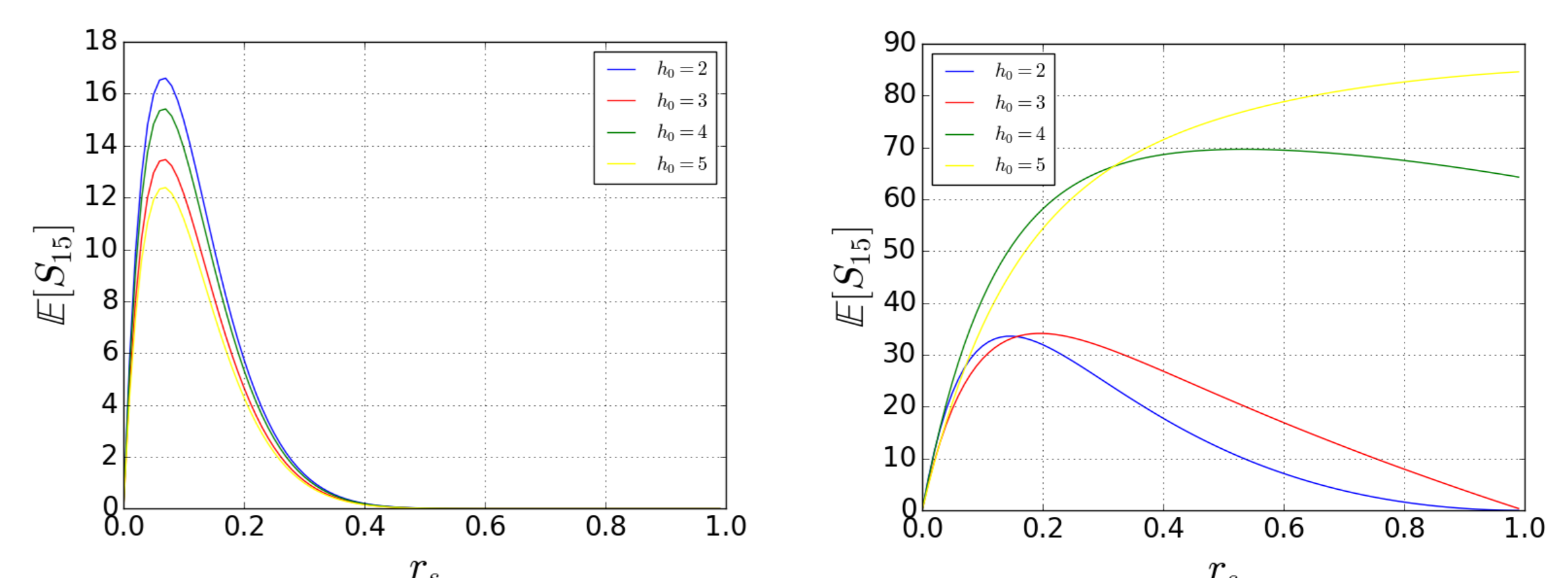


Figure 4: Dependence on r_s of the expected number of selected B-cells after 15 time steps for $N = 10$. **Left:** Model with positive-negative selection. **Right:** Model with only positive selection.

References

- [1] C. Dutta, G. Pandurangan, R. Rajaraman and S. Roche, ACM symposium 2013
- [2] J. M. Gershoni, A. Roitburd-Berman, D. D. Siman-Tov et al., BioDrugs, 2007,
- [3] T. E. Harris, The theory of branching processes, Springer-Verlag, 1963
- [4] L. Lovász, Random walks on graphs: A survey. Combinatorics, Paul erdos is eighty, 1993, vol. 2, no 1, p. 1-46.
- [5] M. Meyer-Hermann, E. Mohr, N. Pelletier et al., Cell reports, 2012, vol. 2, no 1, p. 162-174.