

(Rapid) Local Sequence Alignment  
BLAST

## BLAST: Basic Local Alignment Search Tool

BLAST is a family of **rapid approximate** local alignment algorithms[2]. BLAST is usually used to match a single DNA sequence  $S$  to a database  $D = \{D_1, \dots, D_N\}$  of DNA sequences.

Different variants of BLAST produce alignments for  $S$  and  $D$  represented in either an alphabet of nucleotides or an alphabet of amino acids (in fact,  $S$  and  $D$  may be in different alphabets!).

BLAST outputs two things:

- Good alignments between the query string  $S$  and strings from  $D$ ;
- A  $p$ -value: the estimate of probability that the reported alignment can occur by chance.

The output of BLAST is usually sorted in ascending order by  $p$ -value (i.e., the lower the probability of a chance match, the better the local alignment is).

### BLAST in a Nutshell

BLAST consists of three key procedures:

1. **Rapid search for *seed matches*.** On this stage, for each string  $D_i \in D$ , any locations that can have a "good" local match are rapidly identified. (*the specifics of this part is one of two things what makes BLAST different from other methods*).
2. **Completion of local matches.** *Seed matches* are extended to form local alignments. (*Different variations of BLAST have used different strategies for extending seed matches*).
3. **Estimation of  $p$ -values.** For each local alignment, the probability that it may occur by chance is estimated. The produced matches are sorted in ascending order by the  $p$ -value. (*This is the second "specialty" of BLAST*).

## Rapid Search for Seed Matches

**Inputs.** The problem involves the following inputs:

- alphabet  $\Sigma = \{a_1 \dots, a_M\}$ ;
- string  $S = s_1 \dots s_n$  called *query string*;
- string  $T = t_1 \dots t_m$  called *database string*;
- substitution matrix  $Score : \Sigma \times \Sigma \rightarrow \mathcal{R}$ ;
- value  $\tau$ , a *similarity threshold*, for *seed alignments*;
- an integer  $k \ll \min(m, n)$ , the length of the *seed alignments*.

**Seed alignments.** A pair of substrings  $S_i = s_i \dots s_{i+k-1}$  and  $T_j = t_j \dots t_{j+k-1}$  of length  $k$  is called a *seed alignment* iff

$$Score(S_i, T_j) = \sum_{l=0}^{k-1} Score[s_{i+l}, t_{j+l}] \geq \tau.$$

**Problem:** given the inputs above, find all pairs  $(i, j)$ , such that  $(S_i, T_j)$  is a *seed alignment*. *Do it fast!*

**Idea.**  $\Sigma$  is a constant-length alphabet. In BLAST,  $k$  - the length of a seed alignment **is a constant**.

- For  $\Sigma = \{A, T, C, G\}$  (alphabet of nucleotides),  $k$  is usually set to be in the range between 9 and 12 (a common value is 11).
- For  $\Sigma =$  the amino acid alphabet,  $k = 3$ .

BLAST uses the following key observations:

**Observation 1:** The number of all possible strings of size  $k$  in alphabet  $\Sigma$ ,  $|\Sigma|^k = M^k$  is a constant!

**Observation 2:** For each  $k$ -tuple  $V = v_1 \dots v_k$ , the set of all  $k$ -tuples  $W_1, \dots, W_s$ , such that

$$Score[V, W_i] \geq \tau$$

can be precomputed in advance in constant time!

**Note:** In fact, given  $k$ , BLAST precomputes the matrix  $Score_k$  of similarity scores between all pairs of  $k$ -tuples from  $|\Sigma|^k$ :

$$Score_k : \Sigma^k \times \Sigma^k \rightarrow \mathcal{R},$$

such that

$$Score_k[v_1 \dots v_k, w_1 \dots w_k] = \sum_{i=1}^k Score[v_i, w_i].$$

Then, based on the input value of  $\tau$ , for each  $k$ -tuple  $V$ , BLAST computes the list  $Neighbors(V)$  of all  $k$ -tuples  $W_1, \dots, W_s$ , such that  $Score_k[V, W_i] \geq \tau$ .

**Rapid Search Procedure.** String  $S = s_1 \dots s_n$  has  $n - k + 1$   $k$ -tuples:

$$S_1 = s_1 \dots s_{k-1}$$

$$S_2 = s_2 \dots s_k$$

...

$$S_{n-k+1} = s_{n-k+1} \dots s_n$$

The rapid *seed match* search proceeds as follows:

1. For each  $S_i$ , retrieve  $Neighbors(S_i)$ .
2. Construct set  $Neighbors(S) = \cup_{i=1}^{n-k+1} Neighbors(S_i)$ .
3. For each string  $V \in Neighbors(S)$ , search for all occurrences of  $V$  in  $T$ .
4. Report a *seed match* between each  $S_i$ , such that  $V \in Neighbors(S_i)$  and each  $T_j$ , such that  $T_j = V$ .

**Why this works.** **Step 3** in the procedure above is a repeated search for **exact matches** between a  $k$ -tuple  $V$  and substrings of  $T$ . This can be done in time linear in  $m$  (length of  $T$ ). Because the size of  $Neighbors(S)$  is constant (it is less than  $|\Sigma|^k$ )<sup>1</sup>, **Step 3** takes  $O(m)$  time.

In practice, the rapid matches can be done in a number of ways:

- **Suffix trees.** A suffix tree for  $T$  can be precomputed. If  $k$  is known in advance, a traversal of the suffix tree can be used to label all internal nodes with node-paths of size  $k$  (or the next closest size) with the list of leaf nodes in the subtree.

This suffix tree can be used to efficiently produce the list of seed matches, when searches for  $k$ -tuples from  $Neighbors(S)$ .

- **Indexing.** An index of  $k$ -tuple occurrences in  $T$  can be precomputed in advance and stored in an easy-to-access structure - e.g., in a *hashmap*. Given a  $k$ -tuple  $V$  from  $Neighbors(S)$ , the list of all its occurrences is retrieved in  $O(1)$  time from the index structure.

- **Aho-Corasick algorithm.** Aho-Corasick algorithm[1] solves the problem of searching for a set  $V = \{V_1, \dots, V_s\}$  of exact matches in a string  $T$  in time  $O(n + m + z)$  where,  $n$  is the length of all strings from  $V$ ,  $m$  is the length of  $T$  and  $z$  is the total number of matches found.

The algorithm works by efficiently representing the collection of strings  $V$  as a *keyword tree* with *backlinks* the provide for efficient navigation when mismatches are found.

The algorithm itself uses  $T$  to traverse the keyword tree for  $V$ . Each time a match is found, the keyword tree is navigated following a tree path. Each time, there is a mismatch, a sequence of *fail* jumps occurs.

Aho-Corasick algorithm can be used to match any string  $T$  and the keyword tree constructed from the set  $Neighbors(S)$ .

**By the numbers.** What does it take to precompute the necessary structures?

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<sup>1</sup>Although it can be a rather large number.

**Amino Acid alphabet.** For the alphabet of amino acids we have:

- $|\Sigma| = 21$  (20 amino acids plus the stop codon).
- $k = 3$ . This is the usual length used in BLAST.
- Substitution matrices used. BLOSUM62 is typically used. Other matrices in the BLOSUM family can be used. PAM family is used but not as commonly.
- Total number of combinations:  $21^3 = 9261$ .
- Size of the  $Score_k$  matrix:  $9261 \cdot 9261 = 85,766,121$  (85+ million).

**Nucleotide alphabet.** Things are a bit more "interesting":

- $|\Sigma| = 4$ .
- $k \in \{9, 10, 11, 12\}$ . Usually value is  $k = 11$ .
- Substitution matrices used. Usually,  $Score[X, X] = 5$ ,  $Score = [X, Y] = -4$  for  $X \neq Y$  is used.
- Total number of combinations:  $4^{11} = 2^{22} = 4,194,304$  (over four million).
- Size of  $Score_k$  matrix:  $4^{11} \cdot 4^{11} = 2^{44} = 17,592,186,044,416$  (over 17.5 trillion).

## Completion of Local Matches

**Step 2 of BLAST.** Once *seed matches* are found, each of them needs to be extended to the best/longest possible match.

Different BLAST versions differ on how this step is handled. In the original BLAST algorithm[2], the process of extension was as follows:

1. For each *seed match*  $(i, j)$  reported on **Step 1** of BLAST:
  - extend it in both directions for as long as the score of the new match is above the threshold  $\tau$ .
  - stop, when the match cannot be extended on either side without its score falling below  $\tau$ .
  - Report the computed match.

**Variants.** The original version did not create local alignments with gaps. Subsequent versions of BLAST improved on this process in a number of ways:

- Gapped alignments. When extending the *seed matches* use a substitution matrix **and** *indel score*  $\delta$  to score the alignments.
- Filter out seed matches. Only extend *seed matches* which show up in pairs on the same diagonal within a given number  $A$  of positions. This variant filtered out a large number of *seed matches* and improved the performance of BLAST.

## Computing $p$ -values of Alignments

**Poisson distribution.** A discrete random variable  $X$  has Poisson distribution with the mean (expected) value  $\lambda$  if the probability  $P(X = k)$  is

$$P(X = k) = \frac{\lambda^k \cdot e^{-\lambda}}{k!}$$

**Intuition.** Consider a certain, low-probability event  $\alpha$  that can occur with probability  $p$  at each moment of time. Consider a sequence of  $n$  independent trials for  $\alpha$ . Let  $X$  be a discrete random variable that counts, how many times  $\alpha$  occurs. When  $p$  is reasonable and  $n$  is relatively small, the probability that  $\alpha$  occurs exactly  $k \leq n$  times can be described exactly using the *binomial distribution*:

$$P(X = k) = p^k (1 - p)^{n-k}.$$

However, when  $p$  is *very small*, but  $n$  is *very large*, binomial distribution is not convenient to use. Poisson distribution is an approximation of the binomial distribution in such a situation.

Given  $n$  trials, the expected number of times  $\alpha$  occurs is  $np$ . If  $n$  is very large and  $p$  is very small,  $np$  may be a mid-range number. Variable  $X$  will have Poisson distribution with the expected value  $\lambda = np$ .

**Match by chance.** Consider a query string  $S$  and a database string  $T$ , for which BLAST returns a local alignment with a score  $\tau' \geq \tau$ . In general, two DNA sequences have a good local alignment *if they are related and/or serve similar purposes*. So, *what is the probability that this local alignment of  $S$  and  $T$  occurred by chance?*

**Simple example.** Let  $(i, j)$  be a pair of random positions in  $S$  and  $T$ . Let  $p$  be the probability that two letters occurring at random positions of two strings match<sup>2</sup>.

An *exact match* of length  $k$  starting at position  $i$  in  $S$  and  $j$  in  $T$ , then has the following probability of happening by random chance:

$$p' = (1 - p)p^k.$$

(Here,  $1 - p$  is responsible for the match **starting** at  $s_i$  and  $t_j$ . This means that  $s_{i-1} \neq t_{j-1}$ , which has the probability of  $1 - p$ ).

There are  $nm$  possible alignments of  $S$  and  $T$ . Therefore, the expected number of random alignments of two strings of length  $k$  in  $S$  and  $T$  is

$$\lambda = nm(1 - p)p^k.$$

The number of random alignments is a random variable with Poisson distribution with the expected value of  $\lambda$ .

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<sup>2</sup>For the nucleotide alphabet this probability, under the assumption of uniform distribution of nucleotides in the DNA strings is  $p = \frac{1}{4 \cdot 4} = \frac{1}{16}$ . For the alphabet of amino acids, this probability is  $p = \frac{1}{21 \cdot 21} = \frac{1}{441}$ .

**Altschul-Dembo-Karlin statistics.** In BLAST, the match between two substrings does not have to be exact, in order to qualify for a good local alignment, so the math is a bit more complex. The Altschul-Dembo-Karlin statistic estimates the expected number of such matches in a pair of strings  $S = s_1 \dots s_n$  and  $T = t_1 \dots t_m$  as

$$E(\tau) = K m n e^{-\lambda \cdot \tau},$$

where:

- $K$  is constant.
- $\tau$  is the similarity threshold.
- $\lambda$  is the positive root of the following equation:

$$\sum_{s \in \Sigma} \sum_{t \in \Sigma} p_s \cdot p_t \cdot e^{\lambda \cdot \text{Score}(s,t)} = 1.$$

Here  $p_s$  and  $p_t$  are the frequencies of characters  $s$  and  $t$ .

The probability that there is a match of a score greater than  $\tau$  between two "random" subsequences of  $S$  and  $T$  is

$$P = 1 - e^{E(\tau)}.$$

From these statistics, the  $p$ -values for each alignment are computed.

## References

- [1] Alfred Aho, Margaret Corasick (1975) Efficient String Matching: an Aid to Bibliographic Search, *Communications of the ACM*, Vol. 18, No. 6, pp. 333-340.
- [2] Stephen F. Altschul, Warren Gish, Webb Miller, Eugene W. Myers, David J. Lipman (1990), *Journal of Molecular Biology*, Vol. 215, pp. 403-410.