

Incorporating indel channels into average-case analysis of seed-chain-extend

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Abstract

Given a sequence s_1 of n letters drawn i.i.d. from an alphabet of size σ and a mutated substring s_2 of length $m < n$, we often want to recover the mutation history that generated s_2 from s_1 . Modern sequence aligners are widely used for this task, and many employ the seed-chain-extend heuristic with k -mer seeds. Previously, Shaw and Yu showed that optimal linear-gap cost chaining can produce a chain with $1 - O\left(\sqrt{\frac{1}{m}}\right)$ recoverability, the proportion of the mutation history that is recovered, in $O\left(mn^{2.43\theta} \log n\right)$ expected time, where $\theta < 0.206$ is the mutation rate under a substitution-only channel and s_1 is assumed to be uniformly random. However, a gap remains between theory and practice, since real genomic data includes insertions and deletions (indels), and yet seed-chain-extend remains effective. In this paper, we generalize those prior results under indel channels by introducing mathematical machinery to prove that the expected recoverability of an optimal chain is $\geq 1 - O\left((\log n)^2 n^{-C\alpha}\right)$, when the total mutation rate given by the sum of the substitution, insertion, and deletion mutation rates ($\theta_T = \theta_i + \theta_d + \theta_s$) is less than 0.206, i.e., $0 < \theta_T < 0.206$, and $\alpha = -\log_\sigma(1 - \theta_T)$.

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1 Introduction

String alignment—i.e. determining the best way to match positions of two similar strings s_1 and s_2 under some cost function—has always been one of the central primitives in computational biology, essential for downstream biological analyses like comparing relatedness of genomes or mapping sequenced reads [5, 15, 21, 26]. It is closely related to the edit distance and longest common substring (LCS) problems [4, 8, 14, 23, 29], as the choice of matching positions in the alignment implies a series of insertions, deletions, and substitutions that would be needed to transform s_2 into s_1 or vice versa. To this end, Needleman-Wunsch [20] and Smith-Waterman [27] gave dynamic programming exact solutions to the global and local alignment problems in quadratic $O(mn)$ time, where $|s_1| = n$ and $|s_2| = m$. Although more efficient algorithms exist, e.g., the Four Russians Method has a time complexity of $O(mn/\log(n))$ [1], it turns out that in the worst case, we cannot do polynomially better—Backurs and Indyk showed in 2015 that “edit distance cannot be computed in strongly subquadratic time (unless SETH is false)” [3].

Of course, the landscape of alignment algorithms and heuristics is more complicated. Although in the earliest days of bioinformatics, we had the luxury of algorithms with

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mathematically provable guarantees on accuracy and speed, the rapid growth of biological data necessitated the development of faster heuristics that do not come with those strong guarantees. BLAST [2], one of the most highly cited papers of all time [31], gives a linear-time heuristic for local alignment, at the cost of optimality, and is still to this day one of the primary workhorses of bioinformatics, whereas Smith-Waterman has been relegated to being just a subroutine within heuristic software. More broadly, there are many exact sequence alignment/edit distance algorithms that have subquadratic time complexity under certain conditions—Ukkonen’s method [30] for edit distance runs in $O(s * \min(m, n))$ in the worst case, where s is the edit distance between the two strings, and Myer’s algorithm [19] has $O(m + n + d^2)$ average-case time complexity where d is the minimum edit script between the two strings. Other than exact algorithms, there are numerous heuristics that optimize sequence alignment for specific tradeoffs [7, 11, 12, 16].

One heuristic of particular interest is “seed-chain-extend,” which is used in modern software such as Minimap 2 [18]. The seed-chain-extend heuristic has three stages. In *seeding*, k -mer ‘seeds’ are selected on both s_1 and s_2 , and shared k -mers are marked as ‘anchors’ between the two strings. Afterwards, concordant anchors are *chained* together to form the skeleton of an alignment. Lastly, the space between anchors is filled in using standard worst-case quadratic-time dynamic programming in a process known as *extension*. Seed-chain-extend empirically showcases near quasilinear runtime on the similar genomic strings it is typically applied to, but is not guaranteed to find optimal alignments.

For a long time, bioinformaticians have contented ourselves to this gap between theory and practice. In the last five years, though, theoreticians have made several new breakthroughs by defining a generative model of string evolution and revisiting average-case analysis. Analysis of string algorithms [6, 17], particularly average-case analysis [28], historically made extensive use of generating functions; however, more recent approaches instead used tail-bounds to bound bad events, more akin to the analysis of some randomized probabilistic sketches [33]. Ganesh and Sy’s 2020 breakthrough was to show that under their random mutation model, a modified dynamic programming algorithm will compute edit distance in $O(n \log n)$ time between a random string s_1 and a mutated string s_2 of near equal length [10] with high probability. However, part of what made their analysis work was the concordance of their DP algorithm with their mutation model, which meant that extending their results to more practical but sophisticated heuristics like seed-chain-extend was nontrivial.

To that end, last year, our research group made substantial progress on proving similar results for seed-chain-extend [25]. Unfortunately, the machinery and techniques we introduced in that paper were insufficiently powerful to address insertions and deletions, so we had to restrict our theoretical results to a substitution-only mutation model. Indels tend to complicate analyses and dependence structures, so most bioinformatics theoreticians either avoid directly working with them [9, 22, 24, 32], or adjust their algorithm and model to directly capture them [10]. We were able to run empirical benchmarks with indels that closely tracked our substitution-only theory (including accurate predictions of exponents), so we believed without proof the theorems were also correct for indels [25]. In this sequel, we make progress toward narrowing that remaining gap between theory and practice, generalizing our prior machinery and techniques to handle indels.

2 Strategy Overview

2.1 Challenges to analysis of seed-chain-extend

There are several difficulties in proving average-case results for seed-chain-extend. First, seed-chain-extend has three stages, and chaining and extension have different optimization objectives. When considering the overall performance of the heuristic, it is in theory possible for failures to happen at any stage. By failure here, we mean anything that leads to bad downstream events, which include most prominently not finding the correct string alignment or taking too long (e.g. quadratic time) to find that alignment. A bad chain can result from a failure in chaining, or just a bad selection of anchors in seeding. Similarly, extension failure can be a result of bad chaining, or because the extension procedure does not itself find the right alignment, despite the chaining being “good”. Here we should also note that the alignment problem under a mutation model actually diverges somewhat from the edit distance problem. The best scoring alignment corresponds to some edit distance between the strings, but arguably the “correct” alignment (at least from a biological perspective) is the one that reflects the mutations that happened to transform s_1 into s_2 . Furthermore, seed-chain-extend is known to be an approximate heuristic, and it does not guarantee a correct alignment (under either definition of correct).

To resolve these difficulties, in the prequel [25], we introduced the concept of “recoverability”, which decoupled chaining accuracy from extension accuracy. Extension is only performed in the gaps between anchors on the chain, so roughly speaking, recoverability measures how much of the correct alignment can possibly be recovered given a chain. By structuring our problem thus, we can focus on just the seeding and chaining—i.e. how good is the chain as a starting point for the extension phase. Importantly, recoverability of a chain is a theoretical measure of the goodness of the chain, as opposed to the optimization criterion used to generate the chain, such as linear-gap cost chaining.

2.2 Difficulty from indels

Unfortunately, our formal definition of recoverability in the prequel relied on the substitution-only error model. In the substitution-only regime, the correct alignment of a mutated substring s_2 to s_1 is always a diagonal line in the alignment matrix (which we termed the “homologous diagonal”). Thus, recoverability could be defined as the proportion of the homologous diagonal covered by anchors on the optimal chain and the dynamic programming (DP) extension blocks between anchors. In the presence of indels though, the correct alignment is no longer a straight diagonal. In this sequel, we thus must generalize the homologous diagonal to a “homologous path”.

In the same vein, indels also mess up the matching of indices between s_1 and s_2 . This is not only notationally very tricky to reconcile, but also make it hard to discuss the dependence structure of positions in anchors, which is necessary for the limited-dependence Chernoff bounds we used in the prequel.

Finally, having several different types of mutations raises questions of ordering and reversal that do not appear with only substitutions. One such complication is ‘no operations’ (‘no-ops’) – mutations that at positions in the original string without leaving behind a different sequence of characters – such as an deletion of a letter and an immediate reinsertion of it.

This however creates a problem with defining recoverability of a homologous path. If a letter is inserted and then deleted, or vice versa, then the alignment specified by the

homologous path will have a spurious kink that cannot be found via k -mer matching. For example, if $s_1 = \text{ACGT}$, and it is mutated to $s_2 = \text{ACGT}$ by inserting another G before the original G, and then deleting the original G, then the “correct” alignment is

```
AC-GT
|| |
ACG-T
```

which naturally will not be found by any reasonable alignment method. This is not a problem for recoverability in extension regions, as it still could be an alignment produced by the extension block (given some extension algorithm, however uncommon), but it cannot be found as an anchor, which would produce instead the “incorrect” (but lowest edit distance) alignment

```
ACGT
||||
ACGT
```

Still, in the interest of theoretical consistency, we will apply a recoverability penalty in the latter case, despite it not actually being a problem in practice, because the lower-edit-distance alignment that seed-chain-extend will find does not reflect the actual no-op mutation history.

2.3 Proof structure and motivation

The basic intuition behind the prequel [25] is that given reasonably low mutation rates, the optimal chain under linear-gap-cost chaining will be close to correct in the sense that most of the anchors will lie on the homologous diagonal. A gap between anchors can either be homologous if the anchors flanking it are both on the homologous diagonal, or non-homologous if at least one of the two anchors is off the homologous diagonal. Non-homologous gaps can lead to “breaks”, which are regions of the string where extension through the gaps does not cover the homologous diagonal, leading to a decrease in recoverability. However, with high probability, each break has size $< \sqrt{m}$ and the number of breaks is small, so the recoverability will be high. Additionally, the runtime can be bounded by extension time through all the homologous and non-homologous gaps, which are small in an optimal chain.

Roughly speaking, the intuitive reason the above strategy works is that substitutions are much more likely to break anchors than they are to create spurious anchors. So long as we remain in the regime where sufficiently many anchors can still be found, the optimal chain is close to the correct chain, and the recoverability will be high. When indel channels are added to the mix, the above logic still basically holds (though we do suddenly have pseudo-spurious anchors from indel reversals we have to deal with), but we have to carefully redefine the model and generalize recoverability. Furthermore, because of the redefinition of the model, many of the theorems and lemmas from [25] need to be updated and proved from scratch. We detail the full proof for each theorem/lemma which requires new techniques and omit the proof only when it is exactly analogous to the corresponding proof in the prequel.

The major difference we handle is a new class of anchors, defined below, termed “clipping” anchors, which lie partially on the homologous path. Clipping anchors are a strange phenomenon of indels. They contribute to recoverability in a similar way as homologous anchors: extending through gaps flanked by clipping anchors still contains the bulk majority of the path in the gap. However, they behave like spurious anchors in that the anchors themselves may not recover points on the homologous path. Clipping anchors may seem to be an anomaly but, in fact, they are the most likely anchor type with indels. Intuitively, there

are many ways in which indels can occur without breaking an anchor and the probability of these events is relatively close to the probability of a homologous anchor occurring.

By generalizing all the theorems in the prequel and bounding the number of missed points in regions of clipping anchors, we conclude that the expected recoverability of an optimal chain is $\geq 1 - O((\log n)^2 n^{-C\alpha})$.

2.4 Preliminaries

There are three mutation types used in our analysis, inspired by the mutation models of Shaw and Yu [25], and Ganesh and Sy [10]: substitutions, deletions, and insertions. Each mutation occurs independently at each position.

Before giving a formal definition of the mutation model, we define and bound the constant ρ'_i , the failure probability of the geometric distribution from which insertion lengths are sampled. Intuitively, a larger ρ'_i leads to shorter insertions.

► **Definition 1.** *Conditional on an insertion occurring at a position, the insertion length is sampled from the distribution given by $\text{Geom}(1 - \rho'_i)$. In this analysis, ρ'_i must be bounded away from 1, i.e., $0 < \rho'_i < \gamma < 1$, for any arbitrary $\gamma \in (0, 1)$.*

► **Definition 2 (Mutation Model).** *Let $S = x_1 x_2 \cdots x_{n+k-1}$ be a string where each letter x_i is sampled i.i.d. from an alphabet of size σ . The mutated substring S' is obtained by passing the substring $S[p+1 : p+m']$ through a mutation channel where for each position $p+j$, $1 \leq j \leq m'$, the input symbol $S[p+j]$ undergoes the following mutations independently:*

- **Substitution** (probability θ_s): $S[p+j]$ is replaced by a different letter, chosen uniformly from the other $\sigma - 1$ symbols.
- **Deletion** (probability θ_d): $S[p+j]$ is deleted.
- **Insertion** (probability θ_i): a random string of length $L \sim \text{Geom}(1 - \rho'_i)$ is inserted at position $p+j$.
- **Match** (probability $1 - \theta_s - \theta_d$): $S[p+j]$ is not deleted nor replaced by a different symbol.

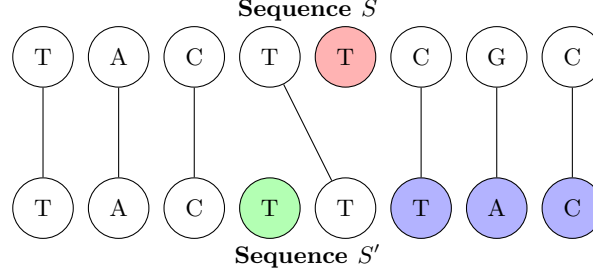
Note that if an insertion occurs then it has length at least 1. In the substitution-only case [25], the optimal alignment is the diagonal matching every position $p+j$ in S to j in S' , i.e., the set of points $\{(p+j, j) \mid 1 \leq j \leq m'\}$. When indels occur, the optimal alignment is nonobvious. We choose to define the optimal alignment as the **path history** of mutations from which S' is generated from S . We call this path history the *homologous path*, which generalizes the homologous diagonal from the prequel [25]. The homologous path, formalized below, is adapted from the canonical alignment from Ganesh and Sy [10].

► **Definition 3 (Inspired by Ganesh and Sy).** *Let S be the original length n substring with i.i.d. letters sampled from the alphabet of size σ , and S' the string resulting from passing $S[p+1, p+m']$ through the mutation channel where edits \mathcal{E} are applied.*

The homologous path P_H between S and S' is represented as a list that initially contains $(p, 0)$. Let (i, j) be the last element of P_H . We iteratively append points to P_H based on the mutations that occur at position $i+1$ in S , assuming $p+1 \leq i+1 \leq p+m'$, following \mathcal{E} :

- *If no insertion or deletion occurs at position $i+1$, i.e., a substitution occurs or there is no mutation, append the point $(i+1, j+1)$ to P_H .*
- *If an insertion of I letters occurs at position $i+1$ and no deletion occurs, append the points $(i, j+1), \dots, (i, j+I), (i+1, j+I+1)$ to P_H .*
- *If a deletion and no insertion occurs at position $i+1$, append the point $(i+1, j)$ to P_H .*
- *If both an insertion of I letters and a deletion occur at position $i+1$, append the points $(i, j+1), \dots, (i, j+I), (i+1, j+I)$ to P_H .*

Figure 1 provides an example of $S = \text{TACTTCGC}$ mutating into $S' = \text{TACTTTAC}$, following the mutation model defined above. Figure 2 shows the homologous path given the edit history.



■ **Figure 1** The match graph resulting from the mutation process that gives S' from $S = \text{TACTTCGC}$, including deletion (red), insertion (green), substitutions (blue), and matches (clear in S'). Lines between nodes represent corresponding positions between the sequences. Specifically, from left to right, an insertion of the letter T occurs at position 4, position 5 is deleted and the characters at positions 6, 7, and 8 are mutated.

► **Definition 4.** The total mutation rate is denoted by $\theta_T = \theta_s + \theta_d + \theta_i$. We will assume that $\theta_T < 0.206$.

We end this section by defining constants and bounds on constants that will be used throughout the paper.

► **Definition 5.** The constant $\alpha = -\log_\sigma(1 - \theta_T)$ represents the expected per-base contribution to the matching length between S and S' . The length of the seeds, k , is given by $k = C \log_\sigma(n)$. We will write $\log = \log_\sigma$ for short. We will always choose C such that $C\alpha < \frac{1}{2}$, which is possible since $\alpha \leq 0.161$ and we can write $C\alpha = \frac{2\alpha}{1-2\alpha} + \delta \leq 0.48 + \delta$ and δ can be chosen to be very small. The length of the generative region in S is $m' = \Omega(n^{2C\alpha+\epsilon})$, for an arbitrarily small enough $\epsilon > 0$. Note such a choice is possible since $C\alpha < 1/2$ and $n^{2C\alpha+\epsilon} < n$.

By our choice of variables, $\sigma^k = n^C$ and $(1 - \theta_T)^k = n^{-C\alpha}$, which we will use repeatedly.

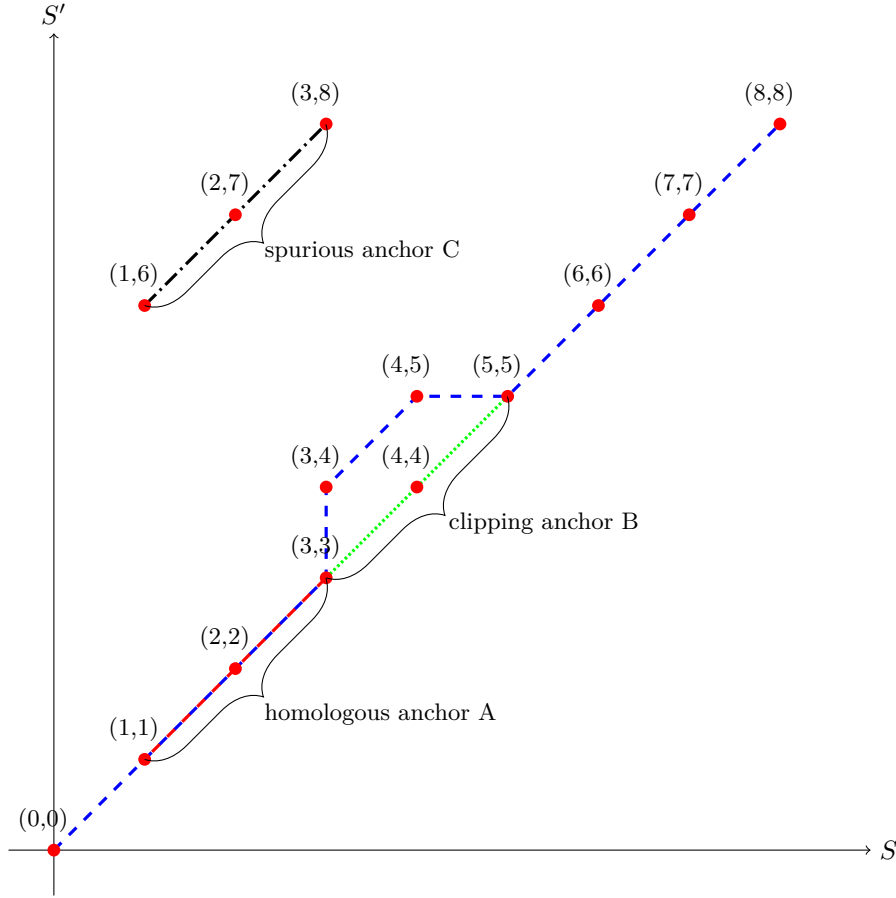
2.5 Tools

In this section, we go through definitions of tools and terms that simplify our analysis. A central concept of seed-chain-extend is an anchor, which we define formally below. An anchor is an exact k -mer match between S and S' .

► **Definition 6.** We say that an anchor of length k occurs at (i, j) , i.e., starting at position i in S and position j in S' , if $S[i : i + k - 1] = S'[j : j + k - 1]$. Note that indexing is right-inclusive. All anchors mentioned throughout the paper are of length k .

We will also find it useful to define the random variable indicating if letters at a position in S and S' match, as well as the random variable indicating if there is an anchor starting at specified positions in S and S' .

► **Definition 7.** Let $M(i, j) = \mathbf{1}\{S[i] = S'[j]\}$ be the indicator random variable detecting if S and S' share the same character at positions i and j . Define $A(i, j) = \prod_{\ell=0}^{k-1} M(i + \ell, j + \ell)$, the indicator random variable for an anchor occurring at (i, j) .



■ **Figure 2** The points along the dashed blue line make up the homologous path given the edits turning $S = \text{TACTTCGC}$ into $S' = \text{TAC TTT TAC}$ following figure 1. In this example, anchors are matching seeds of length 3. Anchor A (red dash) is a *homologous* anchor since it lies entirely on the path. Anchor B (green dash) is a *clipping* anchor since it lies partially on the path, namely, the midpoint of the anchor does not belong to the homologous path. Anchor C (black dash) is *spurious* since it lies entirely off the path.

There are three types of anchors relevant to our analysis: homologous, clipping, and spurious anchors (see figure 2 for examples of each). Visually, anchors always appear as diagonals of length k in the alignment matrix. A homologous anchor lies entirely on the homologous path (anchor A in figure 2), which implies the homologous path contains no mutations at all throughout that anchor. Unlike the substitution-only case, under indel channels, anchors can touch the homologous path at a number of points without lying entirely on it—these are called clipping anchors (anchor B in figure 2). Spurious anchors remain the same as before: anchors that lie entirely off the homologous path (anchor C in figure 2). We now formally define each anchor type.

► **Definition 8.** For notational ease, let $A = \{(i+t, j+t) \mid 0 \leq t \leq k-1\}$ and $B = \{(x, y) \in P_H \mid i \leq x \leq i+k-1 \wedge j \leq y \leq j+k-1\}$. If there exists an anchor at (i, j) , then A is the set of points belonging to that anchor, and B is the set of points on the homologous path between (inclusive) those two points. The anchor at (i, j) is **homologous** if $A = B$, **spurious** if $A \cap B = \emptyset$, and **clipping** otherwise.

We'll now define the notion of a break, which we use throughout the paper to bound recoverability. Intuitively, a break is a maximal region of the chain containing only spurious anchors. Setting the recoverability of these regions to zero allows us to lower bound the recoverability of the entire chain. Later, we'll show that the total break length is short compared to m , so their impact can be ignored.

► **Definition 9.** A **chain** C is a list of anchors $((i_1, j_1), \dots, (i_u, j_u))$ such that $i_k \leq i_{k+1}$ and $j_k \leq j_{k+1}$ for all $1 \leq k \leq u-1$. A **break** B is a maximal interval of spurious anchors in C , $((i_p, j_p), \dots, (i_q, j_q))$, where maximal means that (i_{p-1}, j_{p-1}) and (i_{q+1}, j_{q+1}) are not spurious, if these anchors exist. The break length, $L(B)$ is the total number of points on the homologous path that are covered in this break. Formally,

$$L(B) = |\{(x, y) \in P_H \mid i_p \leq x \leq i_q + k - 1 \wedge j_p \leq y \leq j_q + k - 1\}|.$$

Let $x \in (p+1, p+m')$ be an index of the generative region of S . If the character at position x is not deleted in the mutation process, then there is a unique corresponding position in S' . The character at every other position of S' is independent of $S[x]$. It is convenient to define a function f that maps each position $i \in S$ to its unique corresponding position in S' if one exists. The mapping returns the minimum y such that $(x, y) \in P_H$ and null if no such point exists. To see why this works, consider the following cases: if the character at position x in S is deleted, then there is no corresponding position, so f should return null; otherwise, let $y_0 = \min\{y : (x, y) \in P_H\}$. If there is a substitution at position x or no mutation, then the corresponding position to x is exactly y_0 . If there is an insertion and no deletion at position x , then y_0 will still be the corresponding position in S' to position x in S since insertions occur to the left of mutated positions.

► **Definition 10.** Define the function $f : \{1, \dots, |S|\} \rightarrow \{1, \dots, |S'|\} \cup \{\text{null}\}$ such that

$$f(x) = \begin{cases} \emptyset, & \text{if } x \notin (p+1, p+m') \text{ or } x \text{ is deleted,} \\ \min\{y : (x, y) \in P_H\}, & \text{otherwise.} \end{cases}$$

If in the mutation process, position $x \in (p+1, p+m')$ of S is not deleted, then either no mutation occurred at x or another character is substituted for $S[x]$. In either case, there exists a position y in S' such that $(x, y) \in P_H$, meaning that $\min\{y : (x, y) \in P_H\}$ is well-defined.

3 Methods and Bounds

3.1 Recoverability setup

The recoverability of a chain, $C = ((i_1, j_1), \dots, (i_u, j_u))$, loosely speaking, is the fraction of the homologous path P_H , that lies on the anchors in the chain and in gap extensions. We define recoverability to account for all possible alignments given by the chain, which means that we count any portion of the homologous path in a gap extension as 'recovered' since, in theory, it could be recovered by some extension algorithm. We formalize this intuition below.

► **Definition 11** ((Yu and Shaw) Recoverability). Given a chain $C = ((i_1, j_1), \dots, (i_u, j_u))$, we define the union of all possible alignments for the chain C , $\text{Align}(C)$, as:

$$\text{Align}(C) = \bigcup_{\ell=1}^u \{(i_\ell, j_\ell), \dots, (i_\ell + k - 1, j_\ell + k - 1)\} \cup \bigcup_{\ell=1}^{u-1} \text{Ext}(\ell).$$

Where $\text{Ext}(\ell) = [i_\ell + k, \dots, i_{\ell+1} - 1] \times [j_\ell + k, \dots, j_{\ell+1} - 1]$. If $i_\ell + k > i_{\ell+1} - 1$ or $j_\ell + k > j_{\ell+1} - 1$, then $\text{Ext}(\ell) = \emptyset$.

The recoverability of the chain, $R(C)$, is defined to be:

$$R(C) = \frac{|Align(C) \cap P_H|}{|P_H|}$$

► **Remark 12.** Note that the ‘ends’ of the homologous path that lie outside of the chain are unrecovered. Formally, if the chain $C = ((i_1, j_1), \dots, (i_u, j_u))$ then the set of points $\{(x, y) \in P_H \mid (x < i_1 \wedge y < j_1) \vee (x > i_\ell \wedge y > j_\ell)\}$ are not recovered.

3.2 Independence lemmas and the match graph

The match graph, defined below, is a bipartite graph that captures the dependence structure between positions on S and S' . The match graph contains edges between letters at positions i in S and $f(i)$ in S' , given that $f(i) \neq \emptyset$, i.e., edges occur between letters at corresponding positions in S and S' . When examining the independence structure of a set of random matching variables $\{M(x_1, y_1), \dots, M(x_p, y_p)\}$ where the x_i are positions on S and the y_i are positions on S' , we refer to the *induced* match graph of \mathcal{M} . Here, all edges in the original match graph remain and additional edges occur between the letters at position x_1 in S and y_1 in S' , x_2 in S and y_2 in S' , etc. (see Figure 3 for an example). Intuitively, when the induced match graph of \mathcal{M} does not have any cycles, then the random variables $M(i_\ell, j_\ell) \in \mathcal{M}$ are independent. We will use this fact to calculate the probability that different anchor types occur, and when a pair of anchor indicator variables are independent, which is used extensively in Lemma 24.

► **Definition 13.** Let $\mathcal{M} = \{M(a_1, b_1), \dots, M(a_p, b_p)\}$ be a set of matching variables where a_i are positions in S and b_i are positions in S' . The match graph induced by \mathcal{M} refers to the graph $G = (V, E)$ where the vertices V are the letters in S and S' , and the edges are given by $E = \{(x_i, y_{f(x_i)}) \mid i \in (p+1, p+m') \wedge f(x_i) \neq \emptyset\} \cup \{(x_h, y_l) \mid M(h, l) \in \mathcal{M}\}$. Note that $G(\mathcal{M})$ is bipartite.

► **Lemma 14** ((Yu and Shaw) Conditional independence in the match graph). Consider a set of random variables of the form $\mathcal{M} = \{M(i_1, j_1), \dots, M(i_\ell, j_\ell)\}$. If two vertices u, v lie in separate connected components of $G(\mathcal{M})$, then they are conditionally independent of \mathcal{M} .

The theorem below outlines a sufficient condition for match variables \mathcal{M} to be independent: no variable references corresponding positions: $(i_t, j_t) \notin P_H$, for each $t = 1, \dots, \ell$.

► **Theorem 15** ((Adapted from Yu and Shaw) Spurious match variables are independent in a cycle free match graph). The random variables

$$\mathcal{M} = \{M(i_1, j_1), M(i_2, j_2), \dots\},$$

where $f(i_\ell) \neq j_\ell$ for all $M(i_\ell, j_\ell) \in \mathcal{M}$, are independent if the induced match graph has no cycles.

► **Remark 16.** The above theorem requires that the match variables $M(i_t, j_t)$ use spurious points, i.e., $(i_t, j_t) \notin P_H$, which is captured by the condition $f(i_t) \neq j_t$. This is equivalent to the condition used in the prequel [25], where it is required that $i_t \neq j_t$, implying $(i_t, j_t) \notin P_H$ under the substitution-only mutation model. Although it is more notationally complex, the new condition can simply replace the old one, and the proof is unchanged.

► **Corollary 17.** The match graph induced by a single anchor, $A(i, j)$, has no cycles.

Proof. First note that any position in S that is deleted and any position in S' that is inserted can have degree at most 1, coming from the anchor $A(i, j)$. Any such position cannot be involved in a cycle, so we can remove all such positions from the graph. Relabel the surviving positions in S and S' as their new indices. The interval $(i, i + k - 1)$ becomes $X_{i'}$, possibly empty, and $(j, j + k - 1)$ becomes $X_{j'}$, also possibly empty. If $X_{i'}$ is empty then there cannot be any cycle that uses positions on S since each node has degree at most 1, and hence there are no cycles since the graph is bipartite; we conclude the same if $X_{j'}$ is empty. Note that the unconditioned match graph at this point appears as two sets of vertices of equal size with edges between each corresponding pair.

We now assume $X_{i'}$ and $X_{j'}$ are nonempty and that there exists a cycle. Let $i' = \min H_{i'}$ and $j' = \min X_{j'}$. First, suppose that $i' < j'$. Let x_{i+l} be the first point on S belonging to a cycle with $l \geq 0$. Then x_{i+l} has the neighbor y_{i+l} , so let the first edge of the cycle be (x_{i+l}, y_{i+l}) . Then y_{i+l} must have degree exactly 2 for there to exist a cycle and its remaining edge must be induced from the anchor. Since $i' < j'$, its neighbor lies to the left of x_{i+l} , it is some x_{i+a} where $0 \leq a < l$, contradicting the minimality of l . Similarly, if $i' > j'$, we can apply the same argument but for the largest l such that x_{i+l} is in a cycle. Thus, no cycle exists. ◀

From the above, we get the following corollary, which states that if $\{(i + l, j + l) \mid 0 \leq l \leq k - 1\}$ are spurious positions (off the homologous path), then a spurious anchor occurs at (i, j) with probability $\frac{1}{\sigma^k}$.

► **Corollary 18.** *If $f(i + \ell) \neq j + \ell$ for $\ell = 0, \dots, k - 1$, i.e., (i, j) is a possible spurious anchor, then $\Pr(A(i, j) = 1) = \frac{1}{\sigma^k}$.*

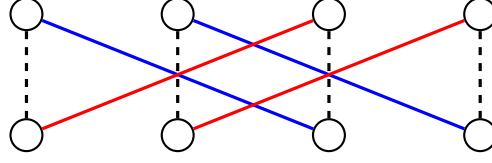
Proof. Corollary 17 shows that the random variables $M(i, j), M(i + 1, j + 1), \dots, M(i + k - 1, j + k - 1)$ are independent, since the induced match graph of $A(i, j)$ has no cycle. Since each position pair does not belong to the homologous path, $\Pr(M(i + l, j + l) = 1) = \frac{1}{\sigma}$ for $l = 0, \dots, k - 1$. Thus, $\Pr(A(i, j) = 1) = \prod_{\ell=0}^{k-1} \Pr(M(i + \ell, j + \ell) = 1) = \frac{1}{\sigma^k}$. ◀

► **Remark 19** ((Yu and Shaw) Independence lemma under substitutions). In the substitution-only regime considered by Yu and Shaw [25], the independence lemma is as follows: for $A(i, j)$ and $A(h, \ell)$, if both of the following conditions hold:

1. $|i - h| \geq k$ or $|j - \ell| \geq k$, and
2. $|i - \ell| \geq k$ or $|j - h| \geq k$,

then the induced match graph on the M variables for $A(i, j)$ and $A(h, \ell)$ has no cycles.

Intuitively, the first condition $|i - h| \geq k$ or $|j - \ell| \geq k$ ensures that the anchors do not overlap too much – the anchors' coverage can overlap on S or S' but not on both. The second condition prevents 'twisting': consider anchors $A(1, 3)$ and $A(3, 1)$ in the substitution-only mutation model as shown in figure 3. There exists a cycle going from $x_1 \rightarrow y_3 \rightarrow x_3 \rightarrow y_1 \rightarrow x_1$ where x_i represents node i on the top set of vertices and y_i represents node i on the bottom set of vertices. The biological interpretation is that under low mutation rates, x_1 and y_1 are likely to be equal, as are x_3 and y_3 . This implies that the events $x_1 = y_3$ and $x_1 = y_3$ are not independent. In the generalized lemma below, the first condition is the remains the same while the second condition is changed to account for possible insertions that complicate indices. The intuition, however, is the same. Although the general lemma is more notationally complex, the match graph in the indel case is not any more complicated because inserted and deleted positions do not have neighbors.



■ **Figure 3** Induced match graph in the substitution-only regime of an initial string of length 4 with anchors $A(1,3)$ and $A(3,1)$. These anchors violate Yu and Shaw's [25] conditions for independence and, as can be seen, there exists a cycle in the graph.

► **Lemma 20** (General independence lemma). *For $A(i,j)$ and $A(h,l)$, if both of the following conditions hold:*

1. $|i - h| \geq k$ or $|j - l| \geq k$, and
 2. $[i : i + k - 1] \cap f^{-1}([l : l + k - 1]) = \emptyset$ or $[h : h + k - 1] \cap f^{-1}([j : j + k - 1]) = \emptyset$,
- then the induced match graph on the M variables for $A(i,j)$ and $A(h,l)$ has no cycles.*

The proof formalizes the intuition that the match graph in the indel case is simpler than in the substitution-only case since any deletion removes an edge and insertions do not add edges.

Proof. We will show that inserted positions in S' and deleted positions in S have degree at most 1, so they cannot be in any cycle. Removing all such positions from the match graph for our cycle analysis will result in the same conditional match graph (except with different k , which is unimportant) as Lemma 2 from Yu and Shaw [25]. Invoking their lemma will finish our proof.

First, note that in the unconditioned match graph, any deleted position in S has no neighbor in S' , and any inserted position in S' has no neighbor in S . This follows from the fact that deleted positions have no dependents and inserted positions take values at random. There are eight total cases to consider, of which we detail one, since the remaining seven are entirely symmetric.

We now turn to the conditioned match graph. Let $|i - h| \geq k$ and $[i : i + k - 1] \cap f^{-1}([l : l + k - 1]) = \emptyset$. Take some $x \in \{i, \dots, i + k - 1\}$ that is deleted in the generation process. Since x is deleted, $x \notin f^{-1}([l : l + k - 1])$. The spacing of the anchors on S is at least k apart, so $x \notin \{h, \dots, h + k - 1\}$, meaning it can have degree at most 1. Thus, it cannot be in a cycle. The same argument, applied verbatim, shows that any deleted position in $[h : h + k - 1]$ cannot be in a cycle. Consider some $x \notin \{i, \dots, i + k - 1\} \cup \{h, \dots, h + k - 1\}$. This position has no additional edge from the anchors $A(i,j), A(h,l)$, and thus has degree at most 1. Thus, any position in S deleted during generation can be entirely removed from the match graph for cycle analysis.

The proof that positions corresponding to inserted characters can be removed for cycle analysis is the same since inserted and deleted nodes share the property that they do not have any edges in the unconditioned match graph. For completion's sake, we provide the full proof. Take some $y \in \{j, \dots, j + k - 1\}$ corresponding to an inserted position. It has no neighbors in the unconditioned match graph and so its degree can only be due to edges added by the anchors. Since $|i - h| \geq k$, there cannot be an edge between y and any $x \in \{h, \dots, h + k - 1\}$, so it has degree 1. Similarly, any inserted $y \in \{l, \dots, l + k - 1\}$ has degree at most 1. For any inserted $y \notin \{j, \dots, j + k - 1\} \cup \{l, \dots, l + k - 1\}$, it does not have any added edge from the anchors and can have degree at most 1. Thus, any inserted position in S' has degree at most 1 and cannot be in a cycle.

If a cycle exists, it must still exist if we remove all inserted positions in S' and deleted positions in S , which yields, up to extra positions with degree 1. Thus, the resulting match graph consists of positions in S and S' that correspond to each other – no position resulting from insertion in S' or deleted in S survive.

For ease, we relabel each surviving position in S and S' to its new position in the string to which it belongs. Note that $(i, i + k - 1)$ will be mapped to a new set of consecutive positions $X_{i'}$. Similarly, $(h, h + k - 1)$ will be mapped to $X_{h'}$, $(j, j + k - 1)$ to $X_{j'}$, and $(l, l + k - 1)$ to $X_{l'}$. These sets can be empty. We proceed with the proof as in the prequel: each $x \in X_{i'}$ or $X_{h'}$ has degree at most 2 since it has degree 1 in the unconditioned match graph and at most one additional edge from the two anchors, since they are disjoint on S . If a cycle exists, then it must contain both points from $X_{i'}$ and $X_{h'}$, otherwise there exists a cycle conditioned on a single anchor, which contradicts corollary 17. Assume that there exists a cycle in this match graph. Let j' be the first entry in $X_{j'}$. We work through two cases: the first is if $j' > i'$, and the second is if $j' < i'$. Note if $j' = i'$ then there are no added edges in the match graph from conditioning on $A(i, j)$, which implies all points in $X_{i'}$ have degree at most 1 and there cannot be a cycle.

Proceeding with the first case, $j' > i'$, let x_l be the leftmost position in $X_{i'}$ belonging to the cycle. This node has degree exactly 2, so its corresponding position y_l in S' must also be in the cycle. Let the first edge in the cycle be (x_l, y_l) . However, since y_l has degree exactly 2 and $j' > i'$, the remaining neighbor of y_l in S is some x_a on S that is to the left of x_l (intuitively, the anchor is right-slanted). This is a contradiction. The same argument applied for the rightmost point on $X_{i'}$ belonging to the cycle reaches a contradiction if $j' < i'$. The remaining cases follow verbatim. ◀

3.3 Anchor-count concentration bounds

The following lemma shows that any k -mer in $S[p + 1 : p + m']$ cannot expand by more than ck when generating S' with high probability. Our approach will be to show that the probability of an arbitrary k -block expanding too much is very low, and then we will take a union bound over all k -blocks in $S[p + 1 : p + m']$ to upper bound the global failure rate by $1/n$. We will find it useful to write k in terms of $\ln(n)$ in moment-generating function and Chernoff style concentration inequalities. To that end, since $k = C \log_\sigma(n) = C \ln(n) \log_\sigma(e)$, let $\beta = \log_\sigma(e)$ so that $k = C\beta \ln(n)$. The constant β does not affect the analysis but allows us to be precise.

The lemma below establishes inequalities that allow us to sum an infinite series and bound probabilities related to the expansion of a k -mer block in the generative region of S .

► **Lemma 21.** For $t_0 = \frac{1}{2} \ln(\frac{9}{1+8\gamma})$, $e^{t_0} \rho'_i < 1$ and $(1 - \theta_i) + \theta_i \left(\frac{(1 - \rho'_i)e^t}{1 - e^t \rho'_i} \right) \leq e$ for all $0 < \rho'_i < \gamma$.

Proof. From the choice of t_0 , it follows that $e^{t_0} \leq \frac{9}{1+8\gamma}$. Note that $f(x) = \frac{x}{1+\gamma(x-1)}$ is increasing for $x > 0$. We have $9 < \frac{e-1}{0.206} + 1 \leq \frac{e-1}{\theta_i} + 1$ because $\theta_i \leq \theta_T < 0.206$. Letting $A = \frac{e-1}{\theta_i} + 1$, the previous two facts give that $e^{t_0} \leq \frac{A}{1+\gamma(A-1)}$, and since $\rho'_i \leq \gamma$, we have additionally that $e^{t_0} \leq \frac{A}{1+\gamma(A-1)} \leq \frac{A}{1+\rho'_i(A-1)}$.

Multiplying both sides of this inequality by ρ'_i gives $e^{t_0} \rho'_i \leq \frac{A \rho'_i}{\rho'_i A + (1 - \rho'_i)} < 1$ since $\rho'_i < \gamma < 1$. This final inequality shows why it is necessary to bound ρ'_i away from 1.

The second inequality in the lemma follows by rearrangement and inserting the expression represented by A :

Expanding the inequality gives $(1 - \rho'_i)e^{t_0} + e^{t_0}\rho'_i A \leq A$, rearranging we get $(1 - \rho'_i)e^{t_0} \leq A(1 - \rho'_i e^{t_0})$. Simplifying yields $\frac{(1-\rho'_i)e^{t_0}}{1-\rho'_i e^{t_0}} \leq A$. Plugging in $A = \frac{e-1}{\theta_i} + 1$ and multiplying both sides by θ_i gives $\theta_i(\frac{(1-\rho'_i)e^{t_0}}{1-\rho'_i e^{t_0}}) \leq e - 1 + \theta_i$, from which we immediately get $(1 - \theta_i) + \theta_i(\frac{(1-\rho'_i)e^{t_0}}{1-\rho'_i e^{t_0}}) \leq e$. This proves the second claim. \blacktriangleleft

► **Lemma 22 (Bounded expansion (E)).** *With probability $\geq 1 - 1/n$, no k -mer in $S[p + 1 : p + m']$ has more than $\frac{1}{t_0}(\frac{2}{\beta} + 1)k$ inserted base pairs, for $t_0 = \frac{1}{2} \ln(\frac{9}{1+8\gamma})$.*

Proof. Denote the random variable representing the total insertion length at the $p + j$ -th coordinate as I_j where

$$I_j = \begin{cases} 0, & \text{with probability } 1 - \theta_i, \\ \text{Geom}(1 - \rho'_i), & \text{with probability } \theta_i, \end{cases}$$

In particular, for $\ell > 0$, $\Pr(I_j = \ell) = \theta_i(1 - \rho'_i)(\rho'_i)^{\ell-1}$. Define $Z = \sum_{j=1}^k I_j$, which represents the total insertion length, the expansion, of the first k -block. A simple Chernoff bound shows that for any $t > 0$,

$$\Pr(Z \geq c) \leq \frac{\mathbb{E}[e^{tZ}]}{e^{tc}} = \frac{\prod_{j=1}^k \mathbb{E}[e^{tI_j}]}{e^{tc}} = \frac{\mathbb{E}[e^{tI_1}]^k}{e^{tc}}.$$

Where the first equality follows since the $\{I_j\}_{j=1}^k$ are independent and the second from them being identically distributed. Choosing $t = t_0 = \frac{1}{2} \ln(\frac{9}{1+8\gamma})$, as in the previous lemma, we can calculate the moment generating function of I_1 can be directly:

$M_{I_1}(t_0) = \mathbb{E}[e^{t_0 I_1}] = (1 - \theta_i) + \sum_{j=1}^{\infty} \theta_i(1 - \rho'_i)(\rho'_i)^{j-1} e^{t_0 j} = (1 - \theta_i) + \theta_i(1 - \rho'_i)e^{t_0} \sum_{j'=0}^{\infty} (\rho'_i e^{t_0})^{j'}$. The last term is a geometric series which converges since $\rho'_i e^{t_0} < 1$ by Lemma 3.3. Thus, $M_{I_1}(t_0) = 1 - \theta_i + \theta_i \frac{(1-\rho'_i)e^{t_0}}{1-\rho'_i e^{t_0}}$, which the previous lemma (Lemma 3.3) shows is at most e .

Thus, $\Pr(Z \geq c) \leq e^{k-t_0 c}$. Choosing $c = \frac{1}{t_0}(\frac{2}{\beta} + 1)$ gives $\Pr(Z_1 \geq \frac{1}{t_0}(\frac{2}{\beta} + 1)k) \leq e^{-(t_0 \frac{1}{t_0}(\frac{2}{\beta} + 1))C\beta \ln(n)} \leq e^{-2\ln(n)} = \frac{1}{n^2}$ since $C > 1$.

Define Z_i to be the random variable denoting the expansion of the $S[p + i : p + i + k - 1]$, the i -th block in $S[p + 1 : p + m']$, formally, Z_i is the sum of the insertion lengths at each position in the k -mer $S[p + i : p + i + k - 1]$. Note that each Z_i has the same distribution as Z .

A simple union bound shows that $\Pr(\exists j : Z_j \geq \frac{1}{t_0}(\frac{2}{\beta} + 1)k) \leq (m - k + 1)\frac{1}{n^2} \leq n\frac{1}{n^2} = \frac{1}{n}$, and the result follows. \blacktriangleleft

We now show a similar lemma for contraction below: namely, sufficiently large blocks in $S[p + 1 : p + m']$ do not have too many deletions.

► **Lemma 23 (Bounded contraction (C)).** *With probability $\geq 1 - 1/n$, no ℓ -block in $S[p + 1 : p + m']$ contracts to size $\leq \frac{(1-\theta_d)\ell}{2}$, where $\ell = \frac{21k}{\beta}$.*

Proof. The proof follows similarly to the previous lemma. Define

$$X_j = \begin{cases} 0, & \text{if the } j\text{-th index of } S[p + 1 : p + m'] \text{ is deleted (with probability } \theta_d), \\ 1, & \text{otherwise.} \end{cases}$$

Note if $X_j = 1$, then the j -th index of $S[p + 1 : p + m']$ survives. Let $X = \sum_{j=1}^{\ell} X_j$ be the total number of surviving indices in the ℓ -block and set $q_d = 1 - \theta_d$ to be the survival

rate of an index. A classic Chernoff bound on the sum of i.i.d. Bernoulli random variables gives for any $0 < \delta < 1$: $\Pr(X \leq (1 - \delta)q_d\ell) \leq \exp(-\frac{\delta^2 q_d\ell}{2})$. Since $\ell = ck = c\beta C \ln(n)$ where $c = \frac{21}{\beta}$ gives $\frac{\delta^2}{2}q_d c_0 C \beta \geq \frac{1}{8}(.794)21 > 2$, when $\delta = 1/2$ and $\theta_d \leq \theta_T \leq 0.206$. Thus, $\Pr(X \leq \frac{1}{2}q_d\ell) \leq \exp(-2\ln(n)) = \frac{1}{n^2}$.

A union bound over all ℓ -blocks in $S[p+1 : p+m']$ gives:

$$\Pr(\exists \ell\text{-block shrunk to less than } \frac{(1-\theta_d)\ell}{2}) \leq n/n^2 = \frac{1}{n}.$$

◀

We will refer to the space where the bounded contraction and expansion lemmas are jointly satisfied as **EC**.

► **Lemma 24 ((EC)).** $\mathbb{E}(N_S^2) \leq \mathbb{E}(N_S)^2 + 2T_0 k^2 \frac{mn}{\sigma^k}$ where $T_0 = \max(\frac{1}{t_0}(\frac{2}{\beta} + 1)\frac{2}{1-\theta_d}\frac{21}{\beta}, 4)$. Thus, $\text{var}(N_S) \leq T_0 k^2 \frac{mn}{\sigma^k}$

Proof. Let $S_p = \{(i, j) \in S \times S' \mid [(i, j), \dots, (i+k-1, j+k-1)] \cap P_H = \emptyset\}$ be the set of all positions where an anchor at that position does not intersect the homologous path.

Define $B_k(i, j) = \{(h, l) \in S_p \mid |h-i| \leq k \cap |l-j| \leq k\}$, and

$$P_k(i, j) = \{(h, l) \in S_p : f^{-1}([l : l+k-1]) \cap [i : i+k-1] \neq \emptyset, \\ f^{-1}([j : j+k-1]) \cap [h : h+k-1] \neq \emptyset\}.$$

By Lemma 20, for any $(h, l) \notin B_k(i, j) \cup P_k(i, j)$, $A(i, j)$ and $A(h, l)$ are independent. Then, $N_S = \sum_{(i, j) \in S_p} A(i, j)$. We calculate the variance as follows:

$$\begin{aligned} N_S^2 &= \sum_{(h, l) \in S_p} \sum_{(i, j) \in S_p} A(i, j) A(h, l) \\ &= \underbrace{\sum_{(h, l) \in S_p \setminus (B_k(i, j) \cup P_k(i, j))} \sum_{(i, j) \in S_p} A(i, j) A(h, l)}_{S_1} + \underbrace{\sum_{(h, l) \in S_p \cap B_k(i, j)} \sum_{(i, j) \in S_p} A(i, j) A(h, l)}_{S_2} \\ &\quad + \underbrace{\sum_{(h, l) \in (S_p \cap P_k(i, j)) \setminus B_k(i, j)} \sum_{(i, j) \in S_p} A(i, j) A(h, l)}_{S_3}. \end{aligned}$$

Dealing first with S_1 : by the independence lemma (Lemma 20), $A(h, l), A(i, j)$ are independent, so:

$$\begin{aligned} \mathbb{E}(S_1) &= \sum_{(h, l) \in S_p \setminus (B_k(i, j) \cup P_k(i, j))} \sum_{(i, j) \in S_p} \mathbb{E}(A(i, j)) \mathbb{E}(A(h, l)) \\ &\leq \sum_{(h, l) \in S_p} \sum_{(i, j) \in S_p} \mathbb{E}(A(i, j)) \mathbb{E}(A(h, l)) = \mathbb{E}(N_S)^2. \end{aligned}$$

Note $\mathbb{E}(A(h, l)A(i, j)) \leq \mathbb{E}(A(i, j))$ since the anchor random variables take values in $\{0, 1\}$. Using that $|B_k(i, j)| \leq 4k^2$, $\Pr(A(i, j) = 1) = \frac{1}{\sigma^k}$ for $(i, j) \in S_p$ from Corollary 18, and the naive bound $|S_p| \leq mn$, we get:

$$\mathbb{E}(S_2) \leq \sum_{(h, l) \in S_p \cap B_k(i, j)} \sum_{(i, j) \in S_p} \mathbb{E}(A(i, j)) \leq 4k^2 \frac{mn}{\sigma^k}.$$

Lastly, we handle the S_3 term. Working under EC, the region $S[i : i + k - 1]$ can expand to have at most $(\frac{1}{t_0}(\frac{2}{\beta} + 1) + 1)k + 2k$ corresponding positions l on S' . Note the additional $2k$ comes from k -mers that start before and after the corresponding region but still intersect it. The same argument shows that there are at most $(\frac{1}{t_0}(\frac{2}{\beta} + 1) + 1)k + 2k$ positions j that correspond to $S[h : h + k - 1]$. Thus, $|P_k(i, j)| \leq ((\frac{2}{\beta} + 1) + 1)k + 2k)^2 = (\frac{1}{t_0}(\frac{2}{\beta} + 1) + 3)^2 k^2$. For simplicity, let the constant $T_0 = 2 \max((\frac{1}{t_0}(\frac{2}{\beta} + 1) + 3)^2, 4)$.

This yields,

$$\mathbb{E}(N_S^2) \leq \mathbb{E}(N_S)^2 + T_0 k^2 \frac{mn}{\sigma^k}.$$

From which it immediately follows that $\text{var}(N_S) \leq T_0 k^2 \frac{mn}{\sigma^k}$. ◀

The previous lemma shows that allowing indels changes the variance of N_S , the number of spurious anchors, by at most a constant compared to the substitution-only case. The below lemma uses the conditional bounded variance of N_S along with the high likelihood of being in the EC space, to bound the number of spurious anchors w.h.p. The spurious anchor bound below is exactly the same as in the prequel up to a constant, which makes no asymptotic difference.

► **Lemma 25 (F1).** *With probability at least $1 - \frac{3}{n}$, the number of spurious anchors is*

$$\leq n^{2-C} + \sqrt{T_0} C \log(n) n^{\frac{3-C}{2}}$$

Mathematically,

$$\Pr\left(N_S \geq n^{2-C} + \sqrt{T_0} C \log(n) n^{\frac{3-C}{2}}\right) \leq \frac{3}{n}$$

Proof. Call the event $X = \{N_S \geq n^{2-C} + \sqrt{T_0} C \log(n) n^{\frac{3-C}{2}}\}$, i.e., the event that there are more spurious anchors than the amount given by the expression. By the law of total probability:

$$\begin{aligned} \Pr(X) &= \Pr(X \mid EC) \Pr(EC) + \Pr(X \mid EC^c) \Pr(EC^c) \\ &\leq \Pr(X \mid EC) + \Pr(EC^c). \end{aligned}$$

We first bound $\Pr(X \mid EC) \Pr(EC) \leq \Pr(X \mid EC)$. To do this, let us first show that $\frac{mn}{\sigma^k} \leq n^{2-C}$. Recall that $k = C \log(n)$. This follows from noting that $\log(m) \leq \log(n)$, since $m \leq n$ under EC, and rearranging, $(C - 1) \log(n) + \log(m) \leq C \log(n) = k$, from which we get $\log(mn) - k \leq (2 - C) \log(n)$ and, finally, $\log(\frac{mn}{\sigma^k}) \leq \log(n^{2-C})$, which gives the inequality.

Now, by Chebyshev, $\Pr(N_S \geq \mathbb{E}(N_S) + \sqrt{n \text{var}(N_S)}) \leq \frac{1}{n}$. We can upper bound $\Pr(X \mid EC)$ by using the variance bound of N_S given by Lemma 24 and the previous inequality. This yields: $\text{var}(N_S) \leq T_0 k^2 \frac{mn}{\sigma^k}$, so $\frac{1}{n} \geq \Pr(N_S \geq \mathbb{E}(N_S) + \sqrt{n \text{var}(N_S)}) \geq \Pr(N_S \geq \mathbb{E}(N_S) + \sqrt{T_0} C \log(n) n^{\frac{3-C}{2}})$. Finally, bounding $\mathbb{E}(N_S) \leq \frac{mn}{\sigma^k}$ and using our previously shown inequality, we get the full result that $\Pr(X \mid EC) \leq \frac{1}{n}$.

For the second term, $\Pr(EC^c) \leq \frac{2}{n}$, by a simple union bound from Lemma 22 and Lemma 23.

Combining terms, $\Pr\left(N_S \geq n^{2-C} + \sqrt{T_0} C \log(n) n^{\frac{3-C}{2}}\right) \leq \frac{3}{n}$. ◀

We finish this section by upper bounding the expected number of clipping anchors, which allows us to upper bound the expected number of missed points due to clipping anchors in the chain. We start by providing the general lemma below, which bounds the number of clipping anchors in terms of the cardinality of the homologous path. In the following corollary, we specialize the below lemma to the EC case.

► **Lemma 26.** *The expected number of clipping anchors, N_C , is at most $|P_H|k(1 - \theta_T)^k$.*

Proof. Consider any point $(i, j) \in P_H$. There are exactly k possible anchors containing that point – anchors that contain it at k different distances from the start of the anchor. Let $N_C(i, j)$ be the number of clipping anchors contributed for the point (i, j) . Then $\mathbb{E}(N_C(i, j)) \leq \sum_{l=1}^k \Pr(A(i - l + 1, j - l + 1) = 1) \leq k(1 - \theta_T)^k$. Thus, $\mathbb{E}(N_C) = \sum_{(i,j) \in P_H} \mathbb{E}(N_C(i, j)) \leq |P_H|k(1 - \theta_T)^k$. ◀

► **Corollary 27 ((EC)).** *The expected number of clipping anchors is at most $O(mk(1 - \theta_T)^k)$.*

Proof. Under EC, each block of k positions in S contributes at most $(\frac{1}{t_0}(\frac{2}{\beta} + 1) + 1)k$ points to the homologous path P_H by Lemma 22. Then there can be at most $(\frac{1}{t_0}(\frac{2}{\beta} + 1) + 1)m = O(m)$ points in P_H . Applying Lemma 26 gives the result. ◀

In the next section, we show that there are no long homologous gaps in the generative region of S and use this fact to bound break lengths with the help of the expansion-contraction lemma.

3.4 Bounding break lengths

A homologous gap is defined to be a region $[a, b]$ in the generative portion of S , $\mathcal{G}(S)$, for which there are no homologous anchors. Equivalently, a homologous gap is a region $\mathcal{G}(S)$ for which every k -mer contains a mutation. We begin by bounding the length of any homologous gap by establishing concentration bounds on homologous anchors in this k -dependence case similar to the prequel [13, 25]. The lemmas in this substring make use of the generative region of S , $\mathcal{G}(S)$, which has length m' but the final inequalities are in terms of $|S'| = m$. This is fine, since under EC, $c_a|\mathcal{G}(S)| \leq |S'| \leq c_b|\mathcal{G}(S)|$, so the two lengths are equivalent up to a constant. We will represent fixed constants with variants of c ; the exact values do not make a difference in the analysis, except that they remain fixed.

► **Theorem 28.** *(Yu and Shaw) Suppose we have $X = \sum_{a \in A} \text{Bernoulli}_a(q)$ for some $0 < q < 1$. A proper cover of A is a family of subsets $\{A_i\}_{i \in I}$ such that all random variables in $A_i \subset A$ are independent and $\bigcup_{i \in I} A_i = A$. Let $\chi(\mathcal{A})$ be the minimum size of the cover, $|I|$, over all possible proper covers. Then for $t \geq 0$,*

$$\Pr(X \leq \mathbb{E}X - t) \leq \exp\left(-\frac{8t^2}{25|A|\chi(\mathcal{A})q}\right).$$

► **Lemma 29.** *$\Pr(N_H \leq m'(1 - \theta_T)^k - t) \leq \exp(-\frac{8t^2}{25mk(1 - \theta_T)^k})$*

Proof. We use the previous theorem with $q = (1 - \theta_T)^k$. Let $A(i)$ denote the random variable taking on the value of 1 if there is a homologous k -mer match starting from index i of S . Then each set $A_j = \{A(j), A(j + k), \dots\}$ contains mutually independent random variables. Note $A = \bigcup_{i \in I} A_i$, and thus the A_i form a partition of A . This implies that $\chi(\mathcal{A}) \leq k$ and the result follows. ◀

We can now apply this lemma to bound homologous gaps in the generative region of S exactly as in the prequel.

► **Lemma 30.** *For any interval consisting of l k -mers in $\mathcal{G}(S)$, the probability that all l homologous anchors are 0 is upper bounded by*

$$\exp\left(-\frac{8l(1-\theta_T)^k}{25k}\right).$$

Proof. For any interval on $\mathcal{G}(S)$ the letters and mutation processes occur with an identical distribution as a $l+k-1$ version of $\mathcal{G}(S)$. Thus, using the previous lemma with $t = l(1-\theta_T)^k$ yields the result. ◀

Working in $\mathcal{G}(S)$, we can now apply the previous lemma to bound homologous gaps w.h.p. exactly as in the prequel.

► **Lemma 31 (F2).** *With probability $\geq 1 - \frac{1}{n}$, no homologous gap in $\mathcal{G}(S)$ has size greater than*

$$g(n) = \frac{50k}{8(1-\theta_T)^k} \ln(n) = \frac{C \cdot 50}{8} \log(n) \ln(n) \cdot n^{C\alpha}$$

plus a small $C \log n$ term we will ignore because it is small asymptotically.

We now turn to bounding the length of any break in the chain w.h.p. The logic will be to show that w.h.p. an optimal chain cannot be completely spurious, breaks flanked on both sides have length at most $O(\sqrt{m})$, and breaks flanked on a single side - breaks at the start or end of the chain - have length at most $O(\sqrt{m})$. Combining each piece, we will conclude that w.h.p. any break has length at most $O(\sqrt{m})$.

► **Lemma 32 ((EC+F1+F2)).** *With probability at least $1 - 6/n$, every optimal chain $((i_1, j_1), \dots, (i_u, j_u))$ contains at least one homologous anchor or clipping anchor, provided n is sufficiently large.*

Proof. It suffices to show that for large n , the chain with only homologous anchors is higher scoring than any completely spurious chain. This is shown in Supplemental Lemma S4 of Yu and Shaw. ◀

The following lemma shows that any break flanked on both sides has length $< cm^{1/2}$ with high probability.

► **Lemma 33 ((EC+F1+F2)).** *A break flanked by two at least clipping anchors (clipping or homologous), has length $< c_0\sqrt{m}$ with probability at least $1 - 6/n$ (for sufficiently large n)*

Proof. We first show that the portion of the homologous path in a break that lies below the left anchor is at most $O(k)$. Call the left flanking anchor (i_l, j_l) . Since the left flanking anchor is at least clipping, it contains some point on the homologous path. In the EC space, the homologous path must move $O(k)$ indices in S' for every k indices traversed in S and vice-versa, i.e., the homologous path has average slope $O(1)$. Thus, there is some index $i \in S$ for which $i - i_l \leq ck$ and $f(i) \geq j_l$. Applying the same argument to the right flanking anchor shows that, again, at most $O(k)$ points are missed in the break region above the right flank on the homologous path. The remainder of the proof ignores these terms since they contribute $O(k)$ to the break length, which is asymptotically negligible.

As in the previous lemma, the exact proof from Yu and Shaw shows that if the break corresponded to a region of length $\geq m^{1/2}$ in $\mathcal{G}(S)$, then all spurious anchors could be

replaced by homologous anchors to obtain a higher scoring chain. Thus, the break in the generative region of S cannot be more than $m^{1/2}$. Under EC, any region of length $m^{1/2}$ cannot correspond to a region of length more than $cm^{1/2}$ in S' and so the total number of missed points is at most $2(m^{1/2} + cm^{1/2}) = c_0 m^{1/2}$, absorbing the constants into c_0 . ◀

The following lemma from the prequel holds with the same exact proof and shows that a break flanked on a single side, i.e. a break at the very start or very end of the chain, has length $< cm^{1/2}$ w.h.p. for a fixed constant $c > 0$.

► **Lemma 34** ((EC+F1+F2)). *A break flanked on one side by an at least clipping anchor has length $< c\sqrt{m}$ with probability at least $1 - 6/n$ (for sufficiently large n)*

We now conclude the main result of this section: the break length lemma. By combining the previous cases of where a break occurs, we show that with probability $\geq 1 - \frac{6}{n}$, any break length is $< cm^{1/2}$ in an optimal chain.

► **Lemma 35** (Break Length Lemma (EC+F1+F2)). *Let $g(n) = \frac{C}{8} \log(n) \ln(n) n^{C\alpha}$, and set $\zeta = \frac{1}{6g(n)}$. Suppose $C > \min(3, \frac{2}{1-2\alpha})$, and that $m = \Omega(n^{2C\alpha+\varepsilon})$ for some $\varepsilon > 0$. Then, for all sufficiently large n , with probability at least $1 - 6/n$ no optimal chain contains a break of length $\geq c\sqrt{m}$.*

Proof. Under (EC+F1+F2), any optimal chain contains at least one clipping anchor or homologous anchor meaning that any break is flanked on one or both sides. Consider a break flanked on both sides. Applying Lemma 33 shows that this break has length $< cm^{1/2}$. If the break is flanked on one side, applying Lemma 34, shows the break has length $< cm^{1/2}$. Since (EC + F1 + F2) holds with probability $\geq 1 - 6/n$, the result follows. ◀

4 Recoverability Theorem

We will now move on to the main result: proving that the expected recoverability of seed-chain-extend with indels is $\geq 1 - O((\log n)^2 n^{-C\alpha})$ for large enough n under the conditions stated in Lemma 35. To show this bound, we will first prove that the recoverability of the chain can be lower bounded as the contribution from three pieces: the aligned fraction, points missed by clipping anchors in the chain, and points missed in breaks. Working in a ‘good’ space, $EC \cap F_1 \cap F_2$, which occurs with probability $\geq 1 - 6/n$, we have shown that with probability $\geq 1 - 6/n$, break lengths are $< cm^{1/2}$, the total number of clipping anchors is small so they cannot miss many points, and the aligned fraction is $\geq 1 - O(\frac{1}{\sqrt{m}})$. Combining these terms, we will conclude that the expected recoverability of an optimal chain is high.

► **Definition 36.** *Let $C = ((i_1, j_1), \dots, (i_u, j_u))$ be a chain. If (i_ℓ, j_ℓ) is a clipping anchor, we define the set of points it misses as:*

$$Cl(i_\ell, j_\ell) = \{(i, j) \in P_H : i_\ell \leq i \leq i_\ell + k - 1, (i, j) \notin \{(i_\ell + t, j_\ell + t) : t = 0, \dots, k - 1\}\}.$$

The set of all missed points from clipping anchors is: $Cl(C) = \bigcup_{(i,j) \in C \text{ clips } C} Cl(i, j)$.

► **Lemma 37** ((EC+F1+F2)). *Let $C = ((i_1, j_1), \dots, (i_u, j_u))$ be a chain. With probability $\geq 1 - 6/n$, the recoverability of C can be lower bounded as*

$$R(C) \geq \frac{\min f^{-1}(j_u) - \max f^{-1}(j_1) - L(C) - |Cl(C)| - O(|C|k)}{|P_H|}$$

Proof. We will lower bound the recoverability of P_H between $(\max f^{-1}(j_1), j_1)$ and $(\min f^{-1}(j_u), j_u)$, which is clearly a lower bound on $R(C)$. Consider any point $(x, y) \in P_H$ that lies between two consecutive homologous or clipping anchors $(i_l, j_l), (i_{l+1}, j_{l+1})$, i.e. $i_l \leq x \leq i_{l+1}$ and $j_l \leq y \leq j_{l+1}$ then $(x, y) \in \text{Ext}(l)$ for all $(x, y) \in P_H$ unless $x_l \in \{i_l, i_{l+1}\}$ or $y \in \{j_l, j_{l+1}\}$. In other words, (x, y) is recovered unless it lies on the edges of the extension 'box'. Note that any point on P_H on the edges of the extension box must be due to an insertion or deletion at the last position of the (i_l, j_l) anchor or directly before the start of the (i_{l+1}, j_{l+1}) anchor. Under EC, all insertions and deletions have $O(k)$ length, so there are at most $O(k)$ missing points on the path in $\text{Ext}(l)$. Applying the same argument shows that at most $O(|C|k)$ points are missed on the edges of extension boxes throughout the entire chain. Exactly $|Cl(C)|$ points on P_H that are covered, but not recovered by clipping chain anchors, are missed. Consider a point $(x, y) \in P_H$ in the break B flanked by $(i_l, j_l), (i_{l+1}, j_{l+1})$, that does not lie on the edge of an extension box and is not covered by a clipping anchor. The anchors $(i_l, j_l), (i_{l+1}, j_{l+1})$ contain points on the homologous path, call them $(x_l, y_l), (x_{l+1}, y_{l+1})$. We have that $x_l \leq i_l < x < i_{l+1} \leq x_{l+1}$ and similarly for y . Thus, (x, y) belongs to the region of the path between those two points, and is counted in $L(B)$.

Let (i_a, j_a) be the first anchor in C that is clipping or homologous, which must exist under our space. Under EC, $|\min f^{-1}(j_a) - i_a| \leq ck$. Call the first break in the chain B . Every point in the break B is counted in $L(C)$. Thus, considering the path between $(\max f^{-1}(j_1), j_1)$ and $(\min f^{-1}(j_u), j_u)$, our expression misses at most $O(k)$ points at the start. By the same logic, at most $O(k)$ missed points at the end of the path are undercounted. Combining terms gives the final inequality. \blacktriangleleft

► **Theorem 38.** *The expected recoverability of any chain is $\geq 1 - O((\log n)^2 n^{-C\alpha})$ for large enough n under the conditions of Lemma 35.*

Proof. Recall that the recoverability of a chain C is defined as $R(C) = \frac{|\text{Align}(C) \cap P_H|}{|P_H|}$. We will work under $\mathcal{F} = \text{EC} \cap \text{F1} \cap \text{F2}$, so all breaks have length $< cm^{1/2}$, by Lemma 35. Note $\Pr(\mathcal{F}) \geq 1 - \frac{2}{n} - \frac{3}{n} - \frac{1}{n} = 1 - \frac{6}{n}$. Writing $R = R(C)$ for shorthand, we have that

$$\mathbb{E}(R \mid \mathcal{F}) \geq \mathbb{E}\left(\frac{\min f^{-1}(j_u) - \max f^{-1}(j_1)}{|P_H|} \mid \mathcal{F}\right) - \mathbb{E}\left(\frac{L(C) + Cl(C)}{|P_H|} \mid \mathcal{F}\right) - c_0 \mathbb{E}\left(\frac{|C|}{|P_H|} \mid \mathcal{F}\right).$$

From Lemma 34, $\min f^{-1}(j_u) - \max f^{-1}(j_1) \geq |P_H| - c\sqrt{m}$, so the bound becomes

$$\mathbb{E}(R \mid \mathcal{F}) \geq 1 - \frac{4c_1}{\sqrt{m}} - \mathbb{E}\left(\frac{L(C) + Cl(C)}{|P_H|} \mid \mathcal{F}\right) - c_0 \mathbb{E}\left(\frac{|C|}{|P_H|} \mid \mathcal{F}\right).$$

We have $L(C) = \sum_{B \in \text{Breaks}} L(B) \leq C_S m^{1/2} \leq N_S m^{1/2}$, since each break contains some spurious anchor in the chain, and that total is at most the total number of spurious anchors. Under \mathcal{F} , $\mathbb{E}(L(C) \mid \mathcal{F}) \leq \mathbb{E}(N_S m^{1/2} \mid \mathcal{F}) \leq \frac{mn^{1-C}}{(1-6/n)}$, using that $\mathbb{E}(N_S) \leq mn^{1-C}$ and the law of total expectation.

Moving on to the second term, note that $Cl(C) \leq kN_C$, so $\mathbb{E}(Cl(C) \mid \mathcal{F}) \leq k\mathbb{E}(N_C \mid \mathcal{F}) \leq c_1 mk^2(1 - \theta_T)^k$ by Corollary 27. Under \mathcal{F} , $|P_H| \geq c_2 m$ for some constant $c_2 > 0$, so $\mathbb{E}\left(\frac{Cl(C)}{|P_H|} \mid \mathcal{F}\right) \leq ck^2(1 - \theta_T)^k = ck^2 n^{-C\alpha}$ for some positive constant c . We can bound the third term in the same way, since $|C| = C_H + C_S + C_c \leq N_H + N_S + N_C$ and $\mathbb{E}(N_H + N_S + N_C \mid \mathcal{F}) \leq c' mkn^{-C\alpha}$, because $\mathbb{E}(N_C)$ dominates, so $\mathbb{E}(c_0 k \frac{|C|}{|P_H|} \mid \mathcal{F}) \leq c' k^2 n^{-C\alpha}$.

Thus,

$$\mathbb{E}(R \mid \mathcal{F}) \geq 1 - \frac{4c_1}{\sqrt{m}} - \frac{mn^{1-C}}{(1-6/n)} - ck^2n^{-C\alpha}$$

Since $\Pr(\mathcal{F}) \geq 1 - 6/n$, we get

$$\begin{aligned} \mathbb{E}(R) &\geq \mathbb{E}(R \mid \mathcal{F}) \Pr(\mathcal{F}) \\ &\geq \left(1 - \frac{4c_1}{\sqrt{m}}\right) \left(1 - \frac{6}{n}\right) - mn^{1-C} - ck^2n^{-C\alpha} \left(1 - \frac{6}{n}\right) \\ &= 1 - O(m^{-1/2}) - O((\log n)^2 n^{-C\alpha}) \\ &= 1 - O((\log n)^2 n^{-C\alpha}). \end{aligned}$$

◀

5 Conclusion

In this work, we have shown that under the assumptions of Lemma 35 and $0 < \theta_T < 0.206$, the expected recoverability of an optimal chain under indels is $\geq 1 - O((\log n)^2 n^{-C\alpha})$. This result is weaker than the substitution-only case, in which Yu and Shaw [25] proved that the expected recoverability of an optimal chain is $\geq 1 - O(\frac{1}{\sqrt{m}})$. The weaker bound is due to the existence of clipping anchors, which uniquely arise with indels. In the substitution-only case, an anchor either lies entirely on or off the path. However, when the path is kinked, as it is with indels, this is no longer true. In the prequel [25], unrecovered points arise from breaks or at the start/end of the chain – no points are missed in regions covered by homologous anchors or on the ‘sides’ of extension boxes. We handled unrecovered points due to clipping anchors and the sides of extension boxes by naively upper bounding the size of the chain C by the total number of anchors $N_H + N_S + N_C$ and the number of clipping anchors in the chain by N_C . In future work, we aim to bridge the gap between recoverability results under indels and only substitutions by making better use of the match graph’s dependence structure, and further analyzing clipping anchor contributions to the chaining score.

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