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S1 $\Pr(\alpha(\theta, k))$ formula and figure

Theorem S1.1. For $2k - 1$ i.i.d Bernoulli trials with success probability $1 - \theta$ and $0 \leq \beta \leq k - 1$,

$$\begin{aligned} \Pr(\alpha(\theta, k) = \beta + 1) &= \Pr(\text{Longest run of successes is } k + \beta) \\ &= \sum_{b=0}^{k-\beta-2} T_k(\beta, b) (1 - \theta)^{k+\beta+b} \cdot \theta^{k-\beta-b-1} \end{aligned}$$

where

$$T_k(\beta, b) = 2 \binom{k-2-\beta}{b} + (k-\beta-2) \binom{k-3-\beta}{b}$$

and binomial coefficients with negative parameters are 0. For $\beta = k - 1$, the probability of $2k - 1$ successes is just $(1 - \theta)^{2k-1}$.

Proof. Suppose $\beta < k - 1$. If the maximum successful run is of length $k + \beta$ in $2k - 1$ trials, this must be the *only* run of $k + \beta$ successes in a row. Label the start and end of this sequence by positions $i, j \in \{1, \dots, 2k - 1\}$, where $j = i + k + \beta - 1$. The possible positions of i are $i \in \{1, \dots, k - \beta\}$. We calculate $\sum_{k=1}^{k-\beta} \Pr(k + \beta \text{ successes in a row}, i = k)$.

Case 1: if $i = 1$ or $i = k - \beta$, then trial $i + 1$ or $i - 1$ has to be a failure respectively, otherwise the run is longer than $k + \beta$. There are $2k - 1 - (k + \beta + 1) = k - \beta - 2$ remaining trials which can be either successes or failures. Letting b be the number of successes in the rest of the trials and conditioning on b , we get the probability of $i = 1$ or $i = k - 1$ as

$$2 \sum_{b=0}^{k-\beta-2} \binom{k-2-\beta}{b} (1 - \theta)^{k+\beta+b} \theta^{k-\beta-b-1}.$$

Case 2: if $i \neq 1$ and $i \neq k - \beta$, then both of the trials $i - 1$ and $j + 1$ have to be failures. This leaves us with $k - \beta - 3$ remaining trials. Conditioning on b again, we get the probability of $i = 2, \dots, k - 2$ as

$$(k - \beta - 2) \sum_{b=0}^{k-\beta-3} \binom{k-\beta-3}{b} (1 - \theta)^{k+\beta+b} \theta^{k-\beta-b-1}.$$

Summing the probabilities together yields the result when $\beta < k - 1$. If $\beta = k - 1$ then clearly the probability is just $(1 - \theta)^{2k-1}$. \square

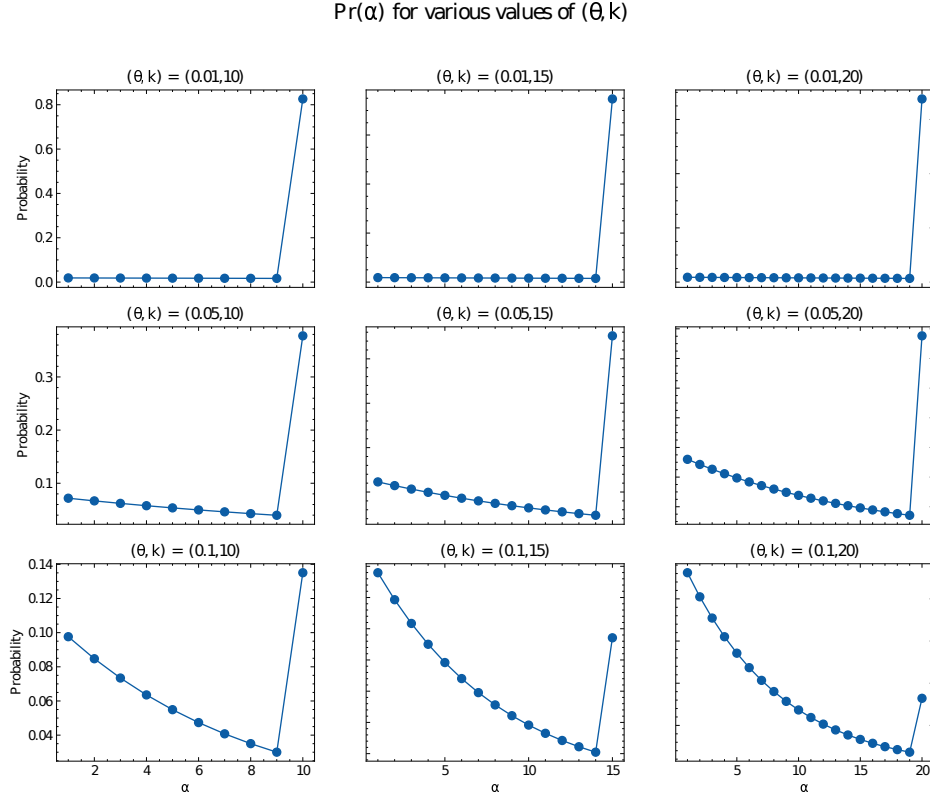


Figure 1: $\Pr(\alpha(\theta, k)) = [\Pr(\alpha(\theta, k) = 1), \dots, \Pr(\alpha(\theta, k) = k)]$ for various values of θ and k .

S2 Open syncmer proofs

Theorem S2.1 (Successful permutations for open syncmers). *Using parameters k, s, t as defined in the definition of open syncmers let $\tau = t - 1$ and $OS(\alpha, k, s, t)$ be the number of permutations in $S_{k-s+\alpha}$ such that for some window $[\sigma(i), \dots, \sigma(i + k - s)]$ the smallest element is $\sigma(i + \tau)$. Define $\ell_1 = \tau, \ell_2 = k - s - \tau$. Then*

$$OS(\alpha, k, s, t) = \alpha(k - s + \alpha - 1)! + R(\alpha, k, s, t, \ell_1) + R(\alpha, k, s, t, \ell_2).$$

We define $R(\alpha, k, s, t, \ell)$ as

$$R(\alpha, k, s, t, \ell) = \sum_{\beta=1}^{\ell} (k - s + \alpha - 1)_{\beta-1} OS(\alpha - \beta, k, s, t)$$

where the subscript indicates falling factorial, and $OS(\alpha - \beta, k, s, t) = 0$ if $\beta \geq \alpha$.

This is proved in the Appendix.

Proof. We condition on the position of the smallest element, i.e. the index β for which $\sigma(\beta) = 1$. Let the set $A_\tau = \{\tau + 1, \tau + 2, \dots, \tau + \alpha\}$

(**Case 1 - if $\beta \in A_\tau$**). In this case, the window

$$[\sigma(\beta - \tau), \dots, \sigma(\beta - \tau + (k - s))]$$

is valid and has the desired property that $\sigma((\beta + \tau) + (\tau)) = \sigma(\beta)$ is the smallest integer in the window, so these permutations all satisfy condition 2 above. There are $\alpha(k - s + \alpha - 1)!$ such permutations.

(Case 2 - if $\beta < \tau + 1$). In this case, β is left of position t . Notice that for all windows containing position β will never be successful since the first window contains β at position $< \tau + 1$, and the relative position of β in subsequent windows will be $< \tau + 1$ as well.

The remaining windows which may still satisfy condition 2 lie are sub-windows of $[\sigma(\beta + 1) \dots \sigma(k - s + \alpha)]$, which may be considered a permutation in $S_{k-s+\alpha-\beta}$ after relabelling elements to be in $\{1, \dots, k - s + \alpha - \beta\}$ to preserve the relative order.

This new permutation has to satisfy condition 2, and the number of such permutations is exactly $OS(\alpha - \beta, k, s, t)$. We have to multiply by an additional $(k - s + \alpha - 1)_{\beta-1}$ to count the possible values for the $\beta - 1$ entries to the left of β , each of which give the same permutation in $S_{w+\alpha-b}$ after relabelling. Summing over $b = 1, \dots, \tau = \ell_1$ gives the $R(\alpha, k, s, t, \ell_1)$ term.

(Case 3 - if $\beta > \tau + \alpha$). This case is identical to case 2 and the same argument works after flipping directions. This works by summing over the $\ell_2 = k - s - \tau$ possible positions $\beta \in \{k - s + \alpha, k - s + \alpha - 1, \dots, \tau + 1 + \alpha\}$ and using the same relabelling after cutting off a portion of the permutation. The number of permutations for $\beta = k - s + \alpha - i$ is the same as for $\beta = i$ by symmetry. Using this correspondence gives the $R(\alpha, k, s, t, \ell_2)$ term and completes the proof. \square

We now prove the following theorem.

Theorem S2.2. Let $\hat{t} = \lceil \frac{k-s+1}{2} \rceil$. Then $OS(\alpha, k, s, \hat{t}) \geq OS(\alpha, k, s, t)$ for any valid choice of t .

Lemma S2.3. Fix k, s, t, α and define $(k - s + \alpha - \beta - 1)_{\beta-1} OS(\alpha - \beta, k, s, t) = \overline{OS}(\alpha, \beta, t)$. If $\gamma \geq \beta$, for any t , we have

$$\overline{OS}(\alpha, \beta, t) \geq \overline{OS}(\alpha, \gamma, t).$$

Proof of Lemma. We show $\overline{OS}(\alpha, \beta - 1, t) \geq \overline{OS}(\alpha, \beta, t)$ for any β , which implies the result. This is equivalent to showing that

$$\begin{aligned} OS(\alpha - \beta + 1, k, s, t) &\geq \frac{(k - s + \alpha - 1)_{\beta-1}}{(k - s + \alpha - 1)_{\beta-2}} OS(\alpha - \beta, k, s, t) \\ &= (k - s + \alpha - \beta + 1) OS(\alpha - \beta, k, s, t). \end{aligned}$$

Notice that

$$OS(\alpha - \beta + 1, k, s, t) / (k - s + \alpha - \beta + 1)! = \Pr(f, \alpha - \beta + 1)$$

and

$$OS(\alpha - \beta, k, s, t) / (k - s + \alpha - \beta)! = \Pr(f, \alpha - \beta)$$

when f is an open syncmer method with fixed parameters k, s, t from our correspondence between random permutations and the event a k -mer is selected by f . By definition, $\Pr(f, \alpha - \beta + 1) \geq \Pr(f, \alpha - \beta)$. Technically, the correspondence is only true up to a small error due to the chance of repeated k -mers appearing in a window, but one can make $OS(x, k, s, t)$ arbitrarily close to $\Pr(f, x)$ by letting the alphabet be very large, making repeats unlikely (see the Section 2.3.1 in [1]). Then

$$OS(\alpha - \beta + 1, k, s, t) \geq (k - s + \alpha - \beta + 1) OS(\alpha - \beta, k, s, t)$$

follows from $\Pr(f, \alpha - \beta + 1) \geq \Pr(f, \alpha - \beta)$, and we're done. \square

Proof of Theorem S2.2. We use the similar notation as Lemma S2.3 for \overline{OS} .

Observe that

$$OS(\alpha, k, s, t) = OS(\alpha, k, s, k - s + 2 - t)$$

since this just swaps the ℓ_1, ℓ_2 in the definition. Since $k - s + 2 - \hat{t} = \hat{t}$ or $\hat{t} + 1$ depending on if $k - s + 1$ is odd or even, we only need to prove that this inequality holds for $t < \hat{t}$. We will assume $k - s + 1$ is odd for exposition since the indices are easier to handle, but the result holds either way after a slight modification.

We proceed by induction on α for $OS(\alpha, k, s, t)$. For the base case $\alpha = 1$, notice that clearly $OS(1, k, s, t) = OS(1, k, s, \hat{t}) = (k - s)!$. Let $\hat{\tau} = \hat{t} - 1$ and $\tau = t - 1$. Therefore we want the following term to be positive:

$$\begin{aligned} OS(\alpha, k, s, \hat{t}) - OS(\alpha, k, s, t) &= \sum_{\beta=1}^{\tau} [\overline{OS}(\alpha, \beta, \hat{t}) - \overline{OS}(\alpha, \beta, t)] \\ &\quad + \sum_{\beta=1}^{k-s-\hat{\tau}} [\overline{OS}(\alpha, \beta, \hat{t}) - \overline{OS}(\alpha, \beta, t)] \\ &\quad + \sum_{\beta=\tau+1}^{\hat{\tau}} \overline{OS}(\alpha, \beta, \hat{t}) - \sum_{\beta=k-s-\hat{\tau}+1}^{k-s-\tau} \overline{OS}(\alpha, \beta, t) \end{aligned} \tag{1}$$

By the induction assumption the first two sums are ≥ 0 since $OS(\alpha - \beta, k, s, \hat{t}) \geq OS(\alpha - \beta, k, s, t)$ for $\beta \geq 1$. For the last term, $k - s + 1$ odd gives us $k - s - \hat{\tau} = \hat{\tau}$. We can rewrite the last line as

$$\sum_{j=1}^{\hat{\tau}-\tau} \overline{OS}(\alpha, (\hat{\tau} - j + 1), \hat{t}) - \overline{OS}(\alpha, (\hat{\tau} + j), t).$$

By the induction assumption, $\overline{OS}(\alpha, (\hat{\tau} - j + 1), \hat{t}) \geq \overline{OS}(\alpha, (\hat{\tau} - j + 1), t)$ and using Lemma S2.3 finishes the proof because $\hat{\tau} - j + 1 \leq \hat{\tau} + j$ for all $j \geq 1$. □

We show in Figure 2 what $\Pr(f)$ looks like for syncmer methods over a range of t .

S3 Proof of random minimizer probability vector

Theorem S3.1 (Successful permutations for random minimizers). *Given parameters (n, w, α, p) with $p + \alpha - 1 \leq n$, let $M(n, w, \alpha, p)$ be the number of permutations in S_n such that for some window $[\sigma(i), \dots, \sigma(i + w - 1)]$, the smallest element is one of $\sigma(p), \sigma(p + 1), \dots, \sigma(p + \alpha - 1)$. Then*

$$M(n, w, \alpha, p) = \begin{cases} (a)(n - 1)! + \tilde{R}(n, w, \alpha, \tilde{\ell}_1) + \tilde{R}(n, w, a, \tilde{\ell}_2) & \text{for } w \leq n \\ 0 & \text{for } w > n \end{cases}$$

where $\tilde{\ell}_1 = p - 1$, $\tilde{\ell}_2 = n - (p + \alpha - 1)$ and using $(x)_n$ to mean the falling factorial,

$$\tilde{R}(n, w, \alpha, \ell) = \sum_{\beta=1}^{\tilde{\ell}} M(n - \beta, w, \alpha, \tilde{\ell} - \beta + 1) \cdot (n - 1)_{\beta-1}.$$

Proof. As in the proof of Theorem S2.1, we condition on the position of the smallest element, i.e. the index β for which $\sigma(\beta) = 1$. Let the set $A_p = \{p, p + 1, \dots, p + \alpha - 1\}$.

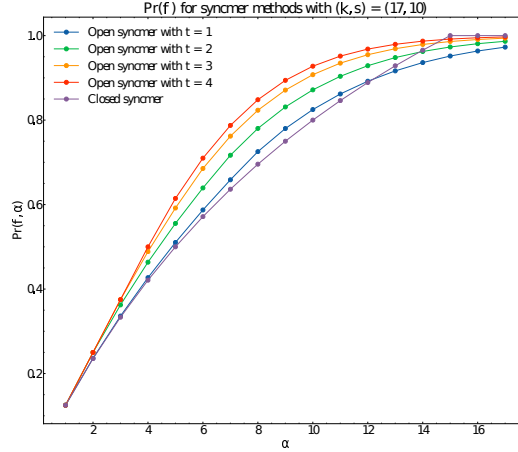


Figure 2: The probability vector for open syncmers with varying t parameters $(k, s) = (17, 10)$. The s parameter for the closed syncmer was chosen so that the densities are equal. We only evaluate for $t \leq 4$ because $t \mapsto k - s + 2 - t$ gives the same probabilities for open syncmers by Theorem S2.1.

(Case 1 - if $\beta \in A_p$). This permutation clearly is successful. There are $\alpha(n-1)!$ such permutations.

(Case 2 - if $\beta < p$). In this case, β is left of position p . All windows containing position β will never be successful since $\beta \notin A_p$ and $\sigma(\beta)$ is the smallest element in the window. The only possible successful windows are the sub-windows of $[\sigma(\beta+1) \dots \sigma(n)]$. We can relabel the positions after shifting by β and consider this as a new permutation on $1, \dots, n-\beta$ after relabelling $\sigma(i)$ while preserving relative order. This sub-problem is exactly counted by $(n-1)_{\beta-1}M(n-\beta, w, \alpha, p-\beta)$ after multiplying by the $(n-1)_{\beta-1}$ possible values for $\sigma(i)$, $i < \beta$. Notice that if $n-\beta < w$, then there are no windows that satisfy our requirement, so $M(n-\beta, w, \alpha, p-\beta) = 0$. Summing over $\beta < p$ gives

$$\sum_{\beta=1}^{p-1} (n-1)_{\beta-1} M(n-\beta, w, \alpha, p-\beta).$$

(Case 3 - if $\beta > p + \alpha - 1$). The exactly same argument follows as in case 2. We see that successful windows must be sub-windows of $[\sigma(1), \dots, \sigma(\beta-1)]$, so this is almost counted by $(n-1)_{n-\beta}M(\beta-1, w, \alpha, p)$ over all $\beta > p + \alpha - 1$. We can shift indices to get

$$\sum_{\beta=p+\alpha}^n (n-1)_{n-\beta} M(\beta-1, w, \alpha, p) = \sum_{\beta=1}^{n-(p+\alpha-1)} M(n-\beta, w, \alpha, p) (n-1)_{\beta-1}.$$

To rewrite the equation to be in a similar form to case 2, one can see that $M(n-\beta, w, \alpha, p) = M(n-\beta, w, \alpha, (n-\beta) - (p+\alpha-1) + 1)$ which corresponds to “flipping” the permutation on $S_{n-\beta}$ so that position $i \mapsto n-\beta-i+1$. This completes the proof. \square

S4 Proof of (a, b, m) -words method probability vector

Theorem S4.1. $\text{Pr}(f, \alpha-1)$ under the (a, b, n) -words method is

$$\sum_{i=1}^{\alpha} (-1)^{i+1} \frac{3^{ni}}{4^{i(n+1)}} \binom{\alpha - n(i-1)}{i}$$

where $\binom{x}{y} = 0$ if $x < 0$.

We first prove an intermediate combinatorial lemma.

Lemma S4.2. *Given a set of α elements labelled $\{1, \dots, \alpha\}$, the number of ways $c(n+1, i)$ to choose i elements x_1, \dots, x_i where we order $x_j < x_{j+1}$ for $j = 1, \dots, i-1$ and $|x_j - x_{j+1}| \geq (n+1)$ for all j is*

$$\binom{\alpha - n(i-1)}{i}.$$

Proof. Let $y_0 = x_1 - 1$, $y_1 = x_2 - x_1 - 1$, ..., $y_i = \alpha - x_i$. The y_i s represent the gaps between x_i s and also the endpoints. A valid choice of x_i s corresponds exactly to a choice of y_i s such that each $y_i \geq n$ for $i = 1, \dots, i-1$ and $y_0, y_i \geq 0$. Furthermore,

$$\sum_{j=0}^i y_j = \sum_{j=1}^{i-1} (x_{j+1} - x_j - 1) + \alpha - x_i + x_1 - 1 = \alpha - i$$

We can take $z_j = y_j - n$ for $j = 1, \dots, i-1$ and $z_j = y_j$ otherwise to get the equivalent problem of finding z_j all ≥ 0 such that

$$\sum_{j=0}^i z_j = \alpha - i - (i-1)n.$$

This problem is equivalent to putting $\alpha - i - (i-1)n$ indistinct balls into $i+1$ distinct jars represented by the variables z_j . The solution is

$$\binom{[i+1] + [\alpha - i - (i-1)n] - 1}{\alpha - i - (i-1)n} = \binom{\alpha - n(i-1)}{i}$$

as desired. □

Proof of Theorem S4.1. The probability that at least one of the k -mers is selected is

$$\Pr\left(\bigcup_{i=1}^{\alpha} E_i\right)$$

where E_i is the event that the i -th k -mer is selected. By inclusion-exclusion, we get

$$\Pr\left(\bigcup_{i=1}^{\alpha} E_i\right) = \sum_{I \subset \{1, \dots, \alpha\}} (-1)^{|I|+1} \Pr(E_I) = \sum_{i=1}^{\alpha} (-1)^{i+1} \sum_{I \subset \{1, \dots, \alpha\}, |I|=i} \Pr(E_I)$$

where $E_I = \bigcup_{i \in I} E_i$. Now note that the probability that $E_{\alpha} \cap E_{\beta}$ for $|\alpha - \beta| < n$ occurs is 0; k -mers with prefix $abbb\dots$ may not be within distance $n+1$ from each other. If the i k -mers are all distance $\geq n+1$ apart, then the probability of that event occurring is just $(\frac{3^n}{4 \cdot 4^n})^i$ because this is just the sequence $abbb\dots$ appearing i times in a string of i.i.d random letters. Therefore, denoting $c(i, n+1)$ to be the number of ways to select i elements from $\{1, \dots, \alpha\}$ such that each element is at least pairwise distance $n+1$ apart, we get

$$\sum_{i=1}^{\alpha} (-1)^{i+1} \sum_{I \subset \{1, \dots, \alpha\}, |I|=i} \Pr(E_I) = \sum_{i=1}^{\alpha} (-1)^{i+1} c(n+1, i) \frac{3^{ni}}{(4^{n+1})^i}.$$

Plugging in the above lemma finishes the proof. □

S5 Comparing $\Pr(f)$

In Figure 3, we plot all $\Pr(f)$ and $UB(d)$ where all methods have density $d = 1/7$ except for the words method, which has density $9/64 \sim 1/7.11$. This is due to the limited range of parameters choices for the methods.

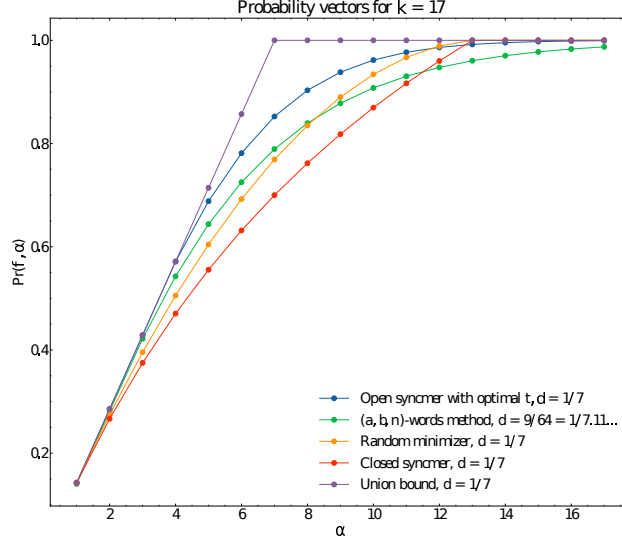


Figure 3: Comparison of $\Pr(f)$ for all methods with exact distributions derived. Note that the density for the words method is slightly smaller.

S6 Defining W_4, W_8

We take the words set W_4 as

$$W_4 = \{rrrrry, rryrry, rryryy, ryrrrr, \\ ryrrry, ryryry, ryyrrr, ryrrry, \\ ryyryr, ryyryy, ryyryy, ryyryy, \\ ryrrry, ryryry, ryyrrr, ryrrry\}. \quad (2)$$

Here $r = \{A, G\}$ and $y = \{C, T\}$ and we mean $rrryrry$ to be all 6-mers that satisfy this condition. This set leads to a $d = 1/4$ method, and was found by an optimization algorithm [2].

We take the words set W_8 as

$$W_8 = \{rrrrrrry, rryrrrry, rrrrrryr, rrrrrryy, \\ rrrrrryy, yrrrrrry, yrrrrryr, yrrrrryy, \\ yrrrrryy, yrrrrryr, yrrrrryy, yrrrrryy, \\ yrrrrryy, yrrrrryr, yrrrrryy, yrrrrryy, \\ yrrrrryy, yrrrrryy, yrrrrryy, yrrrrryy, \\ yrrrrryy, yrrrrryy, yrrrrryy, yrrrrryy\} \quad (3)$$

Here $r = \{A, G\}$ and $y = \{C, T\}$ and we mean $rryrry$ to be all 6-mers that satisfy this condition. This set leads to a $d = 1/8$ method, and was found by an optimization algorithm [2].

References

- [1] H. Zheng, C. Kingsford, and G. Marçais, “Improved design and analysis of practical minimizers,” *Bioinformatics*, vol. 36, pp. i119–i127, July 2020.
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