Tangled Paths: A Random Graph Model from Mallows Permutations

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Abstract

We introduce the random graph $\mathcal{P}(n,q)$ which results from taking the union of two paths of length $n \geq 1$, where the vertices of one of the paths have been relabelled according to a Mallows permutation with parameter $0 < q(n) \leq 1$. This random graph model, the tangled path, goes through an evolution: if q is close to 0 the graph bears resemblance to a path, and as q tends to 1 it becomes an expander. In an effort to understand the evolution of $\mathcal{P}(n,q)$ we determine the treewidth and cutwidth of $\mathcal{P}(n,q)$ up to log factors for all q. We also show that the property of having a separator of size one has a sharp threshold. In addition, we prove bounds on the diameter, and vertex isoperimetric number for specific values of q.

Mathematics Subject Classifications: 05C80, 05A05, 68Q87, 05C78

1 Introduction

Given two graphs G, H on a common vertex set $[n] = \{1, \ldots, n\}$, and a permutation σ on [n], it is natural to consider the following graph

$$\mathsf{layer}(G, \sigma(H)) = ([n], E(G) \cup \{\sigma(x)\sigma(y) : xy \in E(H)\}),$$

which is the union of two graphs where the second graph has been relabelled by a permutation σ . Constructions of graphs via unions are very natural, and have appeared in several contexts, see Section 1.2. Let P_n be the path on [n] that connects i to i + 1, for $i \in [n-1]$, and let S_n be the set of all permutations on [n]. Consider the following scenario: one must choose a permutation $\sigma \in S_n$ with the goal of making layer $(P_n, \sigma(P_n))$ as different from a path as possible. There are many parameters one may use to measure the difference between a connected graph G and a path; for example one may look at the diameter diam(G) or the vertex isoperimetric number $\varphi(G)$, as the path is extremal for these parameters. The treewidth tw(G) which, broadly speaking, measures how far

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(globally) the graph is from being a tree [33], is another natural candidate. Given two or more paths one can build a grid-like graph (see [18, Lemma 8] for more details) which would have treewidth and diameter $\Theta(\sqrt{n})$. If we choose a permutation uniformly at random, then as a consequence of a result of Kim & Wormald [31, Theorem 1], with high probability the resulting graph is a bounded degree expander. Thus, in this case, the graph layer $(P_n, \sigma(P_n))$ has treewidth $\Theta(n)$ and diameter $\Theta(\log n)$, so by these parameters it is essentially as far from a path as a sparse graph can be.

The example above shows that even restricting the input graphs to paths can produce rich classes of graphs. Having seen what happens for a uniformly random permutation, one may ask about the structure of $layer(P_n, \sigma(P_n))$ when σ is drawn from a distribution on S_n that is not uniform. One of the most well known non-uniform distributions on S_n is the Mallows distribution, introduced by Mallows [37] in the late 1950s in the context of statistical ranking theory. Recently it has been the subject of renewed interest for other applications, and as an interesting and natural model to study in its own right, see Section 1.2. The distribution has a parameter q which, roughly speaking, controls the amount of disorder in the permutation.

For real q > 0 and integer $n \ge 1$, the (n, q)-Mallows measure $\mu_{n,q}$ on S_n is given by

$$\mu_{n,q}(\sigma) = \frac{q^{\text{Inv}(\sigma)}}{Z_{n,q}} \quad \text{for any } \sigma \in S_n,$$
 (1)

where $\text{Inv}(\sigma) = |\{(i, j) : i < j \text{ and } \sigma(i) > \sigma(j)\}|$ is the number of inversions in the permutation σ and $Z_{n,q}$ is given explicitly by the following formula [5, Equation (2)]:

$$Z_{n,q} = \prod_{i=1}^{n} (1 + q + \dots + q^{i-1}) = \prod_{i=1}^{n} \frac{1 - q^{i}}{1 - q}.$$

When $q \to 0$, the distribution $\mu_{n,q}$ converges weakly to the degenerate distribution on the identity permutation. We extend $\mu_{n,q}$ to q=0 by setting $\mu_{n,0}$ to be the probability measure assigning 1 to the identity permutation. On the other hand if q=1 then $\mu_{n,1}$ is the uniform measure on S_n . One can see that $\sigma \sim \mu_{n,q}$ has distribution $\mu_{n,1/q}$ when reversed.

We study the random graph given by layer $(P_n, \sigma(P_n))$, where $\sigma \sim \mu_{n,q}$ and $0 \leq q := q(n) \leq 1$. From now on we call this random graph the tangled path model and denote it by $\mathcal{P}(n,q)$. Thus a random graph $\mathcal{P}(n,q)$ has vertex set [n] and (random) edge set $E(P_n) \cup \{\sigma(i)\sigma(i+1) : i \in [n-1]\}$, where $\sigma \sim \mu_{n,q}$. We restrict to $q \in [0,1]$ as reversing the permutation does not affect our construction (up to a relabelling, see (8)). We also identify any multi-edges created as one edge, however this detail is not important for any of our results. This paper will focus on $\mathcal{P}(n,q)$; as we have seen already combining paths can give rise to interesting and varied graphs, and the Mallows permutation gives our model a parameter q which, roughly speaking, increases the 'tangled-ness' of the graph. Other reasons for using Mallows permutations are that they are well studied (see Section 1.2), and they are mathematically tractable since they can be generated by a sequence of independent random variables (see Section 3.1).

By the above, $\mathcal{P}(n,0)$ is a path and $\mathcal{P}(n,1)$ is an expander with high probability; the latter follows from [31, Theorem 1] but we also give a self-contained proof in this paper. Our ultimate aim is to understand the structure of $\mathcal{P}(n,q)$ for intermediate values of q,

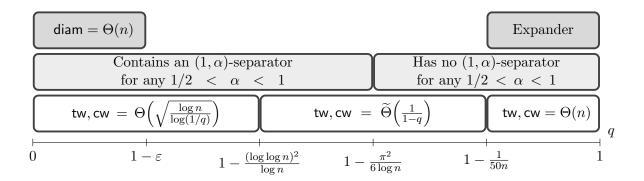


Figure 1: The diagram above gives a representation of our main results. All results above hold with high probability, and we say that $f(n) = \widetilde{\Theta}(g(n))$ if there exist constants c, C > 0 and n_0 such that $c \cdot g(n) / \log g(n) \leqslant f_n \leqslant C \cdot g(n) \cdot \log g(n)$ for all $n \geqslant n_0$.

and this paper takes the first steps in this direction. Informally, if q is not tending to 1 too fast, then $\mathcal{P}(n,q)$ is 'path-like'; we show that if q<1 is fixed the diameter is linear (Theorem 3), and there is a sharp threshold for having a single cut vertex at $q_c=1-\frac{\pi^2}{6\log n}$ (Theorem 2). For $q\to 1$ sufficiently fast, it makes more sense to measure the complexity of the internal structure of $\mathcal{P}(n,q)$ by how much it differs from a tree. Here we show that, up to logarithmic factors, the treewidth [33] of $\mathcal{P}(n,q)$ grows at rate $(1-q)^{-1}$ (Theorem 4) until the graph becomes an expander at around $q=1-\frac{1}{n}$ (Theorem 1), indicating that, in the sense of treewidth, the complexity of the structure grows smoothly with q. This behaviour contrasts with the binomial/Erdős-Rényi random graph [21] where treewidth increases rapidly from being bounded by a constant, to $\Theta(n)$ as the average degree rises from below one to above one [17, 34].

Further motivation for this line of study comes from practical algorithmic applications. Many real-world systems – including social, biological and transport networks – involve qualitatively different types of edges, where each type of edge generates a "layer" with specific structural properties [32, 43]. For example, when modelling the spread of disease in livestock, one layer of interest arises from physical adjacency of farms, and so is determined entirely by geography. A second epidemiologically-relevant layer could describe the pairs of farms which share equipment: this is no longer fully determined by geography, but will nevertheless be influenced by the location of farms, as those that are geographically close are more likely to cooperate in this way. It is known that algorithmically useful structure in individual layers of a graph is typically lost when the layers are combined adversarially [18]. The present work can be seen as an attempt to understand the structure of graphs generated from two simple layers which are both influenced to some extent by a shared underlying "geography". In this setting the treewidth tw is a natural parameter as many NP-hard problems become tractable when parametrised by tw [15, Ch. 7].

1.1 Our Results

In what follows, the integer $n \ge 1$ denotes the number of vertices in the graph (or elements in a permutation) and q := q(n), the parameter of the Mallows permutation (or related tangled path), is a real-valued function of n taking values in [0,1]. We say a sequence of

events \mathcal{E}_n occurs with high probability (w.h.p.) if $\mathbb{P}(\mathcal{E}_n) \to 1$ as $n \to \infty$. Throughout log is base e. See Figure 1 for a summary of our results.

A graph G is a vertex-expander if there exists c > 0 (independent of n) such that any set $S \subseteq V$ with $|S| \leqslant \lceil n/2 \rceil$ is adjacent to at least c|S| vertices in $V \setminus S$, see Section 2.1. As mentioned above, when q = 1 the permutation is uniform, and so the fact that w.h.p. $\mathcal{P}(n,1)$ is an expander follows from [31, Theorem 1]. We give a self-contained proof of this fact, which also shows that for q sufficiently close to 1, this still holds.

Theorem 1. If $q \ge 1 - \frac{1}{50n}$, then w.h.p. $\mathcal{P}(n,q)$ is a vertex-expander.

For an integer $s \ge 1$ and $1/2 \le \alpha < 1$ we say that a graph G with vertex set V has an (s, α) -separator if there is a vertex subset S with $|S| \le s$ such that $V \setminus S$ can be partitioned into two disjoint sets of at most $\alpha |V|$ vertices with no crossing edges, see Section 2. Balanced separators (e.g. $\alpha = 2/3$) are useful for designing divide and conquer algorithms, in particular for problems on planar graphs [36], and are connected to other notions of sparsity [41].

Observe that, for any fixed $1/2 < \alpha < 1$, if G is a vertex expander then there exists a c > 0 such that G has no (cn, α) -separator. At the other extreme, the path has a $(1, \alpha)$ -separator. We show that for $\mathcal{P}(n, q)$ this 'path-like' property disappears around $q_c = 1 - \frac{\pi^2}{6 \log n}$.

Theorem 2. For any fixed $1/2 < \alpha < 1$ we have

$$\lim_{n\to\infty} \mathbb{P}(\,\mathcal{P}(n,q)\,\,has\,\,a\,\,(1,\alpha)\text{-separator}) = \begin{cases} 0 & \text{if}\,\,\frac{\pi^2}{6(1-q)} - \log n + \frac{5\log\log n}{2} \to \infty \\ 1 & \text{if}\,\,\frac{\pi^2}{6(1-q)} - \log n + \frac{9\log\log n}{2} \to -\infty \end{cases}.$$

In other words, Theorem 2 shows that if q is sufficiently below the threshold $q_c = 1 - \frac{\pi^2}{6 \log n}$ then w.h.p. there is a cut vertex which separates the graph into two $\Theta(n)$ vertex subpaths, and if q is sufficiently above it then there does not. We say that q_0 is sharp threshold for a graph property \mathfrak{P} if for any $\varepsilon > 0$ w.h.p. $\mathcal{P}(n,p) \notin \mathfrak{P}$ for any $p \leqslant q_0(1-\varepsilon)$, and $\mathcal{P}(n,r) \in \mathfrak{P}$ for any $r \geqslant q_0(1+\varepsilon)$, see [20]. Theorem 2 is quite precise as it determines the second order of the threshold up to a constant, showing that the property of having a $(1,\alpha)$ -separator has a sharp threshold of width $\mathcal{O}\left(\frac{\log\log n}{(\log n)^2}\right)$. Theorem 2 is established by finding first and second moment thresholds for the property. Positive correlation between cuts suggests this result cannot be significantly improved using standard methods alone (see Remark 26).

The diameter $\operatorname{diam}(G)$ of a graph G is the length of the longest shortest path between any pair of vertices. Theorem 1 implies that $\operatorname{diam}(\mathcal{P}(n,q)) = \mathcal{O}(\log n)$ when q is sufficiently close to 1. On the other hand $\operatorname{diam}(\mathcal{P}(n,0)) = n-1$ as $\mathcal{P}(n,0)$ is a path; we show this holds (up to a constant) for any fixed q < 1.

Theorem 3. For any $0 < \varepsilon < 1$, let $0 < q \leqslant 1 - \varepsilon$. Then, there exists a constant $c := c(\varepsilon) > 0$ such that for $n \geqslant 1/c$, we have $\mathbb{P}(\operatorname{diam}(\mathcal{P}(n,q)) \geqslant cn) \leqslant n^{-1/10}$.

This result follows from bounds on the number of cut vertices used to prove Theorem 2. The treewidth $\mathsf{tw}(G)$ of a graph G is one less than the minimum size of the largest vertex subset (i.e. bag) in a tree decomposition of G, minimised over all such decompositions. The cutwidth $\mathsf{cw}(G)$ is the greatest number of edges crossing any real point under

an injective function $f: V \to \mathbb{Z}$, minimised over all f. See Section 2.1 for full definitions of these quantities. It is known that for any graph G we have $\mathsf{tw}(G) \leqslant \mathsf{cw}(G)$, however there may be a multiplicative discrepancy of order up to n. We show there is at most only a constant factor discrepancy for $G = \mathcal{P}(n,q)$ in certain ranges of q, and give bounds for all q which are tight up to a log factor.

Theorem 4. For any constant $\kappa > 0$, let $0 < q \leqslant 1 - \kappa \cdot \frac{(\log \log n)^2}{\log n}$. Then, there exist constants $c_1, c_2 > 0$ such that w.h.p.

$$c_1 \cdot \left(\sqrt{\frac{\log n}{\log(1/q)}} + 1 \right) \leqslant \mathsf{tw}(\mathcal{P}(n,q)) \leqslant \mathsf{cw}(\mathcal{P}(n,q)) \leqslant c_2 \cdot \left(\sqrt{\frac{\log n}{\log(1/q)}} + 1 \right).$$

Furthermore, if $1 - \frac{(\log \log n)^2}{\log n} \leqslant q \leqslant 1$, then w.h.p.

$$10^{-5} \cdot \min \left\{ \frac{1}{1-q}, \ n \right\} \leqslant \mathsf{tw}(\mathcal{P}(n,q)) \leqslant \mathsf{cw}(\mathcal{P}(n,q)) \leqslant 5 \cdot \max \left\{ \frac{1}{1-q} \cdot \log \left(\frac{1}{1-q} \right), \ n \right\}.$$

Observe that if $q \to 1$ then $\log(1/q) \approx 1-q$ and so when $q = 1-\Theta\left((\log\log n)^2/\log n\right)$ we have $\sqrt{\log(n)/\log(1/q)} \approx -\log(1-q)/(1-q)$. Thus, the two upper bounds on the cutwidth are equal up to constants for this range of q. Hence, for this range of q, the upper bound for the cutwidth given in the second equation is tight and the lower bound for treewidth is off by a multiplicative factor of order $\log\log n$.

The lower bounds on treewidth in Theorem 4 are proved by relating the treewidth to the occurrence of certain permutations as consecutive patterns in the underlying Mallows permutation. The upper bounds on cutwidth are proved by controlling the density of long edges.

1.2 Further Related Work

Many works have studied properties of a typical permutation sampled from the Mallows measure, in particular the longest increasing subsequence [5, 40], cycle structure [23], permutation pattern avoidance [14, 44] and sets of consecutive elements [45]. Mallows permutations also arise as limit objects from stable matchings [4] and have been studied in the contexts of statistical physics [47], Markov chains [16], learning theory [11] and finitely dependent processes [28].

Random graphs have been heavily studied since their introduction in the late 1950s [21]. A random graph arising from the Mallows distribution is introduced in [6]. In this model each edge corresponds to an inversion in the permutation, so it is different to our model. To our knowledge the model in [6] and the tangled path introduced here are the only random graph models based on Mallows permutation. However, some works have studied relations between random graph models and uniform permutations. In particular, in [31] it is proven that the union of two uniformly permuted cycles is contiguous to a random 4-regular graph. Very recently [12] used a union of two paths permuted by a uniform random permutation to get a lower bound on resolution refutations for SAT solvers. Also, [22] shows the union of two uniformly random trees on the same vertex set is an expander with high probability.

There are also several papers which consider the graphs formed from (permuted) unions of graphs. In particular independent sets in the union of two Hamiltonian cycles [2], and the treewidth of a union of two graphs glued using a permutation [3] and [46, Chapter 5]. The clique number of graph unions [1, 42], and unions of cliques have been studied [9]. Unions of dense graphs with random graphs, namely 'randomly perturbed graphs', have been studied intensely, see [8] and citing papers. There is also a connection between graph unions and threshold graphs [27]. From the other direction, decompositions of graphs have been well studied [24, 39, 48].

1.3 Outline of the Paper

In Section 2 we cover some basic notation, definitions and concentration inequalities. In Section 3 we state some known facts about the Mallows distribution, in particular defining the q-Mallows process, before introducing our notions of 'flushing' and 'local' events that are useful later in the paper. Section 4 establishes properties of the tangled path in the case where q is close to one. The first result in this section shows that when q=1 (i.e. a uniformly random permutation) the tangled path is an expander with high probability. Then, a bound on the probability of events under the q-Mallows measure by that of the 1-Mallows measure is shown; this is useful later when bounding the treewidth. The beginning of Section 5 focuses on bounding the probability of flushing events. In the remainder of Section 5, these bounds are used to prove a sharp threshold for $(1,\alpha)$ -separators and a linear bound on the diameter. Arguably the most interesting techniques and proofs are in Section 6. To prove a lower bound on the treewidth we use consecutive patterns in the Mallows permutations to find smaller tangled paths with a higher q parameter as minors in the tangled path. To prove corresponding upper bounds we control the cutwidth by bounding the number of 'long' edges created during the q-Mallows process. Finally, we conclude with some open problems in Section 7.

2 Notation and Preliminaries

For a random variable X and probability measure μ we use $X \sim \mu$ to say that X has distribution μ . Let X, Y be random variables, then X stochastically dominates Y if $\mathbb{P}(X \geqslant x) \geqslant \mathbb{P}(Y \geqslant x)$ for all real x, and we denote this by $X \succeq Y$. We let Ω denote the sample space and $\mathcal{E}^c = \Omega \setminus \mathcal{E}$ to denote the complement of an event \mathcal{E} . We also let $\mathbf{1}_{\mathcal{E}}:\Omega \to \{0,1\}$ denote the indicator random variable where $\mathbf{1}_{\mathcal{E}}(\omega) = 1$ if $\omega \in \mathcal{E}$ and $\mathbf{1}_{\mathcal{E}}(\omega) = 0$ otherwise.

Throughout log denotes the natural logarithm (base e) we will also use the ln notation for this natural logarithm sometimes for reader recognition. We note that for any real x > -1,

$$\frac{x}{1+x} \leqslant \log(1+x) \leqslant x. \tag{2}$$

We use standard asymptotic (big- \mathcal{O} etc.) notation consistent with [21]. A sequence of events (\mathcal{E}_n) holds with high probability (w.h.p.) if $\lim_{n\to\infty} \mathbb{P}(\mathcal{E}_n) = 1$. We use := to indicate suppressed dependency, e.g. C := C(c) if the constant C depends on c.

2.1 Expansion, Width Measures and Graph Minors

Let $S \subseteq V$ and define the edge boundary $\partial(S) = \{uv \in E(G) : u \in S, v \in V(G) \setminus S\}$ of S to be the set of edges with one endpoint in S and the other outside S. Similarly we let the outer vertex boundary $\overline{N}(S) = \{v \in V \setminus S : \text{ there exists } u \in S, uv \in E\}$ of S to be the set of vertices outside S which share an edge with a vertex in S. We then define

$$\phi(G) = \min_{0 < |S| \leqslant n/2} \frac{|\partial(S)|}{|S|} \qquad \text{and} \qquad \varphi(G) = \min_{0 < |S| \leqslant n/2} \frac{|\overline{N}(S)|}{|S|}$$

to be the edge and vertex isoperimetric numbers respectively. For any graph G we have

$$\varphi(G) \leqslant \phi(G) \leqslant \max_{v \in V} |\overline{N}(v)| \cdot \varphi(G).$$
(3)

We say that a graph sequence G_n is an edge (resp. vertex) expander sequence if there exists some fixed $\alpha > 0$ such that $\phi(G_n) \ge \alpha$ (resp. $\varphi(G_n) \ge \alpha$) for all n suitably large, see [29].

Let $\frac{1}{2} \leq \alpha < 1$, $s \geq 0$ an integer, and G = (V, E) a graph. A subset $S \subset V$ is said to be an (s, α) -separator of G [10, 33], if there exists subsets $A, B \subset V$ such that

- $V = A \cup B \cup S$ and A, B, S are pairwise disjoint,
- $|S| \leq s$, |A|, $|B| \leq \alpha |V|$, and
- $\{ab \in E : a \in A, b \in B\} = \emptyset$.

A tree decomposition of a graph G = (V, E) is a pair $(\{X_i : i \in I\}, T = (I, F))$ where $\{X_i : i \in I\}$ is a family of subsets (or 'bags') $X_i \subseteq V$ and T = (I, F) is a tree such that

- $\bullet \bigcup_{i \in I} X_i = V,$
- for every edge $vw \in E$ there exists $i \in I$ with $\{v, w\} \subseteq X_i$,
- for every $i, j, k \in I$, if j lies on the path from i to k in T, then $X_i \cap X_k \subseteq X_j$.

The width of $(\{X_i : i \in I\}, T = (I, F))$ is defined as $\max_{i \in I} |X_i| - 1$. The treewidth $\mathsf{tw}(G)$ of G is the minimum width of any tree decomposition of G. Thus, for any graph G, $\mathsf{tw}(G) \leq n - 1$.

For an injective function $f:V(G)\to\mathbb{Z}$ we define the *cutwidth* [13] of G by

$$\operatorname{cw}(G) = \min_{f:V \to \mathbb{Z}, \text{ injective }} \max_{x \in \mathbb{R}} |\{ij \in E(G) : f(i) \leqslant x < f(j)\}|; \tag{4}$$

this is the maximum number of edges crossing a real point when the vertices are arranged in a line according to f, minimised over all injections $f: V \to \mathbb{Z}$. By [26, Proposition 1] and [7],

$$\lfloor \varphi(G) \cdot n/4 \rfloor \leqslant \mathsf{tw}(G) \leqslant \mathsf{cw}(G) \leqslant |E(G)|. \tag{5}$$

A graph H is called a *minor* of the graph G if H can be formed from G by deleting edges and vertices and by contracting edges.

Lemma 5 (Folklore, see [33]). If H is a minor of G then $tw(H) \leq tw(G)$.

2.2 Concentration Inequalities

Let Geo(p) denote the geometric distribution with success probability p. That is, if a random variable $X \sim Geo(p)$ then $\mathbb{P}(X = k) = (1 - p)^{k-1}p$ for any integer $k \ge 1$.

Lemma 6 ([30, Theorem 2.3]). For any $n \ge 1$ and $p_1, ..., p_n \in (0, 1]$, let $X = \sum_{i=1}^n X_i$ where $X_i \sim \text{Geo}(p_i)$. Let $p_* = \min_{i \in [n]} p_i$ and $\mu = \mathbb{E}[X] = \sum_{i=1}^n \frac{1}{p_i}$. Then for any $\lambda \ge 1$,

$$\mathbb{P}(X \geqslant \lambda \mu) \leqslant \lambda^{-1} (1 - p_*)^{\mu(\lambda - 1 - \log \lambda)}.$$

Let $\operatorname{Ber}(p)$ denote the Bernoulli distribution with success probability $0 \leq p \leq 1$. If $X \sim \operatorname{Ber}(p)$ then $\mathbb{P}(X=1) = p$ and $\mathbb{P}(X=0) = 1 - p$.

Lemma 7 ([38, Theorem 4.4]). Let $n \ge 1$ be an integer, $X = \sum_{i=1}^{n} X_i$ where X_i are independent Bernoulli random variables, and $\mu = \mathbb{E}[X]$. Then, for any real $\delta > 0$,

$$\mathbb{P}(X \geqslant (1+\delta)\mu) \leqslant \left(\frac{e^{\delta}}{(1+\delta)^{1+\delta}}\right)^{\mu}.$$

3 The q-Mallows Process & Technical Tools for Tangled Paths

In this section we describe a random process which generates a Mallows permutation. It is easier to prove results via this process rather than with $\mu_{n,q}$ directly as the process is driven by independent random input variables. We also introduce a special class of events related to the inputs of this process, and prove a concentration result for sums of indicators of these events.

3.1 The q-Mallows Process

In this section we introduce the q-Mallows process [5], a permutation-valued stochastic process $(r_n)_{n\geqslant 1}$, where each $r_n\in S_n$. In what follows assume q>0, and $q\neq 1$ (unless specified otherwise). The process is initialized by setting r_1 to be the (only) permutation on one element. The process iteratively constructs r_n from r_{n-1} and an independent random variable $v_n\sim \nu_{n,q}$, where $\nu_{n,q}$ is the truncated geometric distribution given by

$$\nu_{n,q}(j) = \frac{q^{j-1}}{1+q+\dots+q^{n-1}} = \frac{(1-q)q^{j-1}}{1-q^n} \quad (1 \le j \le n).$$
 (6)

For the case q=1 we define $\nu_{n,1}(j)=1/n$ for all $1 \leq j \leq n$. The random sequence $(r_n)_{n\geqslant 1}$ can now be defined inductively by sampling $v_n \sim \nu_{n,q}$ then setting

$$r_n(i) = \begin{cases} r_{n-1}(i) & i < v_n \\ n & i = v_n \\ r_{n-1}(i-1) & i > v_n \end{cases}$$
 (1 \leq i \leq n). (7)

To visualise this process: start with an empty bookshelf and at each time $i \ge 1$ insert the book with label i at position $v_i - 1$ then shift the remaining books one position to the right.

See Figure 2 for an example of the q-Mallows process (r_n) and the resulting tangled path. There is (at least) one other process which generates Mallows permutations, see

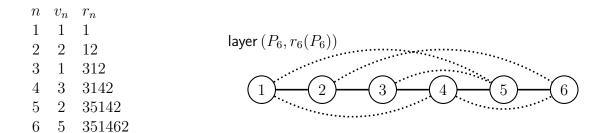


Figure 2: The table on the left gives the sequences of permutations (r_n) generated by the sequence (v_n) for i = 1, ..., 6. On the right we have a tangled path generated by r_6 , where the edges of $r_6(P_6)$ are dotted.

[5, Section 2] for more details. We use (r_n) as it is convenient for our proof methods, and often describe events in terms of the random sequence (v_n) generating the process (r_n) . The following lemma tells us that the reversed output permutations r_n^R has the Mallows distribution.

Lemma 8 ([5, Corollary 2.3]). Let q > 0 and (r_n) be the q-Mallows process (7). Then the permutation $\sigma \in S_n$ given by $\sigma(i) = r_n(n+1-i)$, for $1 \le i \le n$, has distribution $\mu_{n,q}$.

Recall that $\mathcal{P}(n,q)$ is the tangled path on n vertices with parameter q. Any edge in the tangled path $\mathcal{P}(n,q)$ must belong to one of the two paths P_n or $\sigma(P_n)$, thus

$$E(\mathcal{P}(n,q)) = \{i(i+1); 1 \le i < n\} \cup \{\sigma(i)\sigma(i+1) : 1 \le i < n\}.$$

However given a permutation σ , for any $1 \le i \le n-1$ there exists $1 \le j \le n-1$ such that $\{\sigma(i), \sigma(i+1)\} = \{\sigma^R(j), \sigma^R(j+1)\}$. It follows that

$$\mathcal{P}(n,q) \sim \mathsf{layer}(P_n, \sigma(P)) \sim \mathsf{layer}(P_n, \sigma^R(P_n)).$$
 (8)

The equation above shows that it does not matter if we use a (n, q)-Mallows permutation or its reverse (which is a (n, 1/q)-Mallows permutation). This justifies our use of the (unreversed) permutation r_n to generate $\mathcal{P}(n, q)$ and our restriction of q to the range $0 \leq q \leq 1$.

We now give some useful bounds on densities of the distribution $\nu_{n,q}$. Observe that if $v_k \sim \nu_{k,q}$ then for any integer $1 \leq x \leq k$, and 0 < q < 1, by (6) we have

$$\mathbb{P}(v_k \geqslant x) = \frac{(1-q)\left(q^{x-1} + \dots + q^{k-1}\right)}{1-q^k} = \frac{q^{x-1} - q^k}{1-q^k} = q^{x-1} \cdot \frac{1-q^{k-x+1}}{1-q^k} \leqslant q^{x-1}. \quad (9)$$

From this we can also obtain

$$\mathbb{P}(v_k \leqslant x) = 1 - \mathbb{P}(v_k > x) = 1 - \frac{q^x - q^k}{1 - q^k} = \frac{1 - q^x}{1 - q^k} \geqslant 1 - q^x. \tag{10}$$

Bhatnagar & Peled [5] proved the following tail bounds on the displacement of an element in under the action of a Mallows permutation.

Theorem 9 ([5, Theorem 1.1]). Let $1 \le i \le n$ and $t \ge 1$ be integers, and 0 < q < 1 be real. Then, if $\sigma \sim \mu_{n,q}$ we have $\mathbb{P}(|\sigma(i) - i| \ge t) \le 2q^t$.

3.2 Local Events and Concentration

As the q-Mallows process (r_i) is generated by a sequence (v_i) of independent random variables, it is convenient to describe events in terms of (v_i) . An example of such an event is the flush (13) given by $\mathcal{F}_k = \{\text{for each } i > k \text{ we have } v_i \leq i - k\}$. This event will be introduced formally in Section 3.3 and is closely related to the existence of cut vertices, see Section 5. In this section we are interested in events defined in terms of the sequence (v_i) which, unlike the flush, become pairwise (asymptotically) independent when the indices of the events are suitably well spaced. To begin we describe a class of events satisfying this criterion; we call these local events. We then show in Lemma 10 that sums of indicators of local events concentrate.

For integers $n \ge 1$, $i \in [n]$, and $\ell \ge 1$, we say that an event is ℓ -local to i if the event is completely determined by the values of v_j in the range $j \in [i-\ell, i+\ell]$. As an example the event $(\{v_i \le 5\} \cup \{v_{i+1} = 2\}) \cap \{v_{i-2} = 3\}$ is 2-local to i. We say a random variable X_i is ℓ -local to i if it is a weighted indicator random variable of an event \mathcal{E}_i which is ℓ -local to i, that is $X_i = c_i \cdot \mathbf{1}_{\mathcal{E}_i}$ where $c_i \ge 0$ are real numbers. We prove several results by showing that certain small subgraphs are present. Often the number of these subgraphs can be expressed as sums of local random variables, the following lemma, based on Chebyshev's inequality, is then useful.

Lemma 10. Let $n \ge 1$, $\ell \ge 1$ be integers and $S \subseteq [n]$ non-empty. Let $X = \sum_{i \in S} X_i$ where each X_i is a non-negative random variable ℓ -local to i and $M = \max_{i \in S} \mathbb{E}[X_i^2]$. Then, for any $x \ge 0$,

$$\mathbb{P}\Big(\left|X - \mathbb{E}\left[X\right]\right| > x \cdot \sqrt{M \cdot |S| \cdot \ell}\,\Big) \leqslant 5/x^2.$$

Proof. We seek to apply the second moment method to X, and thus we must calculate or bound terms of the form $\mathbb{E}[X_i \cdot X_j]$. For any $i, j \in S$ the Cauchy-Schwarz inequality gives

$$\mathbb{E}\left[X_i \cdot X_j\right] \leqslant \sqrt{\mathbb{E}\left[X_i^2\right] \cdot \mathbb{E}\left[X_j^2\right]} \leqslant M. \tag{11}$$

Observe, each X_i only depends on (at most) the values v_a for $a \in [i - \ell, i + \ell]$, and so

$$\mathbb{E}\left[X_i \cdot X_j\right] = \mathbb{E}\left[X_i\right] \cdot \mathbb{E}\left[X_j\right] \quad \text{for any } i, j \text{ with } |i-j| > 2\ell, \tag{12}$$

by independence of the sequence $(v_i)_{i=1}^n$. Thus, since $X_i \ge 0$, by (11) and (12) we have

$$\operatorname{Var}(X) = \sum_{i,j \in S} \mathbb{E}\left[X_i \cdot X_j\right] - \mathbb{E}\left[X_i\right] \mathbb{E}\left[X_j\right]$$

$$\leqslant \sum_{i,j \in S} \mathbf{1}_{\{|i-j| \leqslant 2\ell\}} \cdot M$$

$$\leqslant |S|(4\ell+1)M \leqslant 5|S|\ell M.$$

Thus applying Chebyshev's inequality [38, Theorem 3.6], for any $x \ge 0$, gives

$$\mathbb{P}\left(\left|X - \mathbb{E}\left[X\right]\right| \geqslant (x/\sqrt{5}) \cdot \sqrt{5|S|\ell M}\right) \leqslant \mathbb{P}\left(\left|X - \mathbb{E}\left[X\right]\right| \geqslant (x/\sqrt{5}) \cdot \sqrt{\operatorname{Var}(X)}\right) \leqslant (x/\sqrt{5})^{-2},$$

concluding the proof.

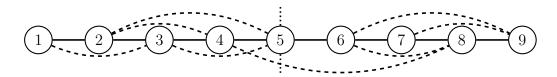


Figure 3: A representation of a flush event \mathcal{F}_5 (see (13)) holding in the graph layer $(P_9, \sigma(P_9))$. In this example $\sigma = (7, 9, 6, 8, 4, 2, 5, 3, 1)$ was generated by the sequence $\mathbf{x} = (1, 1, 2, 1, 3, 1, 1, 3, 2)$ satisfying \mathcal{F}_5 . Observe that only one edge crosses vertex 5.

3.3 Flushing Events

In this section we define two events with respect to the sequence of values $(v_i)_{i=1}^n$ generating the Mallows process (7). These events will help us describe certain graph properties and prove concentration using Lemma 10. First, we will briefly recall how the q-Mallows process $(r_i)_{i=1}^n$ evolves: At step $i \ge 1$ we have a permutation $r_i \in S_i$, starting from $r_1 = (1)$. Then, to generate the next permutation r_{i+1} in the sequence we sample a random variable $v_{i+1} \sim v_{i+1,q}$ and insert the value i+1 at relative position $v_{i+1} \in [1, i+1]$ from the left-hand side of r_i . To insert i+1 at v_{i+1} we shift each of the values in relative positions v_{i+1}, \ldots, i one place to the right.

For an integer $k \in [n]$ we define the *flush* event by

$$\mathcal{F}_k = \{ \text{for each } i > k \text{ we have } v_i \leqslant i - k \},$$
(13)

and say there is a flush at step k (of the q-Mallows process) if \mathcal{F}_k holds (see Figure 3).

If there is a flush at step k then no subsequent element with value greater than k is inserted into any of the k rightmost positions in the array. Thus, if \mathcal{F}_k holds, then for any $n \geq k$ the permutation $r_n = sr_k$ will be a concatenation of two strings s and r_k , where s is a permutation of the elements $\{k+1,\ldots,n\}$ and r_k is the state of the process at step k. This property is very useful if we are trying to find certain structures in the permutation r_n ; since if the structure appears in r_k and \mathcal{F}_k holds (these events are independent) then the elements inserted after k will not affect the structure in r_k . One simple example of this is that if $\{v_k=1\}\cap\mathcal{F}_k$ holds then vertex k is a cut vertex of $\mathcal{P}(n,q)$; see Section 5 for details and also precise bounds on $\mathbb{P}(\mathcal{F}_k)$). Thus, if we wish to show that there is a cut vertex w.h.p., then a standard approach would be to let K count the number of $K \in [n]$ such that $K \in [n$

Observe that, for $q \in (0,1)$, the position each element is inserted is biased towards the left-hand end. Suppose we can find a $b_n := b_n(q)$ such that w.h.p. for all $i \in [n]$ we have $v_i \leq b_n$. Then, we can define a new event \mathcal{L}_k which only specifies the first b_n values in the flush \mathcal{F}_k , and if we condition on \mathcal{L}_k then w.h.p. \mathcal{F}_k will also hold. To make this precise let

$$b_n = \left\lceil \frac{8\log n}{\log(1/q)} \right\rceil. \tag{14}$$

Then, as $\mathbb{P}(v_k > i) \leq q^i$ for i < k by (9), for $n \geq 2$ the union bound gives

$$\mathbb{P}\left(\bigcup_{k\in[n]} \{v_k > b_n\}\right) \leqslant \sum_{k\in[n]} \mathbb{P}(v_k > b_n) = \sum_{k=b_n+1}^n q^{b_n} \leqslant n \cdot q^{b_n} \leqslant n^{-7}.$$
 (15)

If we recall that v_i is the insert position of i relative to the left-hand end, then (15) shows that w.h.p. no element is inserted at relative position $r > b_n$. The constant 8 in (14) is fairly arbitrary but works in our analysis. Thus, conditional on the event $\{v_i \leq b_n \text{ for all } i \in [n]\}$, if an element i' < i is not within b_n places from the left-hand end of r_{i-1} then when element i is inserted it will not be adjacent to i'. This is significant since adjacency of i and i' in the final permutation r_n determines the presence of the edge ii' in $\mathcal{P}(n,q)$. So, if we want to preserve the subpermutation r_k w.h.p. then we do not necessarily need to condition on all elements i > k not being inserted in the rightmost k places, we only need to specify this for the next k elements. We can now define, for any $1 \leq k \leq n$, the local flush event given by

$$\mathcal{L}_k = \{ \text{ For each } k < i \leqslant k + b_n(q) \text{ we have } v_i \leqslant i - k \}.$$
 (16)

The local flush event \mathcal{L}_k captures the desirable property of preserving r_k , provided we condition on the event $\mathcal{E} = \bigcap_{i \in [n]} \{v_i \leqslant b_n(q)\}$. As \mathcal{E} occurs w.h.p. by (15), we can essentially use the local flush \mathcal{L}_k in the same way as \mathcal{F}_k . The advantage of \mathcal{L}_k is that it only fixes the positions of elements $k+1,\ldots,k+b_n$, as opposed to the positions of all elements i > k in \mathcal{F}_k , and thus it is $\lceil b_n/2 \rceil$ -local with respect to $k + \lceil b_n/2 \rceil$. The next result 'localises' events involving flushes.

Lemma 11. Let $n, \ell \geqslant 1$ be integers, 0 < q < 1, and $(c_i)_{i \in [n]}$ be any non-negative real sequence. Let $\mathcal{E}_i = \mathcal{B}_i \cap \mathcal{F}_i$ where \mathcal{B}_i is ℓ -local to i. Then, the event $\mathcal{D}_i = \mathcal{B}_i \cap \mathcal{L}_i$ is $\max\{b_n, \ell\}$ -local to i, where $b_n := b_n(q)$ is given by (14). Furthermore, if $X = \sum_i c_i \cdot \mathbf{1}_{\mathcal{E}_i}$ and $Y = \sum_i c_i \cdot \mathbf{1}_{\mathcal{D}_i}$, then

$$\mathbb{P}(X \neq Y) \leqslant n^{-7}, \quad and \quad \mathbb{E}[X] \leqslant \mathbb{E}[Y] \leqslant \mathbb{E}[X] + n^{-7} \cdot \sum_{i} c_{i}.$$

Proof. The event \mathcal{D}_i is $\max\{b_n, \ell\}$ -local to i as \mathcal{B}_i is ℓ -local to i and \mathcal{L}_i , given by (16), is b_n -local to i. Recall $v_i \sim \nu_{i,q}$ for $i \in [n]$ and $\mathcal{E} = \bigcap_{i \in [n]} \{v_i \leqslant b_n(q)\}$. Note that $\mathbb{P}(\mathcal{E}^c) \leqslant n^{-7}$ by (15), and although (15) only holds for $n \geqslant 2$ we can assume this or else $\mathbb{P}(\mathcal{E}^c) = 0$ trivially.

Observe that $\mathcal{E}_i \subseteq \mathcal{D}_i$ for any $i \in [n]$. Furthermore, the events \mathcal{E}_i and \mathcal{D}_i impose the same restrictions on the random variables $(v_j)_{j \leqslant i+b_n(q)}$. In particular, $(v_j)_{i < j \leqslant i+b_n(q)}$ satisfy both the flush \mathcal{F}_i and local flush \mathcal{L}_i event. However, conditional on \mathcal{E} no random variable v_i is larger than $b_n(q)$ and so the remaining values $(v_j)_{j>i+b_n}$ also satisfy \mathcal{F}_i . It follows that $\mathcal{E} \cap \mathcal{D}_i \subseteq \mathcal{E}_i$ for any $i \in [n]$ and hence $\mathbb{P}(X \neq Y) \leqslant \mathbb{P}(\mathcal{E}^c)$.

For the upper bound in the last part of the claim note $Y = \sum_i c_i \cdot \mathbf{1}_{\mathcal{E}_i} \leqslant \sum_i c_i$, thus

$$\mathbb{E}\left[\,Y\,\right] = \mathbb{E}\left[\,X \cdot \mathbf{1}_{\{Y = X\}}\,\right] + \mathbb{E}\left[\,Y \mid Y \neq X\,\right] \cdot \mathbb{P}(\,Y \neq X\,) \leqslant \mathbb{E}\left[\,X\,\right] + \left(\sum_{i} c_{i}\right) \cdot n^{-7}.$$

The lower bound $\mathbb{E}[Y] \geqslant \mathbb{E}[X]$ holds since $\mathcal{D}_i \supseteq \mathcal{E}_i$ and $c_i \geqslant 0$ for any $i \in [n]$.

4 Expansion and Relations Between q-Mallows Measures

This section concerns $\mathcal{P}(n,q)$ with q sufficiently close to 1. In Lemma 13 we will prove that $\mathcal{P}(n,1)$ is an expander with probability $1 - e^{-\Omega(n)}$. We then establish Lemma 14 which allows us to relate properties of $\mathcal{P}(n,q)$ to those of $\mathcal{P}(n,1)$. This is then used to extend Lemma 13 to Lemma 16, which shows that, if q is sufficiently close to 1, then w.h.p. $\mathcal{P}(n,q)$ is an expander.

As mentioned in the introduction, it is possible to show that $\mathcal{P}(n,1)$ is an expander with probability 1-o(1) by adapting a result of Kim & Wormald [31, Theorem 1], however it is the fact that Lemma 13 holds with probability $1-\mathrm{e}^{-\Omega(n)}$ which allows us to relate this to a smaller q in Lemma 16. This combination of Lemmas 13 and 14 is also used to prove lower bounds on the treewidth in Section 6 by finding a subdivided expander on roughly 1/(1-q) vertices as a subgraph in $\mathcal{P}(n,q)$. Again, if we knew only that the $\mathcal{P}(n,1)$ was an expander with probability 1-o(1), then we would not be able to get such good bounds later in Section 6.

4.1 Expansion in the case q = 1

Recall the definition of $\partial(S) = \{uv \in E(G) : u \in S, v \in V(G) \setminus S\}$, the edge boundary of $S \subseteq V$, from Section 2.1. The following bound is quite crude but (crucially) it is independent of |S|.

Lemma 12. For any integer $1 \le k \le n-1$ there are at most $2 \cdot \binom{n-1}{k}$ distinct vertex subsets S of an n-vertex path such that $|\partial(S)| = k$.

Proof. Observe that there are $\binom{n-1}{k}$ ways to choose the k edges of the boundary set $\partial(S)$. Each set $\partial(S)$ of boundary edges gives two possible sets S depending on whether the first vertex of the path is in S or not.

Recall the definitions of the edge and vertex isoperimetric numbers ϕ and φ from Section 2.1. We now prove that w.h.p. a tangled path generated from a uniform permutation is an expander.

Lemma 13. For any integer $n \ge 100$, we have

$$\mathbb{P}\bigg(\varphi(\mathcal{P}(n,1)) \leqslant \frac{1}{40}\bigg) \leqslant 1000 \cdot n^{7/2} \cdot \left(\frac{3}{4}\right)^n.$$

Proof. Observe that $\varphi(\mathcal{P}(n,1)) \ge \phi(\mathcal{P}(n,1))/4$ by (3) since $\mathcal{P}(n,1)$ has degree at most 4. Thus, to prove this Lemma it suffices to bound $\mathbb{P}(\phi(\mathcal{P}(n,1)) \le 1/10)$ from above.

Let $\mathcal{P} = \mathsf{layer}(P_n, \sigma(P_n))$ where $\sigma \sim \mu_{n,1}$. Now, if the edge isoperimetric number of \mathcal{P} is at most α then for any set $S \subset [n]$ there can be at most $\alpha |S|$ edges of either the permuted or un-permuted path in $\partial(S)$. That is, for any $\alpha > 0$ we have

$$\{\phi(\mathcal{P}) \leqslant \alpha\} \subseteq \bigcup_{s=1}^{\lfloor n/2 \rfloor} \bigcup_{k=1}^{\lfloor \alpha s \rfloor} \bigcup_{S \subseteq V: |S|=s, |\partial(S) \cap E(P_n)|=k} \{|\partial(S) \cap E(\sigma(P_n))| \leqslant \lfloor \alpha s \rfloor - k\}. \tag{17}$$

By Lemma 12 there are at most $2 \cdot {n-1 \choose k}$ sets S (of any size) such that $|\partial(S) \cap E(P_n)| = k$. So by applying this bound, the union bound, and rewriting (17), where we ignore the -k, we have

$$\mathbb{P}(\phi(\mathcal{P}) \leqslant \alpha) \leqslant 2 \sum_{s=1}^{\lfloor n/2 \rfloor} \sum_{k=1}^{\lfloor \alpha s \rfloor} \binom{n-1}{k} \max_{S \subseteq V: |S| = s} \mathbb{P}(|\partial(S) \cap E(\sigma(P_n))| \leqslant \lfloor \alpha s \rfloor). \tag{18}$$

We will now bound the probability on the right-hand side, note that since σ is a uniform permutation this is the same for all S with |S| = s.

To begin we will view the action of the uniform random permutation σ on P_n as a relabelling of $V(P_n) = [n]$. Under this relabelling a given set $S \subseteq V(P_n)$, with |S| = s, is equally likely to be mapped to any other $S' \subseteq V(P_n)$ with |S'| = s. The number of $S' \subseteq V(P_n)$ with an edge boundary of size $k \ge 0$ is at most $2 \cdot \binom{n-1}{k}$ by Lemma 12. Since |S'| = s there are s! ways to order the elements of S within S' and (n-s)! ways to organise the remaining elements within the path. Thus, if we restrict to a fixed $0 < \alpha \le 1/10$, then for any $S \subseteq V$ with |S| = s we have

$$\mathbb{P}(|\partial(S) \cap E(\sigma(P_n))| \leqslant \lfloor \alpha s \rfloor) \leqslant 2 \sum_{k=1}^{\lfloor \alpha s \rfloor} \binom{n-1}{k} \frac{s!(n-s)!}{n!} \leqslant 2n \binom{n-1}{\lfloor \alpha s \rfloor} \frac{s!(n-s)!}{n!}, \quad (19)$$

where the last bound holds since $\alpha s \leq n/20$ and $\binom{x}{y}$ is increasing in y provided $y < \lfloor x/2 \rfloor$. Recall that for any integers $1 \leq k \leq n$ and any real number x such that $|x| \leq n$,

$$\binom{n}{k} \leqslant \left(\frac{ne}{k}\right)^k, \quad \sqrt{2\pi n} \left(\frac{n}{e}\right)^n \leqslant n! \leqslant e\sqrt{n} \left(\frac{n}{e}\right)^n, \quad \text{and} \quad \left(1 + \frac{x}{n}\right)^n \leqslant e^x.$$
 (20)

Thus, as $s \leq \lfloor n/2 \rfloor$ and $0 < \alpha \leq 1/10$ are fixed, by (20) and monotonicity of $\binom{x}{y}$ we have

$$\begin{pmatrix} n-1 \\ \lfloor \alpha s \rfloor \end{pmatrix} \leqslant \left(\frac{ne}{\lfloor \alpha \lfloor \frac{n}{2} \rfloor \rfloor} \right)^{\lfloor \alpha \lfloor \frac{n}{2} \rfloor \rfloor}
\leqslant \left(\frac{2e}{\alpha \left(1 - \frac{4}{\alpha n} \right)} \right)^{\frac{\alpha n}{2}}
\leqslant \left(e^{-\frac{4}{\alpha}} \right)^{-\frac{\alpha}{2}} \left(\frac{2e}{\alpha} \right)^{\frac{\alpha n}{2}} = e^{2} \left(\frac{2e}{\alpha} \right)^{\frac{\alpha n}{2}},$$
(21)

provided that $n > 4/\alpha$. Observe that s = n/2 maximises $s^s (n-s)^{n-s}$, thus (20) gives

$$s! \cdot (n-s)! \leqslant e\sqrt{s} \left(\frac{s}{e}\right)^s \cdot e\sqrt{n-s} \left(\frac{n-s}{e}\right)^{n-s} \leqslant \frac{e^2n}{e^n} \cdot s^s (n-s)^{n-s} \leqslant \frac{e^2n}{e^n} \cdot \left(\frac{n}{2}\right)^n. \tag{22}$$

For $n > 4\alpha$, inserting the bounds from (20), (21) and (22) into (18) and (19) gives

$$\mathbb{P}(\phi(\mathcal{P}) \leqslant \alpha) \leqslant 2 \sum_{n=1}^{\lfloor n/2 \rfloor} \sum_{k=1}^{\lfloor \alpha s \rfloor} \binom{n-1}{k} \cdot 2n \binom{n-1}{\lfloor \alpha s \rfloor} \frac{s!(n-s)!}{n!}$$

$$\leqslant 2n^{2}e^{2} \left(\frac{2e}{\alpha}\right)^{\frac{\alpha n}{2}} \cdot 2ne^{2} \left(\frac{2e}{\alpha}\right)^{\frac{\alpha n}{2}} \frac{\frac{e^{2}n}{e^{n}} \cdot \left(\frac{n}{2}\right)^{n}}{\sqrt{2\pi n} \left(\frac{n}{e}\right)^{n}}$$

$$= \frac{4e^{6}}{\sqrt{2\pi}} \cdot n^{7/2} \cdot \left(\frac{2e}{\alpha}\right)^{\alpha n} 2^{-n}.$$

One can check that if we fix $\alpha = 1/10$ then $\left(\frac{2e}{\alpha}\right)^{\alpha} \approx 1.4912 < 3/2$. Thus taking $\alpha = 1/10$ gives $\mathbb{P}(\phi(\mathcal{P}) \leq 1/10) \leq 1000n^{7/2}(3/4)^n$ for $n \geq 100$ as $4e^6/\sqrt{2\pi} < 1000$.

4.2 Relating Different q-Mallows Measures

The aim of this section is to prove the following result which allows us to relate properties satisfied by the non-uniform tangled path $(q \neq 1)$ to those satisfied in the uniform case. Let $2^{\binom{n}{2}}$ denote the set of all (labelled) *n*-vertex graphs.

Lemma 14. Let $n \ge 1$ be an integer, and 0 < q < 1. Then, the following holds:

- (i) For any $A \subseteq S_n$, we have $\mu_{n,q}(A) \leqslant e^{9n^2(1-q)} \cdot \mu_{n,1}(A)$.
- (ii) For any $B \subseteq 2^{\binom{n}{2}}$, we have $\mathbb{P}(\mathcal{P}(n,q) \in B) \leqslant e^{9n^2(1-q)} \cdot \mathbb{P}(\mathcal{P}(n,1) \in B)$.

(iii) If
$$q = 1 - o(n^{-2})$$
, then for any $B \subseteq 2^{\binom{n}{2}}$, $\mathbb{P}(\mathcal{P}(n,q) \in B) = \mathbb{P}(\mathcal{P}(n,1) \in B) + o(1)$.

In order to prove this result we must first prove a lemma which bounds the ratio between densities of the truncated geometric and uniform distributions.

Lemma 15. For any integers $1 \le i \le k$, and 0 < q < 1, we have

$$\frac{\nu_{k,q}(i)}{\nu_{k,1}(i)} \le 1 + 9k(1-q) \le e^{9k(1-q)}.$$

Proof. For ease of notation we will sometimes use the parametrisation q = 1 - x. By the definition (6) of $\nu_{k,q}$, for any $i \leq k$ we have

$$\nu_{k,q}(i) = \frac{(1-q)q^i}{1-q^k} = \frac{x(1-x)^i}{1-(1-x)^k}.$$
(23)

We need a 'reverse Bernoulli inequality': for any integer $r \ge 1$ and real y satisfying $|y| \le 1/(2r)$, we have

$$(1+y)^r = \sum_{i=0}^r \binom{r}{i} y^i \leqslant 1 + ry + \sum_{i=2}^r (ry)^i \leqslant 1 + ry + (ry)^2 \sum_{i=0}^{r-2} 2^{-i} \leqslant 1 + ry + 2(ry)^2.$$
 (24)

We will now begin with the case $1 - 1/(4k) \le q < 1$. Applying (24) to (23), for $|x| \le 1/(4k)$ and $i \le k$, gives

$$\nu_{k,q}(i) \leqslant \frac{x(1-ix+2(ix)^2)}{1-(1-kx+2(kx)^2)} = \frac{1}{k} \cdot \frac{1-ix+2(ix)^2}{1-2kx} = \frac{1}{k} \cdot \left(1 + \frac{2kx-ix+2(ix)^2}{1-2kx}\right).$$

Recall that $\nu_{k,1}(i) = 1/k$ for all $1 \le i \le k$. Thus, by the above and $|x| \le 1/(4k)$, we have

$$\frac{\nu_{k,q}(i)}{\nu_{k,1}(i)} \le 1 + \frac{2kx + 2(kx) \cdot (1/4)}{1 - 2/4} = 1 + 5kx = 1 + 5k(1 - q). \tag{25}$$

We now deal with the remaining case 0 < q < p where p := 1 - 1/(4k). By (23),

$$\frac{\nu_{k,q}(i)}{\nu_{k,p}(i)} = \frac{(1-q)q^i}{1-q^k} \cdot \frac{1-p^k}{(1-p)p^i} = \frac{1-q}{1-p} \cdot \left(\frac{q}{p}\right)^i \cdot \frac{1-p^k}{1-q^k} \leqslant \frac{1-q}{1-p} = 4k(1-q). \tag{26}$$

Thus by (25) and (26), for any $0 < q \le 1 - 1/(4k)$ we have

$$\frac{\nu_{k,q}(i)}{\nu_{k,1}(i)} = \frac{\nu_{k,q}(i)}{\nu_{k,p}(i)} \cdot \frac{\nu_{k,p}(i)}{\nu_{k,1}(i)} = 4k(1-q) \cdot (1+5k\cdot 1/(4k)) = 9k(1-q). \tag{27}$$

The first inequality in the statement follows as 1 + 9k(1 - q) is an upper bound for (25) and (27). Then the second follows since $1 + y \leq e^y$ for all y.

Using this lemma we can now prove the main result in this section.

Proof of Lemma 14. By the description of the q-Mallows process we see that for every integer $n \ge 1$, $0 < q \le 1$, and $A \subset S_n$, there exists a set $\{(x_k^i)_{k=1}^n : i \in \mathcal{I}\}$ of inputs to the q-Mallows process such that

$$\mu_{n,q}(A) = \sum_{i \in \mathcal{I}} \prod_{k \in [n]} \nu_{k,q}(x_k^i).$$

Furthermore, since the q-Mallows is a deterministic function of random inputs, the set $\{(x_k^i)_{k=1}^n : i \in \mathcal{I}\}$ does not depend on q. Thus, by Lemma 15, for any event $A \subseteq S_n$,

$$\mu_{n,q}(A) = \sum_{i \in \mathcal{I}} \prod_{k \in [n]} \nu_{k,q}(x_i^j) \leqslant \sum_{i \in \mathcal{I}} \prod_{k \in [n]} e^{9k(1-q)} \cdot \nu_{k,1}(x_i^j).$$

Then, using the bound $k \leq n$, we can deduce Item (i) since

$$\mu_{n,q}(A) \leqslant e^{9n^2(1-q)} \cdot \sum_{i \in \mathcal{I}} \prod_{k \in [n]} \nu_{k,1}(x_i^j) = e^{9n^2(1-q)} \cdot \mu_{n,1}(A).$$

Item (ii): Recall that the tangled path $\mathcal{P}(n,q)$ is generated deterministically from a Mallows permutation $\sigma \sim \mu_{n,q}$ by the construction layer $(\sigma(P_n), P_n)$. Thus for any $B \subseteq 2^{\binom{n}{2}}$ there is a corresponding set of permutations $A \subseteq S_n$ such that layer $(\sigma(P_n), P_n) \in B$ if and only if $\sigma \in A$. The result now follows from Item (i).

Item (iii): Since $1 - q = o(n^{-2})$, for any $B \subseteq 2^{\binom{n}{2}}$ we have

$$\mathbb{P}(\mathcal{P}(n,q) \in B) \leqslant e^{9n^2(1-q)} \cdot \mathbb{P}(\mathcal{P}(n,1) \in B)$$

$$= e^{o(1)} \cdot \mathbb{P}(\mathcal{P}(n,1) \in B)$$

$$= \mathbb{P}(\mathcal{P}(n,1) \in B) + o(1),$$

by Item (ii) and as $e^{o(1)} = 1 + o(1)$ by the Taylor expansion.

4.3 Proof of Theorem 1

We can now apply Lemmas 13 and 14 to prove Lemma 16. Theorem 1 is simply a less explicit restatement of Lemma 16, so it follows directly from this.

Lemma 16. If
$$q \ge 1 - \frac{1}{50n}$$
, then $\mathbb{P}(\varphi(\mathcal{P}(n,q)) \le \frac{1}{40}) = o(1)$.

Proof. Let $B = \{G : \varphi(G) \leq 1/40\} \subseteq 2^{\binom{n}{2}}$. Then Lemmas 14(ii) and 13 give

$$\mathbb{P}(\mathcal{P}(n,q) \in B) \leqslant e^{9n^2(1-q)} \cdot \mathbb{P}(\mathcal{P}(n,1) \in B) \leqslant e^{n/5} \cdot 1000n^{7/2} \left(\frac{3}{4}\right)^n = o(1),$$

since
$$\ln(3/4) < -1/4$$
.

We note that the constant 1/50 in the assumption on q in Lemma 16 has not been optimised. However, we believe that $q = 1 - \Theta(1/n)$ should be the threshold for $\mathcal{P}(n,q)$ being an expander.

5 Cut Vertices and Diameter

The main aim of this section is to prove Theorem 2 which shows that $q = 1 - \frac{\pi^2}{6 \log n}$ is a sharp threshold for the existence of a cut vertex. This is achieved by equating the existence of a cut vertex to a combination of flush events, then bounding the number of cut vertices. We then use bounds on the number of cut vertices obtained while proving Theorem 2 to establish Theorem 3, which shows that the diameter is linear when 0 < q < 1 is fixed.

5.1 Relating Unit Separators to Flush Events

Given a graph G = (V, E) we say that a vertex v is a cut vertex in G if its removal separates the graph into two or more disjoint components. The aim of this section is to prove Lemma 17 which shows that cut vertices are determined by flush events.

Let $1 \leq k \leq n$ be integers and $v_i \sim \nu_{i,q}$ for all $i \in [n]$. Recall the flush event given by (13):

$$\mathcal{F}_k = \{ \text{For each } i > k \text{ we have } v_i \leqslant i - k \}.$$

For an integer $k \in [n]$ we define also define the reverse flush event by

$$\mathcal{R}_k = \{ \text{for each } i > k \text{ we have } v_i > k \}.$$

The name we give to this event is quite fitting: Suppose the flush event \mathcal{F}_k holds in a permutation σ and let σ^R be the reverse of σ . Then, the reverse flush event \mathcal{R}_k holds for σ^R .

We define two events $C_k^{\mathcal{F}}$ and $C_k^{\mathcal{R}}$ which stipulate that k separates all elements i > k from all elements i < k in σ . In $C_k^{\mathcal{F}}$ the elements i > k are to the left of k in σ , and in $C_k^{\mathcal{R}}$ they are to the right of k in σ (see Figure 4). These events are disjoint for n > 1, and are given by

$$C_k^{\mathcal{F}} = \mathcal{F}_k \cap \{v_k = 1\}$$
 and $C_k^{\mathcal{R}} = \mathcal{R}_k \cap \{v_k = k\}.$ (28)

Our next result shows that together these two events characterise cut vertices in $\mathcal{P}(n,q)$.

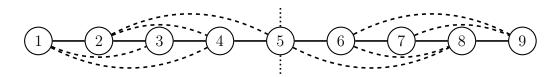


Figure 4: A representation of the event $C_5^{\mathcal{F}}$ from (28) holding in the graph layer $(P_9, \sigma(P_9))$. In this example $\sigma = (7, 9, 6, 8, 5, 2, 4, 1, 3)$ was generated by the sequence $\mathbf{x} = (1, 1, 3, 2, 1, 1, 1, 3, 2)$ satisfying $C_5^{\mathcal{F}}$. Observe that vertex 5 is a cut vertex.

Lemma 17. Let $n \ge 3$ be an integer. Then, a vertex $k \in \{2, ..., n-1\}$ is a cut vertex in $\mathcal{P}(n,q)$, if and only if one of the disjoint events $\mathcal{C}_k^{\mathcal{F}}$ or $\mathcal{C}_k^{\mathcal{R}}$ holds.

Proof. We consider the tangled path $\mathsf{layer}(P_n, r_n(P_n))$ with $r_n \sim \mu_{n,q}$ generated from the random sequence $(v_i)_{i=1}^n$, where $v_i \sim \nu_{i,q}$, via the q-Mallows process. Recall from (7) that given $(v_i)_{i=1}^n$ we generate the sequence $(r_i)_{i=1}^n$ inductively by the rule

$$r_i(j) = \begin{cases} r_{i-1}(j) & j < v_i \\ i & j = v_i \\ r_{i-1}(j-1) & i > v_j \end{cases}$$
 $(1 \le j \le i).$

First assume that $C_k^{\mathcal{F}}$ holds for some $k \in \{2, \ldots, n-1\}$. Conditional on $\{v_k = 1\}$ we have $r_k(1) = k$ and $r_k(i) < k$ for all i > 1. Once k is inserted then, conditional on \mathcal{F}_k , each i > k the element i is inserted to the left of k in the permutation (r_{i-1}) . It follows that, conditional on $C_k^{\mathcal{F}}$, the final permutation r_n satisfies $r_k(i) > k$ for all $i \leq n - k$, $r_n(n-k+1) = k$, and $r_n(j) < k$ for all j > n-k+1. Thus, since no element i > k is adjacent to any element j < k in r_n and thus k is a cut vertex of \mathcal{P} . See Figure 4 for an example. We now assume $C_k^{\mathcal{R}}$ holds. Conditional on $\{v_k = k\}$ we have $r_k(k) = k$ and $r_k(i) < k$ for all i < k. Then, conditioning on \mathcal{R}_k , each element i > k is inserted to the right of k. So, conditional on $C_k^{\mathcal{R}}$, we have $r_k(i) < k$ for all i < k, $r_n(k) = k$ and $r_n(j) > k$ for all j > k. Thus, as before, k is a cut vertex of \mathcal{P} .

For the other direction suppose $k \in \{2, \ldots, n-1\}$ is a cut vertex. Then, since \mathcal{P} contains a path on [n] as a subgraph, k's removal must separate the graph into vertex sets $\{1, \ldots, k-1\}$ and $\{k+1, \ldots, n\}$ and there can be no element j > k adjacent to any i < k in r_n . By the definition of $(r_i)_{i=1}^n$, at the time k is inserted all elements i < k have been inserted. If k is a cut then either $\{v_k = 1\}$ or $\{v_k = k\}$ holds, since otherwise when k+1 is inserted it will be adjacent to at least one element j < k. Now if $\{v_k = 1\}$ holds then all elements i > k must be inserted to the left of k, this is precisely the event \mathcal{F}_k . Otherwise if $\{v_k = k\}$ holds then all i > k must be inserted right of k, this gives \mathcal{R}_k . Hence, if k is a cut then either $\mathcal{C}_k^{\mathcal{F}}$ or $\mathcal{C}_k^{\mathcal{R}}$ holds.

5.2 Bounds on the Probability of Flush Events

Our first result in this section gives algebraic expressions for $\mathbb{P}(\mathcal{F}_k)$ and $\mathbb{P}(\mathcal{R}_k)$.

Lemma 18. For integers $1 \le k \le n$, and 0 < q < 1, we have

$$\mathbb{P}(\mathcal{F}_k) = \prod_{i=1}^k \frac{1 - q^i}{1 - q^{n-k+i}} \quad and \quad \mathbb{P}(\mathcal{R}_k) = q^{k(n-k)} \cdot \mathbb{P}(\mathcal{F}_k).$$

Proof. Observe that for each i > k we have $\mathbb{P}(v_i \leqslant i - k) = \frac{1 - q^{i-k}}{1 - q^i}$ by (10), thus

$$\mathbb{P}(\mathcal{F}_k) = \prod_{i=k+1}^n \mathbb{P}(v_i \leqslant i - k)
= \prod_{i=k+1}^n \frac{1 - q^{i-k}}{1 - q^i}
= \left(\prod_{i=k+1}^n (1 - q^i)\right)^{-1} \cdot \prod_{i=k+1}^n (1 - q^{i-k}),$$
(29)

where the first equality holds since (v_i) are independent. Shifting some indices gives

$$\mathbb{P}(\mathcal{F}_k) = \left(\prod_{i=1}^k (1 - q^{n-k+i}) \cdot \prod_{i=k+1}^{n-k} (1 - q^i)\right)^{-1} \cdot \prod_{i=1}^{n-k} (1 - q^i) = \prod_{i=1}^k \frac{1 - q^i}{1 - q^{n-k+i}},$$

as claimed. Now, again by (9), we have $\mathbb{P}(v_i > k) = \frac{q^k - q^i}{1 - q^i}$ for any i > k and thus

$$\mathbb{P}(\mathcal{R}_k) = \prod_{i=k+1}^n \mathbb{P}(v_i > k) = \prod_{i=k+1}^n \frac{q^k - q^i}{1 - q^i} = q^{k(n-k)} \prod_{i=k+1}^n \frac{1 - q^{i-k}}{1 - q^i} = q^{k(n-k)} \cdot \mathbb{P}(\mathcal{F}_k),$$

where the last equality follows from (29).

The following technical lemma is used for obtaining tight bounds on $\mathbb{P}(\mathcal{F}_k)$.

Lemma 19. For any 0 < q < 1, we have

$$\frac{q \log q}{6(1-q)} \leqslant \sum_{i=1}^{\infty} \log(1-q^i) - \frac{\pi^2}{6 \log q} - \frac{3 \log(1-q)}{2} \leqslant -\frac{1-q}{q \log q}.$$

Proof. Recall the Euler-Maclaurin summation formula [25, Section 9.5]: for any real function f(x) with derivative f'(x) and any integers $-\infty < a < b < \infty$ we have

$$\sum_{a \le i \le b} f(i) - \int_a^b f(x) \, \mathrm{d}x + \frac{1}{2} f(x) \Big|_a^b = \frac{1}{12} f'(x) \Big|_a^b + R_2, \tag{30}$$

where $f(x)\Big|_a^b = f(b) - f(a)$ and R_2 is real and satisfies $|R_2| \leqslant \frac{1}{12} |f'(x)|_a^b$. In our case we set $f(x) = \log(1 - q^x)$, thus

$$\frac{1}{2}f(x)\Big|_a^b = -\frac{1}{2}\log\left(\frac{1-q^a}{1-q^b}\right).$$

The derivative of f is

$$f'(x) = -q^x \log(q)/(1 - q^x) = \log(1/q) \left(1/(1 - q^x) - 1\right) > 0.$$

Notice that f'(x) is decreasing in x, thus as f'(a) > f'(b) since b > a, we have

$$0 \geqslant \frac{1}{12}f'(x)\Big|_a^b + R_2 \geqslant \frac{1}{6}f'(x)\Big|_a^b = \frac{q^a \log q}{6(1 - q^a)} - \frac{q^b \log q}{6(1 - q^b)}.$$

So by (30) and setting a = 1, for any $b \ge 1$ we have

$$\frac{q \log q}{6(1-q)} - \frac{q^b \log q}{6(1-q^b)} \leqslant \sum_{1 \leqslant i \leqslant b} \log(1-q^i) - \int_1^b \log(1-q^x) \, \mathrm{d}x - \frac{1}{2} \log\left(\frac{1-q}{1-q^b}\right) \leqslant 0.$$

Now if we take $b \to \infty$ then the integral and sum converge since q < 1. Thus, we have

$$\frac{q \log q}{6(1-q)} \leqslant \sum_{i=1}^{\infty} \log(1-q^i) - \int_{1}^{\infty} \log(1-q^x) \, \mathrm{d}x - \frac{\log(1-q)}{2} \leqslant 0. \tag{31}$$

Observe that the substitution $y = q^x$ yields

$$\int_{1}^{\infty} \log(1 - q^{x}) dx = \frac{1}{\log q} \int_{0}^{q} \frac{\log(1 - y)}{y} dy.$$
 (32)

The right-hand side of (32) contains the Dilogarithm function $Li_2(q)$ given by

$$\operatorname{Li}_{2}(q) := \int_{0}^{q} \frac{\log(1-y)}{y} \, \mathrm{d}y = \sum_{j=1}^{\infty} \frac{q^{j}}{j^{2}}, \quad \text{for } 0 \leqslant q \leqslant 1,$$
 (33)

where the second equality is by [35, (1.3)], thus $\text{Li}_2(1) = \pi^2/6$. Integration by parts gives

$$\int_0^q \frac{\log(1-y)}{y} \, dy = \ln(y) \ln(1-y) \Big|_0^q + \int_0^q \frac{\log(y)}{1-y} \, dy$$
$$= \ln(q) \ln(1-q) + \int_{1-q}^1 \frac{\log(1-x)}{x} \, dx,$$

where we used the substitution x = 1 - y. By (33) and the above we have

$$\int_0^q \frac{\log(1-y)}{y} \, dy = \ln(q) \ln(1-q) + \text{Li}_2(1) - \text{Li}_2(1-q)$$
$$= \ln(q) \ln(1-q) + \frac{\pi^2}{6} - \sum_{j=1}^\infty \frac{(1-q)^j}{j^2}.$$

Note that $0 \leqslant \sum_{j=1}^{\infty} \frac{(1-q)^j}{j^2} \leqslant \sum_{j=1}^{\infty} (1-q)^j = \frac{1-q}{q}$. The result then follows from (31).

We can now apply this approximation to prove bounds on $\mathbb{P}(\mathcal{F}_k)$.

Lemma 20. For any integers $1 \le k \le n$, and 0 < q < 1, we have

$$\frac{q \log q}{6(1-q)} \leqslant \log \mathbb{P}(\mathcal{F}_k) - \frac{\pi^2}{6 \log q} - \frac{3 \log(1-q)}{2} \leqslant \frac{2q^{\min\{n-k,k\}}}{(1-q)(1-q^{\min\{n-k,k\}})} - \frac{1-q}{q \log q}.$$

Proof. By Lemma 18 we have $\mathbb{P}(\mathcal{F}_k) = \prod_{i=1}^k \frac{1-q^i}{1-q^{n-k+i}}$ for any $1 \leqslant k \leqslant n$. Thus

$$\log \mathbb{P}(\mathcal{F}_k) = \sum_{i=1}^k \log (1 - q^i) - \sum_{i=1}^k \log (1 - q^{n-k+i})$$

$$= \sum_{i=1}^\infty \log (1 - q^i) - \sum_{i=1}^\infty \log (1 - q^{k+i}) - \sum_{i=1}^k \log (1 - q^{n-k+i}).$$
(34)

Observe that $-\sum_{i=1}^k \log(1-q^{n-k+i}) \ge 0$. As $\log(1-x) \ge -x/(1-x)$ for all x < 1,

$$-\sum_{i=1}^{k} \log \left(1 - q^{n-k+i}\right) \leqslant \sum_{i=1}^{k} \frac{q^{n-k+i}}{1 - q^{n-k+i}} \leqslant \frac{1}{1 - q^{n-k+1}} \sum_{i=1}^{k} q^{n-k+i} \leqslant \frac{q^{n-k+1}}{(1 - q)(1 - q^{n-k+1})}.$$

Similarly, we have $0 \leqslant -\sum_{i=1}^{\infty} \log (1-q^{k+i}) \leqslant \frac{q^{k+1}}{(1-q)(1-q^{k+1})}$. Thus

$$0 \leqslant -\sum_{i=1}^{k} \log \left(1 - q^{n-k+i}\right) - \sum_{i=1}^{\infty} \log \left(1 - q^{k+i}\right) \leqslant \frac{2q^{\min\{n-k,k\}}}{(1-q)(1-q^{\min\{n-k,k\}})}.$$

The result now follows by inserting these bounds into (34) and then using Lemma 19 to bound the remaining sum in (34).

The next lemma gives a better bound on $\mathbb{P}(\mathcal{F}_k)$ than Lemma 20 when q is close to 1.

Lemma 21. For any integers $1 \le k \le n$, and 0 < q < 1, we have

$$\mathbb{P}(\mathcal{F}_k) \leqslant \exp\left(-\frac{q\left(1-q^{\min\{k,n-k\}}\right)}{2(1-q)}\right).$$

Proof. By (10), $\mathbb{P}(v_i \leq i - k) = 1 - \frac{q^{i-k} - q^i}{1 - q^i}$. Independence of the random variables v_i yields

$$\mathbb{P}(\mathcal{F}_k) = \prod_{i=k+1}^n \mathbb{P}(v_i \leqslant i - k) = \prod_{i=1}^{n-k} \left(1 - \frac{q^i - q^{k+i}}{1 - q^{k+i}}\right) \leqslant \prod_{i=1}^{n-k} \exp\left(-\frac{q^i - q^{k+i}}{1 - q^{k+i}}\right),$$

where the last inequality follows since $1 + x \le e^x$ for all real x. This gives the following, where in the last step we apply the expression for the sum of a geometric series:

$$\mathbb{P}(\mathcal{F}_k) \leqslant \exp\left(-\sum_{i=1}^{n-k} \frac{q^i - q^{k+i}}{1 - q^{k+i}}\right)$$

$$\leqslant \exp\left(-\frac{1 - q^k}{1 - q^n} \sum_{i=1}^{n-k} q^i\right) = \exp\left(-\frac{q(1 - q^k)(1 - q^{n-k})}{(1 - q)(1 - q^n)}\right).$$

Observe that $1 - q^n = (1 + q^{n/2})(1 - q^{n/2})$ and $\max\{k, n - k\} \ge n/2$, thus

$$\mathbb{P}(\mathcal{F}_k) \leqslant \exp\left(-\frac{q}{1-q} \cdot \frac{(1-q^k)(1-q^{n-k})}{(1+q^{n/2})(1-q^{n/2})}\right) \leqslant \exp\left(-\frac{q}{1-q} \cdot \frac{1-q^{\min\{n-k,k\}}}{1+q^{n/2}}\right),$$
and finally $\mathbb{P}(\mathcal{F}_k) \leqslant \exp\left(-\frac{q(1-q^{\min\{n-k,k\}})}{2(1-q)}\right)$ since $1+q^{n/2} \leqslant 2$.

5.3 Bounds on the Number of Cuts

For $1/2 < \alpha < 1$ we let the random variable $X_n(\alpha)$ denote the number of vertices $k \in [\lceil (1-\alpha)n \rceil, \lfloor \alpha n \rfloor]$ which are cut vertices in $\mathcal{P}(n,q)$. We have the following by Lemma 17.

Corollary 22. Let $n \ge 1$ be an integer, $1/2 < \alpha < 1$, and $0 \le q \le 1$. Then

$$X_n(\alpha) = \sum_{k=\lceil (1-\alpha)n\rceil}^{\lfloor \alpha n\rfloor} \left(\mathbf{1}_{\mathcal{C}_k^{\mathcal{F}}} + \mathbf{1}_{\mathcal{C}_k^{\mathcal{R}}} \right).$$

Restricting the ranges of q and k in Lemma 20 gives the following result.

Lemma 23. Let $1/2 < \alpha < 1$, $n \geqslant (\frac{100}{1-\alpha})^5$ be an integer, and $0 < q \leqslant 1 - n^{4/5}$. Then, for any integer $k \in [\lceil (1-\alpha)n \rceil, \lfloor \alpha n \rfloor]$, we have

$$e^{-1/6} \cdot (1-q)^{5/2} \cdot \exp\left(-\frac{\pi^2}{6(1-q)}\right) \leqslant \mathbb{P}\left(C_k^{\mathcal{F}}\right) \leqslant e^6 \cdot (1-q)^{5/2} \cdot \exp\left(-\frac{\pi^2}{6(1-q)}\right).$$

Proof. Recall $\mathbb{P}(v_k = 1) = \frac{1-q}{1-q^k} \geqslant 1-q$ from (6). Since $k \in [\lceil (1-\alpha)n \rceil, \lfloor \alpha n \rfloor]$, we have

$$q^k \leqslant q^{\min\{k,n-k\}} \leqslant (1 - 1/n^{4/5})^{(1-\alpha)n} \leqslant (e^{-1/n^{4/5}})^{(1-\alpha)n} \leqslant e^{-(1-\alpha)n^{1/5}} \leqslant e^{-100},$$
 (35)

where the last step uses $n \geqslant (\frac{100}{1-\alpha})^5$. By independence $\mathbb{P}(\mathcal{C}_k^{\mathcal{F}}) = \mathbb{P}(\mathcal{F}_k) \cdot \mathbb{P}(v_k = 1)$, thus

$$(1-q)\cdot\mathbb{P}(\mathcal{F}_k)\leqslant\mathbb{P}(\mathcal{C}_k^{\mathcal{F}})=\frac{1-q}{1-q^k}\cdot\mathbb{P}(\mathcal{F}_k)\leqslant\mathrm{e}\cdot(1-q)\cdot\mathbb{P}(\mathcal{F}_k). \tag{36}$$

By (36) it suffices to bound $\mathbb{P}(\mathcal{F}_k)$. Taking x = q - 1 > -1, the bounds on log from (2) give

$$\frac{1}{\log q} \geqslant -\frac{1}{1-q}, \quad \log q \geqslant -\frac{1-q}{q}, \quad \text{and} \quad \frac{1}{\log q} \leqslant 1 - \frac{1}{1-q}.$$
 (37)

Thus, applying (37) to the bound on $\mathbb{P}(\mathcal{F}_k)$ in Lemma 20 gives

$$\log \mathbb{P}(\mathcal{F}_k) \geqslant \frac{\pi^2}{6\log q} + \frac{3\log(1-q)}{2} + \frac{q\log q}{6(1-q)} \geqslant -\frac{\pi^2}{6(1-q)} + \frac{3\log(1-q)}{2} - \frac{1}{6}.$$
 (38)

Inserting (38) into (36) gives the lower bound in the statement.

For the upper bound applying (37) and (35) to Lemma 20 gives

$$\log \mathbb{P}(\mathcal{F}_{k}) + \frac{\pi^{2}}{6(1-q)} - \frac{3\log(1-q)}{2} \leqslant \frac{\pi^{2}}{6} + \frac{2q^{\min\{k,n-k\}}}{(1-q)(1-q^{\min\{k,n-k\}})} - \frac{1-q}{q\log q}$$

$$\leqslant \frac{\pi^{2}}{6} + 3n^{4/5}e^{-n^{1/5}/4} + \frac{1}{q}$$

$$\leqslant 2 + \frac{1}{q}.$$
(39)

Let $f(q) = e^{3+1/q} \cdot (1-q)^{5/2} \cdot e^{-\frac{\pi^2}{6(1-q)}}$ and observe that applying (39) to (36) gives the bound $\mathbb{P}(\mathcal{C}_k^{\mathcal{F}}) \leq f(q)$. Now, as f(q) is monotone decreasing in q, for any $q \leq 1/3$,

$$f(q) \ge f(1/3) = e^6 e^{-\frac{\pi^2}{4}} (2/3)^{5/2} > 1.$$

Thus since $\mathbb{P}(\mathcal{C}_k^{\mathcal{F}}) \leq \max\{f(q), 1\}$ holds and $e^{3+1/q}$ is monotone decreasing in q we can simplify the bound to $\mathbb{P}(\mathcal{C}_k^{\mathcal{F}}) \leq e^6 \cdot (1-q)^{5/2} \cdot e^{-\frac{\pi^2}{6(1-q)}}$, as claimed.

The next lemma shows $\mathcal{C}_k^{\mathcal{R}}$ is 'rarer' than $\mathcal{C}_k^{\mathcal{F}}$, except if q=1 where they are equiprobable.

Lemma 24. For integers $1 \le k \le n$, and 0 < q < 1, we have

$$\mathbb{P}(\mathcal{C}_k^{\mathcal{R}}) = q^{k(n-k+1)-1} \mathbb{P}(\mathcal{C}_k^{\mathcal{F}}).$$

Proof. Recall that $\mathbb{P}(v_k = j) = \frac{(1-q)q^{j-1}}{1-q^k}$ by (6), and so $\mathbb{P}(v_k = k) = q^{k-1}\mathbb{P}(v_k = 1)$. The result follows from independence and since $\mathbb{P}(\mathcal{R}_k) = q^{k(n-k)} \cdot \mathbb{P}(\mathcal{F}_k)$ by Lemma 18. \square

Lemma 25 below shows that for a certain range of q there are many cut vertices. This proves one side of Theorem 2 and will also be key to the proof of Theorem 3.

Lemma 25. Let $1/2 < \alpha < 1$, $n \ge \left(\frac{100}{\min\{1-\alpha, 2\alpha-1\}}\right)^5$ be an integer, and $0 \le q \le 1 - n^{-4/5}$. Then, for any x > 0,

$$\mathbb{P}\left(X_n(\alpha) < \frac{2\alpha - 1}{10} \cdot n(1 - q)^{5/2} e^{-\frac{\pi^2}{6(1 - q)}} \left(1 - \frac{1000x}{\sqrt{2\alpha - 1}} \cdot \sqrt{\frac{e^{\frac{\pi^2}{6(1 - q)}} \log n}{n(1 - q)^{7/2}}}\right)\right) \leqslant \frac{5}{x^2} + \frac{1}{n^7}.$$

Proof. To begin, by Corollary 22 and Lemma 23,

$$\sum_{k=\lceil (1-\alpha)n \rceil}^{\lfloor \alpha n \rfloor} \mathbb{P}(\mathcal{C}_k^{\mathcal{F}}) \geqslant ((2\alpha - 1)n - 2) \cdot e^{-1/6} (1 - q)^{5/2} e^{-\frac{\pi^2}{6(1-q)}}.$$

Since n satisfies $n \ge \left(\frac{100}{\min\{1-\alpha, 2\alpha-1\}}\right)^5$, we have

$$\mathbb{E}[X_n(\alpha)] \geqslant \frac{2\alpha - 1}{10} \cdot n \cdot (1 - q)^{5/2} \cdot e^{-\frac{\pi^2}{6(1 - q)}}.$$
 (40)

We want to show that $X_n(\alpha)$ concentrates around $\mathbb{E}[X_n(\alpha)]$, but we cannot apply Lemma 10 directly as $\mathcal{C}_k^{\mathcal{F}}$ is not ℓ -local for any reasonable ℓ . Recalling $\mathcal{L}_k = \{v_i \leqslant i - k \text{ for all } k < i \leqslant k + b_n\}$ from (16), where $b_n = \left\lceil \frac{8 \log n}{\log(1/q)} \right\rceil$ by (14), define the event $\mathcal{C}_k^{\mathcal{L}} = \mathcal{L}_k \cap \{v_k = 1\}$, and let

$$Y_n(\alpha) = \sum_{k=\lceil (1-\alpha)n \rceil}^{\lfloor \alpha n \rfloor} \mathbf{1}_{\mathcal{C}_k^{\mathcal{L}}}.$$

We note two facts that follow from the first part of the statement in Lemma 11, the bound $\log(1/q) \geqslant 1 - q$, and the assumption $n \geqslant \left(\frac{100}{\min\{1-\alpha, 2\alpha-1\}}\right)^5$. Firstly, $\mathcal{C}_k^{\mathcal{L}}$ is ℓ -local to $k + \lceil \frac{b}{2} \rceil$, where $\ell = \lceil \frac{1+b_n}{2} \rceil \leqslant \frac{5\log n}{1-q}$. Secondly, $\mathbb{P}(\mathcal{C}_k^{\mathcal{F}}) \leqslant \mathbb{P}(\mathcal{C}_k^{\mathcal{L}}) \leqslant (1+n^{-7})\mathbb{P}(\mathcal{C}_k^{\mathcal{F}})$. It follows that $Y(\alpha)$ is the 'localisation' of $X(\alpha)$ according to Lemma 11, where $c_i = 1$ for all $i \in [n]$, also

$$\mathbb{E}\left[X_n(\alpha)\right] = \sum_{k=\lceil (1-\alpha)n\rceil}^{\lfloor \alpha n\rfloor} \mathbb{P}\left(\mathcal{C}_k^{\mathcal{F}}\right) \leqslant \mathbb{E}\left[Y_n(\alpha)\right] \leqslant (1+n^{-7}) \cdot \mathbb{E}\left[X_n(\alpha)\right] \leqslant 2 \cdot \mathbb{E}\left[X_n(\alpha)\right]. \tag{41}$$

Recall S and M from Lemma 10. In this setting $S = \lceil \lceil (1-\alpha)n \rceil, \lfloor \alpha n \rfloor \rceil$ and, by Lemma 23,

$$M = \max_{k \in S} \mathbb{P}\left(\mathcal{C}_k^{\mathcal{L}}\right) \leqslant (1 + n^{-7}) \cdot \max_{k \in S} \mathbb{P}\left(\mathcal{C}_k^{\mathcal{F}}\right) \leqslant 2 \cdot e^6 (1 - q)^{5/2} e^{-\frac{\pi^2}{6(1 - q)}}.$$

Since $10^2 e^6 \left(1 - \frac{2}{(2\alpha - 1)n}\right)^{-1} \leq 250^2$, we have the following by comparison with (40):

$$M \cdot |S| \cdot \ell \leqslant 2e^{6} (1 - q)^{5/2} e^{-\frac{\pi^{2}}{6(1 - q)}} \cdot (2\alpha - 1) \, n \cdot \frac{5 \log n}{1 - q} \leqslant \frac{250^{2} \log n}{1 - q} \cdot \mathbb{E} \left[X_{n}(\alpha) \right]. \tag{42}$$

Finally, for any x > 0, by Lemma 10, (41), and (42)

$$\mathbb{P}\left(Y_{n}(\alpha) < \frac{\mathbb{E}\left[X_{n}(\alpha)\right]}{2} - \sqrt{\frac{(250x)^{2} \log n}{1 - q} \cdot \mathbb{E}\left[X_{n}(\alpha)\right]}\right)
\leq \mathbb{P}\left(Y_{n}(\alpha) < \mathbb{E}\left[Y_{n}(\alpha)\right] - x \cdot \sqrt{M \cdot |S| \cdot \ell}\right)
\leq 5/x^{2}.$$
(43)

Now, by (40), we have

$$\frac{(250x)^2 \cdot \log n}{\mathbb{E}\left[X_n(\alpha)\right] \cdot (1-q)} \leqslant \frac{(250x)^2 \cdot \log n}{\frac{2\alpha-1}{10} \cdot n \cdot (1-q)^{5/2} e^{-\frac{\pi^2}{6(1-q)}} \cdot (1-q)} \leqslant \frac{(1000x)^2}{2\alpha-1} \cdot \frac{e^{\frac{\pi^2}{6(1-q)}} \log n}{n \cdot (1-q)^{7/2}}.$$

The result follows by (43) as $\mathbb{P}(Y_n(\alpha) \neq X_n(\alpha)) \leq n^{-7}$ by Lemma 11.

5.4 Proof of Theorem 2

We now use our bounds on $\mathbb{P}(\mathcal{C}_k^{\mathcal{F}})$ and $X_n(\alpha)$ to establish Theorem 2, which shows that $q = 1 - \frac{\pi^2}{6 \log n}$ is a sharp threshold for having a vertex cut separating the graph into two macroscopic pieces.

Proof of Theorem 2. As we are proving a statement concerning a limit in n we can assume $n \geqslant \left(\frac{100}{\min\{1-\alpha,2\alpha-1\}}\right)^5$. We break the proof into three cases depending on the value of q; in the first two cover the 0-statement, and the last case deals with the 1-statement.

Recall that $X_n(\alpha)$ is the number of cut vertices in $[\lceil (1-\alpha)n \rceil, \lfloor \alpha n \rfloor]$ and thus $\{X_n(\alpha) \ge 1\} = \{\exists \text{ an } (1, \alpha)\text{-separator}\}$. Markov's inequality, Corollary 22, and Lemma 24 give

$$\mathbb{P}(X_n(\alpha) \geqslant 1) \leqslant \mathbb{E}[X_n(\alpha)] \leqslant 2n \cdot \max_{k \in [\lceil (1-\alpha)n \rceil, \lfloor \alpha n \rfloor]} \mathbb{P}(C_k^{\mathcal{F}}) \leqslant 2n \cdot \max_{k \in [\lceil (1-\alpha)n \rceil, \lfloor \alpha n \rfloor]} \mathbb{P}(\mathcal{F}_k).$$

To prove the 0-statement it suffices to show $\mathbb{P}(C_k^{\mathcal{F}})$ or $\mathbb{P}(\mathcal{F}_k)$ is $o(\frac{1}{n})$ whenever

$$k \in [\lceil (1-\alpha)n \rceil, \lfloor \alpha n \rfloor].$$

Case (i) $[(1-q)^{-1} \ge n^{4/5}]$: If q=1 then the result follows from Lemma 16, so we can assume $q \ne 1$. In this case we have $q \ge 1 - n^{-4/5}$ and $\min\{k, n-k\} - 1 \ge 5\sqrt{n}$, so

$$\frac{1 - q^{\min\{k, n-k\}}}{1 - q} = \frac{(1 - q)(1 + q + q^2 + \dots + q^{\min\{k, n-k\} - 1})}{1 - q} \geqslant \sum_{i=0}^{5\sqrt{n}} (1 - n^{-4/5})^i.$$
 (44)

Now, by Bernoulli's inequality, we have

$$\frac{1 - q^{\min\{k, n - k\}}}{1 - q} \geqslant \sum_{i = 0}^{5\sqrt{n}} (1 - i \cdot n^{-4/5}) \geqslant 4\sqrt{n}.$$
(45)

Using this bound in combination with Lemma 21 and the fact $q \ge 1 - n^{-4/5} \ge 1/2$ yields

$$\mathbb{P}(\mathcal{F}_k) \leqslant \exp\left(-\frac{q}{2} \cdot \frac{\left(1 - q^{\min\{k, n - k\}}\right)}{1 - q}\right) \leqslant \exp\left(-\frac{1}{4} \cdot 4\sqrt{n}\right) \leqslant e^{-\sqrt{n}}.$$

Case (ii) $[(1-q)^{-1} < n^{4/5}$, and $\frac{\pi^2}{6(1-q)} - \log n + \frac{5 \log \log n}{2} \to \infty]$: By Lemma 23 we have

$$\mathbb{P}\left(C_k^{\mathcal{F}}\right) \leqslant e^6 \cdot \exp\left(-\frac{\pi^2}{6(1-q)}\right) \cdot (1-q)^{5/2} = \exp\left(6 - \frac{\pi^2}{6(1-q)} + \frac{5\log(1-q)}{2}\right), (46)$$

for $k \in [\lceil (1-\alpha)n \rceil, \lfloor \alpha n \rfloor]$ as $n \geqslant \left(\frac{100}{\min\{1-\alpha, 2\alpha-1\}}\right)^5$. Differentiating the exponent of (46) gives

$$\frac{\mathrm{d}}{\mathrm{d}\,q}\left(6 - \frac{\pi^2}{6(1-q)} + \frac{5\log(1-q)}{2}\right) = -\frac{\pi^2}{6(1-q)^2} - \frac{5}{2(1-q)} = \frac{15q - \pi^2 - 15}{6(1-q)^2}.\tag{47}$$

From (47) we see that the exponent of the bound on $\mathbb{P}(\mathcal{C}_k^{\mathcal{F}})$ from (46) is monotone decreasing in q provided $q \in [0,1)$. Hence to bound $\mathbb{P}(\mathcal{C}_k^{\mathcal{F}})$ it suffices to evaluate (46) for the smallest q in the scope of this case. Thus, if we let $w(n) = \frac{\pi^2}{6(1-q)} - \log n + \frac{5\log\log n}{2}$ then it suffices to bound $\mathbb{P}(\mathcal{E}_k)$ in the case $w(n) \to \infty$ arbitrary slowly and we will assume, for convenience, that $w(n) \leq \sqrt{\log\log n}$. As $q \geq 1-2/\log n$ we have $\log(1-q) \leq 2-\log\log n$. Thus by (46)

$$\mathbb{P}\left(\mathcal{C}_{k}^{\mathcal{F}}\right) \leqslant \exp\left(6 - \frac{\pi^{2}}{6(1-q)} + \frac{5\log(1-q)}{2}\right)$$
$$\leqslant \exp\left(7 - \log n - w(n)\right) = \frac{e^{7} \cdot e^{-w(n)}}{n} = o\left(\frac{1}{n}\right).$$

Case (iii) $\left[q \geqslant 0, \text{ and } \frac{\pi^2}{6(1-q)} - \log n + \frac{9\log\log n}{2} \to -\infty\right]$: In this final case we will show that $\mathbb{P}(X_n(\alpha) \geqslant 1) = 1 - o(1)$ by applying Lemma 25 which bounds $X_n(\alpha)$ from below. If we let $w(n) = -\frac{\pi^2}{6(1-q)} + \log n - \frac{9\log\log n}{2}$ then for this case we have $w(n) \to \infty$ and

$$\exp\left(-\frac{\pi^2}{6(1-q)}\right) = \exp\left(-\log n + \frac{9\log\log n}{2} + w(n)\right) = \frac{e^{w(n)} \cdot \log^{9/2} n}{n}.$$
 (48)

For this case the (loose) bound $q \leqslant 1 - 1/\log n$ holds, thus $(1 - q)^{5/2} \geqslant 1/\log^{5/2} n$, hence

$$n \cdot (1-q)^{5/2} \cdot e^{-\frac{\pi^2}{6(1-q)}} \geqslant n \cdot \frac{1}{\log^{5/2} n} \cdot \frac{e^{w(n)} \cdot \log^{9/2} n}{n} = e^{w(n)} \log^2 n, \tag{49}$$

by (48). The bound $q \le 1 - 1/\log n$ also implies $1/(1-q)^{7/2} \le \log^{7/2} n$, so (48) gives

$$\frac{\log n}{n(1-q)^{7/2}} \cdot e^{\frac{\pi^2}{6(1-q)}} \leqslant \frac{\log^{9/2} n}{n} \cdot \frac{n}{e^{w(n)} \cdot \log^{9/2} n} = e^{-w(n)}.$$
 (50)

Inserting the bounds (49) and (50) into Lemma 25 and choosing x = w(n) gives

$$\mathbb{P}\left(X_n(\alpha) < \frac{(2\alpha - 1)\log^2 n}{10} \cdot e^{w(n)} \left(1 - \frac{1000}{\sqrt{2\alpha - 1}} \cdot w(n) \cdot e^{-w(n)/2}\right)\right) \leqslant \frac{5}{w(n)^2} + \frac{1}{n^7}.$$

The result follows since $\alpha > 1/2$ is fixed and $w(n) \to \infty$.

Remark 26. The events \mathcal{F}_k are positively correlated, i.e. $\mathbb{P}(\mathcal{F}_k \cap \mathcal{F}_{k+j}) > \mathbb{P}(\mathcal{F}_k) \mathbb{P}(\mathcal{F}_{k+j})$. This inequality also holds for \mathcal{F} replaced by \mathcal{L} for j not too large. This is an obstruction to proving a (significantly) improved lower threshold in Case (iii) using the second moment method.

5.5 Proof of Theorem 3

We now prove Theorem 3, which shows the diameter is linear when q is bounded away from 1. Theorem 3 follows from the next lemma and an earlier bound on the number of cut vertices.

Lemma 27. Let G be an n-vertex graph containing a Hamiltonian path. Then if there exists a set $C \subseteq V$ such that each vertex in C is a cut vertex of G, then $diam(G) \ge |C|+1$.

Proof. Label the vertices from 1 to n along the Hamiltonian path P. Denote the cut vertices by $c_1 < \cdots < c_k$, where k = |C|, ordered with respect to the vertex labelling of P.

Claim 28. Any path from vertex 1 to vertex n must include every vertex of C.

Proof of Claim. Since the graph contains an n-vertex path as a subgraph (i.e. a Hamiltonian path), we must have that for $1 \le i \le |C|$ the removal of the cut vertex c_i separates the graph into exactly two connected graphs, one containing the subpath $[1, \ldots, c_i - 1]$ and the other containing the subpath $[c_i + 1, \ldots, n]$, with no edges between them. Now assume, for a contradiction, that there exists a path Q from 1 to n in G and a vertex $c_i \in C$ which is not contained in Q. Since the path does not go through c_i it must use an edge xy where $x \in [1, \ldots, c_i - 1]$ and $y \in [c_i + 1, n]$ however this contradicts our earlier observation about the cut vertex c_i .

By the claim, the shortest path connecting 1 to n has length at least |C| + 1. This follows as 1 and n cannot be in C, since removing them does not disconnect the graph due to the existence of the path P on [n] as a subgraph.

We note that one cannot completely remove the assumption of a subpath of length n in Lemma 27. To see this consider the binary tree. We are now ready to prove Theorem 3.

Proof of Theorem 3. Let $\alpha = 2/3$ and recall that $X_n(\alpha)$ denotes the number of cut vertices in $[\lceil (1-\alpha)n \rceil, \lfloor \alpha n \rfloor] = [\lceil n/3 \rceil, \lfloor 2n/3 \rfloor]$. Since $0 < q \leqslant 1 - \varepsilon$ we have $\mathrm{e}^{\frac{\pi^2}{6(1-q)}} \leqslant \mathrm{e}^{\frac{\pi^2}{6\varepsilon}}$. Assume that $n \geqslant 300^5 \cdot \mathrm{e}^{\frac{\pi^2}{6\varepsilon}} \geqslant \left(\frac{100}{\max\{1-\alpha,2\alpha-1\}}\right)^5$. We will now apply Lemma 25 to bound $X_n(\alpha)$. First observe that, by the restrictions on n above and $0 < q \leqslant 1 - \varepsilon$,

$$\frac{200}{\sqrt{2 \cdot (2/3) - 1}} \sqrt{\frac{e^{\frac{\pi^2}{6(1-q)}} \log n}{n(1-q)^{7/2}}} \leqslant n^{-1/3},$$

and

$$\frac{2 \cdot (2/3) - 1}{10} \cdot n(1 - q)^{5/2} e^{-\frac{\pi^2}{6(1 - q)}} \geqslant \frac{\varepsilon^{5/2} e^{-\frac{\pi^2}{6\varepsilon}}}{30} n.$$

Thus, is we set $c := c(\varepsilon) = \varepsilon^{5/2} e^{-\frac{\pi^2}{6\varepsilon}} \cdot 300^{-5}$, then our earlier assumption on n is implied by $n \ge 1/c$. Furthermore, if we set $x = n^{1/6}$ and $\alpha = 2/3$ in Lemma 25, then this yields $\mathbb{P}(X_n(\alpha) < cn) \le 3n^{-1/6} \le n^{-1/10}$. The result follows as there are at least $X_n(\alpha)$ cut vertices and $\{\operatorname{diam}(\mathcal{P}(n,q)) \le d+1\} \subseteq \{X_n(\alpha) \le d\}$ holds for any integer $d \ge 0$ by Lemma 27.

6 The Treewidth and Cutwidth, Proof of Theorem 4

The aim of this section is to prove Theorem 4, which gives lower and upper bounds on the treewith and cutwidth in the tangled path, respectively. This is an amalgamation of several bounds proved in this section. In particular, the lower bounds on treewidth follow from Lemmas 33, 34, and the bound $\mathsf{tw}(G) \geqslant \lfloor \varphi(G) \cdot n/4 \rfloor$ from (5) in combination with Theorem 1. Then the upper bounds on cutwidth follow from Lemma 35, and as $\mathsf{cw}(G) \leqslant |E(G)| \leqslant 2n-2$ by (5).

6.1 Lower Bound on Treewidth

The main results in this section are Lemmas 33 and 34, which give two separate bounds on treewidth. The first bound is tight, but it only holds for small values of q. Whereas the second holds for any $q \leq 1 - 1/n$, however it is only known to be tight up to a log factor. Both bounds follow from finding a minor with good expansion properties, however the probability of finding such a minor is calculated slightly differently in the two different cases.

Before we begin in earnest, we must introduce consecutive patterns in permutations. Given a list of k distinct integers $w = w_1, \ldots, w_k$, the *standardization* of w, written $\mathsf{st}(w)$, is the unique permutation of [k] that is order-isomorphic to w. That is, we obtain $\mathsf{st}(w)$ by replacing the smallest element among $\{w_1, \ldots, w_k\}$ with 1, the second smallest with 2, and so on. We say that $\pi \in S_n$ contains $\sigma \in S_k$ consecutively if there exists an index $i \in [n-k+1]$ such that $\mathsf{st}(\pi(i), \ldots, \pi(i+k-1)) = \sigma$; otherwise, we say that π avoids σ consecutively.

We now give a lemma relating consecutive patterns in permutations to minors in the related tangled paths. Figure 5 illustrates this Lemma.

Lemma 29. Let $1 \le k \le n$ be integers and let $\pi \in S_n$ contain $\sigma \in S_k$ consecutively. Then, $\mathsf{layer}(\pi(P_n), P_n)$ contains $\mathsf{layer}(\sigma(P_k), P_k)$ as a minor.

Proof. Let $\pi(i), \ldots, \pi(i+k-1)$ be such that $\operatorname{st}(\pi(i), \ldots, \pi(i+k-1)) = \sigma$ and $s_1 < \cdots < s_k$ be the elements of $\{\pi(i), \ldots, \pi(i+k-1)\}$ ordered increasingly. Define P' to be the path of length $s_k - s_1 + 1$ on $\{s_1, s_1 + 1, \ldots, s_k\}$. Finally, let the graph H be the union of P' and the set of edges $\{\pi(i+j)\pi(i+j+1)\}_{j=0}^{k-2}$. Observe that H is isomorphic to a copy of $\operatorname{layer}(\sigma(P_k), P_k)$ where the edges of P_k have been subdivided. Now, as $H \subseteq \operatorname{layer}(\pi(P_n), P_n)$ the result follows.

We now recall two results from [5], the first is a restricted form of independence for patterns in a q-Mallows permutation.

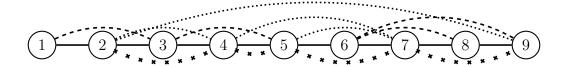


Figure 5: An example of Lemma 29. Let $\sigma = (3, 4, 2, 1, 5)$ and $\pi = (1, 3, 5, 7, 4, 2, 9, 6, 8)$, thus $st(5, 7, 4, 2, 9) = \sigma$. The graph induced by the crossed and dotted edges is isomorphic to $layer(\sigma(P_5), P_5)$. The graph induced by the dashed, solid and dotted edges is $layer(\pi(P_9), P_9)$. The crossed edges can be subdivided to give a path on $\{2, \ldots, 9\}$.

Lemma 30 ([5, Lemma 2.5]). Let (i_1, \ldots, i_k) and (i'_1, \ldots, i'_ℓ) be two increasing sequences with $i_k < i'_1$. Let $\pi \sim \mu_{n,q}$ for $n \ge i'_\ell$. Then, $\operatorname{st}(\pi(i_1), \ldots, \pi(i_k))$ and $\operatorname{st}(\pi(i'_1), \ldots, \pi(i'_k))$ are independent.

The second follows by translation invariance of Mallows permutations [5, Lemma 2.6].

Lemma 31 ([5, Corollary 2.7]). Let $(i, i+1, ..., i+k-1) \subseteq [n]$ be a sequence of consecutive elements. If $\pi \sim \mu_{n,q}$ then $\operatorname{st}(\pi(i), ..., \pi(i+k-1)) \sim \mu_{k,q}$.

Given $\sigma \in S_k$, we let $\mathsf{Forb}_n(\sigma) \subseteq S_n$ denote the set of permutations in S_n that avoid σ consecutively. Similarly, given $S \subseteq S_k$ we let $\mathsf{Forb}_n(S) = \bigcap_{\sigma \in S} \mathsf{Forb}_n(\sigma)$. Recall the (n,q)-Mallows measure $\mu_{n,q}(\sigma)$ of $\sigma \in S_n$ from (1). Thus $\mu_{n,q}(S) = \sum_{\sigma \in S} \mu_{n,q}(\sigma)$ for any $S \subseteq S_n$.

Lemma 32. Let $1 \leqslant k \leqslant n$ be integers, and $0 \leqslant q \leqslant 1$. Then, for any $S \subseteq S_k$ we have

$$\mu_{n,q}\left(\mathsf{Forb}_n(S)\right) \leqslant \mu_{k,q}\left(\mathsf{Forb}_k(S)\right)^{\lfloor n/k \rfloor}.$$

Proof. To begin, observe that if a permutation $\pi \in S_n$ avoids a set of permutations $S \subseteq S_k$ consecutively then $\mathsf{st}(\pi(ik+1), \ldots, \pi(ik+k)) \notin S$ for any $0 \leqslant i \leqslant \lfloor n/k \rfloor - 1$. The permutations $(\mathsf{st}(\pi(ik+1), \ldots, \pi(ik+k)))_{i=0}^{\lfloor n/k \rfloor - 1}$ are independent by Lemma 30 and by Lemma 31 each permutation $\mathsf{st}(\pi(ik+1), \ldots, \pi(ik+k))$ has distribution $\mu_{k,q}$. Thus, for any $S \subseteq S_k$, we have

$$\mu_{n,q}\left(\mathsf{Forb}_n(S)\right) \leqslant \prod_{i=0}^{\lfloor n/k\rfloor-1} \mathbb{P}(\operatorname{st}(\pi(ik+1),\ldots,\pi(ik+k)) \notin S) = \mu_{k,q}\left(\mathsf{Forb}_k(S)\right)^{\lfloor n/k\rfloor},$$

as claimed. \Box

We now have what we need to prove our fist lower bound on treewidth.

Lemma 33. Let $n \ge 1$ be an integer, $\kappa > 0$ be any constant, and $0 < q \le 1 - \kappa(\log \log n)^2/\log n$. Then, there exists a constant c > 0 such that for any n > 1/c we have

$$\mathbb{P}\bigg(\operatorname{tw}(\mathcal{P}(n,q)) < c\left(\sqrt{\frac{\log n}{\log(1/q)}} + 1\right)\bigg) \leqslant \exp(-\sqrt{n}).$$

Proof. By (4) and Lemma 16, for any k, the bound $\operatorname{tw}(\mathcal{P}(k,1)) \geqslant k/200$ holds with probability 1 - f(k) where $f(k) \to 0$ as $k \to \infty$. Thus, there exist fixed constants C > 0 such that for any $k \geqslant C$ there is at least one permutation $\sigma_k \in S_k$ satisfying $\operatorname{tw}(\operatorname{layer}(\sigma_k(P_k), P_k)) \geqslant k/200$. We wish to show that, for some k := k(n, q) defined later, a permutation $\pi \sim \mu_{n,q}$ contains a given $\sigma_k \sim \mu_{k,q}$ as a consecutive pattern. We will do this by appealing to Lemmas 30 and 31. The result will then follow from Lemmas 32 and 5. To begin, by (1) we have

$$\mu_{k,q}(\sigma_k) = q^{\text{Inv}(\sigma_k)} \cdot (Z_{k,q})^{-1} = q^{\binom{k}{2}} \cdot \prod_{i=1}^k \frac{(1-q)}{1-q^i} \geqslant q^{k^2} \cdot (1-q)^k.$$
 (51)

Recall that $0 < q \le 1 - \kappa (\log \log n)^2 / \log n$, for some $\kappa > 0$ and let

$$k = \sqrt{\frac{\alpha \log n}{\log(1/q)}}$$
 where $\alpha = \min\left\{\frac{1}{100}, \frac{\kappa}{25}\right\}.$ (52)

As $\log q \leqslant q - 1$ by (2) we have $1/\log(1/q) \leqslant 1/(1-q) \leqslant \log(n)/(\kappa(\log\log n)^2)$. Thus,

$$k \leqslant \sqrt{\frac{\alpha \log^2 n}{\kappa (\log \log n)^2}} = \sqrt{\frac{\alpha}{\kappa}} \cdot \frac{\log n}{\log \log n} \leqslant \frac{\log n}{5 \log \log n}.$$
 (53)

Returning to the bound on $\mu_{k,q}(\sigma_k)$, by (51) and (53), for any large enough n we have

$$\mu_{k,q}(\sigma_k) \geqslant q^{k^2} \cdot (1-q)^k$$

$$\geqslant e^{k^2 \log q} \cdot \left(\frac{\kappa (\log \log n)^2}{\log n}\right)^k$$

$$\geqslant e^{-\alpha \log n} \cdot \left(e^{-\log \log n}\right)^{\sqrt{\alpha/\kappa} \cdot \frac{\log n}{\log \log n}}.$$
(54)

Now by our choice of α in (52) we have

$$\alpha + \sqrt{\alpha/\kappa} \leqslant 1/100 + \sqrt{\kappa/(25\kappa)} < 1/4.$$

Thus by (54):

$$\mu_{k,q}\left(\mathsf{Forb}_{k}(\sigma_{k})\right) = 1 - \mu_{k,q}(\sigma_{k}) \leqslant 1 - n^{-\left(\alpha + \sqrt{\alpha/\kappa}\right)} \leqslant 1 - n^{-1/4}.$$
 (55)

Observe that, for suitably large n, $\lfloor n/k \rfloor \geqslant n^{3/4}$ by (53). Thus, by Lemma 32 and (55),

$$\mu_{n,q}\left(\mathsf{Forb}_n(\sigma_k)\right) \leqslant \mu_{k,q}\left(\mathsf{Forb}_k(\sigma_k)\right)^{\lfloor n/k \rfloor} \leqslant \left(1 - n^{-1/4}\right)^{n^{3/4}} \leqslant e^{-\sqrt{n}}.$$
 (56)

It follows from (52) that there is a constant C' such that if $n \ge C'$, then $k \ge C$. Thus if we take $c \le \min\{\sqrt{\alpha}/200, 1/C'\}$ so that $n \ge 1/c$ is sufficiently large, then

$$\mathbb{P}\Bigg(\operatorname{tw}\left(\mathcal{P}(n,q)\right) < c \cdot \sqrt{\frac{\log n}{\log(1/q)}}\,\Bigg) \leqslant \mu_{n,q}(\operatorname{Forb}_n(\sigma_k)) \leqslant \mathrm{e}^{-\sqrt{n}},$$

by Lemmas 32 and 5, (56), and the expression for k given by (58).

The proof of the following lower bound is similar in its use of Lemma 29, but to calculate the probability it uses Lemma 14 to relate the q-Mallows measure to the 1-Mallows measure.

Lemma 34. Let $n \ge 1$ be an integer, and $q \le 1 - \frac{1}{n}$ satisfy $\lim_{n \to \infty} q = 1$. Then, for sufficiently large n,

$$\mathbb{P}(\mathsf{tw}(\mathcal{P}(n,q)) < (1-q)^{-1}/10^5) \leqslant \exp(-n/100)$$
.

Proof. For an integer $k \ge 1$, we define the set

$$T_k = \{ \sigma \in S_k : \mathsf{tw}(\mathsf{layer}(\sigma(P_k), P_k)) \geqslant k/50 \} \subseteq S_k.$$

By Lemma 13 $\mathbb{P}\left(\varphi(\mathcal{P}(n,1)) \leq \frac{1}{40}\right) \leq 1000 \cdot k^{7/2} \cdot \left(\frac{3}{4}\right)^k$ holds for any $k \geq 100$. By (5) for any k-vertex graph G we have $\mathsf{tw}(G) \geq \lfloor \varphi(G) \cdot k \rfloor - 1$. Thus, for $k \geq 100$, we have

$$\mu_{k,1}(\mathsf{Forb}_k(T_k)) \leqslant 1000 \cdot k^{7/2} \cdot \left(\frac{3}{4}\right)^k \leqslant e^{-k/20}.$$
 (57)

We now fix

$$k = \left\lfloor \frac{1}{1000} \cdot \frac{1}{1 - q} \right\rfloor. \tag{58}$$

Since $\lim_{n\to\infty} q = 1$, we can assume that k satisfies $k \ge 100$ by taking a suitably large n. By Lemma 14, (57) and (58), we have

$$\mu_{k,q}(\mathsf{Forb}_k(T_k)) \leqslant e^{9k^2(1-q)} \cdot \mu_{k,q}(\mathsf{Forb}_k(T_k)) \leqslant e^{\frac{9}{1000} \cdot k} \cdot e^{-k/20} \leqslant e^{-4k/100}.$$

Note that $k \leq \lfloor n/10 \rfloor$ by (58) and $q \leq 1 - \frac{1}{n}$. Thus, by Lemma 32, for sufficiently large n.

$$\mu_{n,q}\left(\mathsf{Forb}_n(T_k)\right) \leqslant \mu_{k,q}\left(\mathsf{Forb}_k(T_k)\right)^{\lfloor n/k \rfloor} \leqslant e^{-(4k/100) \cdot \lfloor n/k \rfloor} \leqslant e^{-n/100}. \tag{59}$$

Recall that $\mu_{n,q}(\mathsf{Forb}_k(T_k))$ is the probability $\pi \sim \mu_{n,q}$ avoids all permutations $\sigma \in T_k$ consecutively. Thus, for any $\pi \in S_n$ such that $\pi \notin \mathsf{Forb}_n(T_k)$, Lemmas 5 and 29 yield

$$\mathsf{tw}\left(\mathsf{layer}(\pi(P_n), P_n)\right) \geqslant \min_{\sigma \in T_k} \mathsf{tw}\left(\mathsf{layer}(\sigma(P_k), P_k)\right) \geqslant k/50.$$

Recall also that $\mathsf{layer}(\pi(P_n), P_n) \sim \mathcal{P}(n, q)$ as $\pi \sim \mu_{n,q}$. Thus by (59), for large n,

$$\mathbb{P}(\mathsf{\,tw\,}(\mathcal{P}(n,q)) < k/50\,) \leqslant \mu_{n,q}(\mathsf{Forb}_n(T_k)) \leqslant \mathrm{e}^{-n/100}.$$

The result now follows from the expression for k given by (58).

6.2 Upper Bound on Cutwidth

Recall the definition of the cutwidth of a graph G given by (4):

$$\operatorname{cw}(G) = \min_{f:V \to \mathbb{Z}, \text{ injective }} \max_{x \in \mathbb{R}} |\{ij \in E(G) : f(i) \leqslant x < f(j)\}|. \tag{60}$$

In this section we prove Lemma 35, which gives the upper bounds on cutwidth in Theorem 4.

Lemma 35. Let $n \ge 1$ be an integer, c > 0 be any constant, and $0 \le q \le 1 - c(\log \log n)^2/\log n$. Then there exists a constant C such that

$$\mathbb{P}\bigg(\operatorname{cw}(\mathcal{P}(n,q)) > C\left(\sqrt{\frac{\log n}{\log(1/q)}} + 1\right)\bigg) = \mathcal{O}\bigg(\frac{1}{n^3}\bigg)\,.$$

Additionally, if $q \ge 1 - (\log \log n)^2 / \log n$, then

$$\mathbb{P}\bigg(\operatorname{cw}(\mathcal{P}(n,q)) > \frac{5}{1-q}\log\left(\frac{1}{1-q}\right)\bigg) = \mathcal{O}\bigg(\frac{1}{n^3}\bigg)\,.$$

We will begin by outlining a sketch of the proof of Lemma 35.

Proof Sketch: For an upper bound on $\operatorname{cw}(\mathcal{P}(n,q))$ we fix $f:[n] \to [n]$ in (60) to be the identity map denoted Id. That is, we order the vertices of $\mathcal{P}(n,q)$ using the ordering of the un-permuted path P_n with edges (i)(i+1) for $i \in [n-1]$. We bound the number of edges crossing any vertex i by bounding the number of elements with values j > i are inserted next to elements k < i by the q-Mallows process. To do this we show that, for $b_n = \lceil \frac{8 \log n}{\log(1/q)} \rceil$ given by (14) and some suitable ℓ , L where $L \geqslant \ell$, the following events hold with high probability:

- (i) no insert position v_i has value greater than b_n ,
- (ii) after L steps the leftmost b_n places contain only elements added at most L steps ago,
- (iii) within any window of L steps there are at most ℓ values of v_i greater than ℓ .

The events (i) and (iii) control the number of long edges created from new entries being added far away from the left-hand end of the process. The event (ii) is more subtle, it ensures that the left-hand end of the permutation grown by the q-Mallows process cannot retain entries that were inserted long ago, again preventing long edges caused by new elements lying next to old ones. We show that if (i) - (iii) hold, then the number of edges crossing any vertex is $\mathcal{O}(\ell)$.

Having concluded the proof sketch we now formalise event described in (ii). To do this we introduce the sparse flush event, which is a relaxation of the local flush event \mathcal{L}_k given by (16). For positive integers n, k, b where $1 \leq k \leq n$ and $b \leq n - k$ and real $L \leq n - k$ we say there is a sparse flush $\mathcal{S}(k, b, L)$ of b items at step k with length L if the following event holds:

$$S(k, b, L) = \{ \text{there exist } k \leqslant t_1 < \dots < t_b \leqslant k + L \text{ such that } v_{t_i} \leqslant i \text{ for all } 1 \leqslant i \leqslant b \}.$$
(61)

To give an intuition on the sparse flush, first recall the local flush event (16) given by $\mathcal{L}_k = \{\text{for each } k < i \leq k + b_n(q) \text{ we have } v_i \leq i - k\}$, where $b_n = \lceil 8\log(n)/\log(1/q) \rceil$. Recall also that no insert position is greater than b_n w.h.p. by (15). Thus, if the local flush event \mathcal{L}_k holds for some vertex k then w.h.p., no element $j \geq k + b_n$ is inserted next to any element $i \leq k$. If we choose $b_n = \lceil 8\log(n)/\log(1/q) \rceil$ in (61) then, conditional on $\mathcal{S}(k, b_n, L)$, the values $(v_{t_i})_{i=1}^{b_n}$ form a (non-consecutive) local-flush, that is $v_{t_1} \leq 1, v_{t_2} \leq 2, \ldots, v_{t_{b_n}} \leq b_n$. This sequence of insert positions ensures that w.h.p. no element $j \geq k + L$ is inserted next to any element $i \leq k$, and thus no such edge ij is present in the permuted

path. If q is very close to 1, the local flush event only occurs with small probability; the next two lemmas show that if we take L significantly larger than b_n , then the sparse flush holds w.h.p., even for q tending to 1 quite fast.

Lemma 36. Let $n, b \ge 1$ be integers, 0 < q < 1, and $\lambda \ge 1$. Let

$$L = \lambda \left(b + \frac{q}{1 - q} + \frac{1}{\log q} \log \frac{1 - q}{1 - q^b} \right),$$

and $1 \leq k \leq n-L$ be an integer. Then,

$$\mathbb{P}(\mathcal{S}(k,b,L)^c) \leqslant \left(\frac{(1-q)q^b}{1-q^b}\right)^{\lambda-1-\log\lambda}.$$

Proof. Recall that each element x in the q-Mallows process (r_n) is inserted at step x and relative position $v_x \sim \nu_{x,q}$. The sparse flush $\mathcal{S}(k,b,L)$ consists of a sequence of b steps $(t_i)_{i=1}^b$ where element t_i is inserted at relative position at most i. We call these steps good. Conditional on t_{i-1} , the $(i-1)^{\text{th}}$ good step, we define the random variable $\tau_i = t_i - t_{i-1}$ to be the additional number of steps we must wait for the i^{th} good step (where $1 \leq i \leq b$ and $t_0 = k$).

Let $T = \sum_{i=1}^{b} \tau_i$ be the total number of steps we have to wait to have b good steps and observe that $S(k, b, L) = \{T \leq L\}$. We now aim to bound T. Since $(v_x)_{x \geq 0}$ are independent and $\mathbb{P}(v_k \leq i) \geq 1 - q^i$ by (10), it follows that $\tau_i \leq X_i$, for independent $X_i \sim \text{Geo}(p_i)$, where $p_i = 1 - q^i$. That is, the time between the $i - 1^{\text{th}}$ and i^{th} good step is stochastically dominated by a geometric random variable with success probability $1 - q^i$. We now set

$$X = \sum_{i=1}^{b} X_i$$
 and $\mu = \mathbb{E}[X] = \sum_{i=1}^{b} \frac{1}{p_i}$.

Observe that $T \leq X$ and so $\mathbb{P}(\mathcal{S}(k, b, L)^c) \leq \mathbb{P}(X > L)$. We bound the latter probability using Lemma 6. To do this we need bounds on $\mu = \mathbb{E}[X]$. For the upper bound:

$$\mu = \sum_{i=1}^{b} \frac{1}{1 - q^i} \leqslant \frac{1}{1 - q} + \int_1^b \frac{1}{1 - q^x} \, \mathrm{d}x = \frac{1}{1 - q} + \int_1^b \left(1 + \frac{q^x}{1 - q^x} \right) \, \mathrm{d}x,$$

since $1/(1-q^x) = 1 + q^x/(1-q^x)$ is decreasing in x. Thus, by integrating we have

$$\mu \leqslant \frac{1}{1-q} + b - 1 - \left\lceil \frac{\log(1-q^x)}{\log q} \right\rceil_{x=1}^b = b + \frac{q}{1-q} + \log\left(\frac{1-q}{1-q^b}\right) / \log(q). \tag{62}$$

Similarly, we can obtain the following lower bound:

$$\mu = \sum_{i=1}^{b} \frac{1}{1 - q^i} \geqslant \int_1^b \frac{1}{1 - q^x} \, \mathrm{d}x \geqslant b + \log\left(\frac{1 - q}{1 - q^b}\right) / \log(q). \tag{63}$$

We apply Lemma 6 to X where $p_* = \min_i p_i \ge 1 - q$, $\mu = \mathbb{E}[X]$ and $\lambda > 1$:

$$\mathbb{P}(\mathcal{S}(k,b,\lambda\mu)^c) \leqslant \mathbb{P}(X \geqslant \lambda\mu) \leqslant \lambda^{-1}(1-p_*)^{\mu(\lambda-1-\log\lambda)} \leqslant q^{\mu(\lambda-1-\log\lambda)}.$$

We now insert the lower bound on μ from (63) into the bound in the line above to give

$$\mathbb{P}(\mathcal{S}(k,b,\lambda\mu)^c) \leqslant q^{\left(b + \log\left(\frac{1-q}{1-q^b}\right)/\log(q)\right) \cdot (\lambda - 1 - \log \lambda)} = \left(\frac{(1-q)q^b}{1-q^b}\right)^{\lambda - 1 - \log \lambda}.$$

To conclude we observe that $L \geqslant \lambda \mu$ by (62).

We now plug some specific values into Lemma 36 for use in the proof of Lemma 35.

Lemma 37. Let $n \ge 1$ be an integer, 0 < q < 1, and $b_n = \left\lceil \frac{8 \log n}{\log(1/q)} \right\rceil \ge 1$. Then for any $L \ge \frac{100}{1-q} \left(\frac{1}{1-q} + \log n \right)$ and integer $1 \le k \le n - L$ we have $\mathbb{P}(\mathcal{S}(k, b_n, L)^c) = o(n^{-10})$.

Proof. Observe that $q^{b_n} \leqslant n^{-8}$ and $\mathcal{S}(k, b_n, L) \supseteq \mathcal{S}(k, b_n, D)$ for any $L \geqslant D$. Now let

$$D = 10 \left(b_n + \frac{q}{1-q} + \log \left(\frac{1-q}{1-q^{b_n}} \right) / \log(q) \right).$$

Applying the inequalities $-1/\log q \leq 1/(1-q)$ and $-\log(1-q) \leq 1/(1-q)$ yields

$$D \leqslant 10 \left(-\frac{8\log n}{\log q} + 1 + \frac{q}{1-q} - \frac{\log(1-q)}{-\log(q)} \right) \leqslant \frac{100}{1-q} \left(\frac{1}{1-q} + \log n \right).$$

Finally,
$$\mathbb{P}(S(k, b_n, D)^c) \leqslant \left(\frac{(1-q)q_n^b}{1-q^{b_n+1}}\right)^{10-1-\log 10} \leqslant \left(\frac{n^{-8}}{1-n^{-8}}\right)^6 = o(n^{-10})$$
 by Lemma 36. \square

We are now ready to prove Lemma 35, which provides upper bounds on the cutwidth.

Proof of Lemma 35. As mentioned at the start of this section we bound the cutwidth from above by fixing the injection $f:[n] \to [n]$ in (60) to be the identity $\mathrm{Id}:[n] \to [n]$.

Before beginning in earnest, we treat the case $0 \le q \le 1/n^2$ as this is a straightforward deduction from an earlier result. Indeed, if $q < 1/n^2$ then by Theorem 9 we have

$$\mathbb{P}(\operatorname{cw}(\mathcal{P}) \geqslant 2) \leqslant \sum_{i \in [n]} \mathbb{P}(|\sigma(i) - i| \geqslant 2) \leqslant n \cdot 2q^2 = \mathcal{O}(n^{-3}). \tag{64}$$

Thus, from now on we can assume $q \ge 1/n^2$. The remainder of the proof is as follows. First, we introduce the event $\mathcal{W}(\ell, L)$, which essentially controls how many long edges originate from any small set of consecutive vertices. We then prove a claim bounding the cutwidth conditional on this event. To apply the claim we then bound the probability that the event fails, breaking into two cases for different values of q. Finally, we conclude by relating the different cases to the lemma's statement. We now begin with the formal definition of the event $\mathcal{W}(\ell, L)$.

Let $b_n = \lceil 8\log(n)/\log(1/q) \rceil$ and define the event $\mathcal{E} = \bigcap_{k \in [n]} \{v_k \leq b_n\}$. For integers $i \in [n]$ and $L \geq \ell \geq 1$, we recall the sparse flush event $\mathcal{S}(i, b_n, L)$ from (61), and also define the event

 $\mathcal{B}(i,\ell,L) = \{ \text{there are at most } \ell \text{ values } k \in [i,\max\{i+L,n\}] \text{ such that } \nu_k > \ell \} \,.$

The aforementioned event $\mathcal{W}(\ell, L)$ is an intersection of these previous three events

$$\mathcal{W}(\ell,L) = \mathcal{E} \cap \left(\bigcap_{1 \leqslant i \leqslant n-L} \mathcal{S}(i,b_n,L)\right) \cap \left(\bigcap_{i \in [n]} \mathcal{B}(i,\ell,L)\right).$$

The first two intersected events in $\mathcal{W}(\ell, L)$ imply that for every element i inserted by the q-Mallows process, at most the L elements following it can potentially create an edge that crosses the vertex $i \in [n]$. The final event in the intersection states that of these L elements, all but at most ℓ will be inserted in the ℓ closest positions to the left-hand end of the permutation at the time of insertion. The next claim shows that if this event holds then the cutwidth is bounded.

Claim 38. For any integers
$$L \ge \ell \ge 1$$
 we have $W(\ell, L) \subseteq \{cw(\mathcal{P}) \le 4(\ell+1)\}$.

Proof of Claim. Recall that we fixed the injection f in the definition of the cutwidth (60) to be Id. Under f the edges of the un-permuted path $P_n \subseteq \mathsf{layer}(\sigma(P_n), P_n)$ contribute at most 1 to the cutwidth. We now bound the contribution by the edges of the permuted path $\sigma(P_n) \subseteq \mathcal{P}(n,q)$.

Given vertices $i, j, k \in [n]$ we say that the edge $jk \in E(\sigma(P_n))$ is bad (with respect to i) if j < i and k > i. We claim that if $\mathcal{W}(\ell, L)$ holds then, for any $i \in [n]$, there are at most 4ℓ edges which are bad for i. We show this by keeping track of where elements are inserted during the q-Mallows process $(r_a)_{a=1}^n$. By definition if an edge is bad for i, then exactly one of its endpoints must have value greater than i. If i = n then there are no bad edges for i, so we assume that i < n. Observe for any element $k \in \{i+1,\ldots,n\}$ there are at most 2 bad edges with endpoint k, since k can be adjacent to at most two elements j < i in the final permutation r_n . We now partition the elements $k \in \{i+1,\ldots,n\}$ into three disjoint sets A_i , B_i and C_i based on their value k and v_k , the insert position of the element k relative to the left-hand end of r_{k-1} . These sets are given by

$$A_i = \{k \in \{i+1, \dots, i+L\} : v_k > \ell\},\$$

 $B_i = \{i+1, \dots, i+L\} \setminus A_i,\$
 $C_i = \{i+L+1, \dots, n\},\$

see Figure 6 for an example. We will count the contribution of each set to the bad edges. Contribution from (A_i) : Conditional on $\mathcal{B}(i,\ell,L)$, we have $|A_i| \leq \ell$. Thus, in total the elements in A_i contribute at most 2ℓ bad edges.

Contribution from (B_i) : The left-most ℓ positions of r_i each have value at most i. By the definition of the q-Mallows process (7), no additional element with value at most i can ever occupy any of the ℓ left-most positions. Since each element $k \in B_i$ is inserted within the ℓ left-most places, there can only be 2ℓ total bad edges for i with an endpoint $k \in B_i$.

Contribution from (C_i) : Observe that if $i \ge n - L$ then C_i is empty. Otherwise, as $S(i, b_n, L)$ holds, we have $r_k(j) > i$ for all k > i + L and $j \le b_n$. That is, after step i + L of the q-Mallows process the permutation will only have elements with value at least i in its b_n left-most positions. However, conditional on \mathcal{E} , no element is inserted at position greater than b_n relative to the left-hand end of the permutation. Thus, the elements k > i + L create no bad edges.

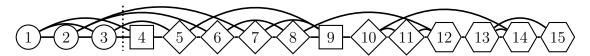


Figure 6: A visual aid to the claim in proof of Lemma 35. Observe the 'random' sequence $\mathbf{x} = (1,1,2,4,2,1,3,1,5,1,2,3,1,2,1)$ satisfies $\mathcal{W}(\ell,L)$ where with $\ell=3, L=8$ and b=5. The permutation generated by the q-Mallows process with input \mathbf{x} is $\pi=(15,13,14,10,11,12,8,6,2,7,9,5,3,1,4)$. We let x=3.5 be the location of the cut (as this maximises the cut) and observe that there are 3 bad edges for vertex 3. We also have the sets $A_3=\{4,9\}, B_3=\{5,6,7,8,10,11\}$ and $C_3=\{12,13,14,15\}$ whose vertices are represented as squares, diamonds and hexagons respectively.

Collecting these contributions, if we condition on $\mathcal{W}(\ell, L)$ then there are in total at most 4ℓ bad edges for any $i \in [n]$. For any $x \in \mathbb{R}$ there exists an i, such that all edges crossing x are either bad for i or have i as an endpoint. The result follows as there are at most 4 edges with endpoint i, since $\mathcal{P}(n,q)$ has maximum degree 4.

We now bound the failure probability of each of the constituent events in $\mathcal{W}(\ell, L)$. To begin, we set

$$L = \frac{100}{1 - q} \left(\frac{1}{1 - q} + \log n \right).$$

Recall that $q \ge 1/n^2$ and so $b_n = \lceil 8\log(n)/\log(1/q) \rceil \ge \lceil 8\log(n)/(2\log n) \rceil \ge 1$. Thus, by (15), and Lemma 37 (since $b_n \ge 1$) plus the union bound, respectively, we have

$$\mathbb{P}(\mathcal{E}^c) \leqslant n^{-7} \quad \text{and} \quad \mathbb{P}(\cup_{1 \leqslant i \leqslant n-L} (\mathcal{S}_{i,b_n,L})^c) \leqslant n \cdot o(n^{-10}) = o(n^{-9}). \quad (65)$$

We bound $\mathbb{P}(\mathcal{B}(i,\ell,L)^c)$ by showing the number of large inputs to the q-Mallows process is stochastically dominated by a sum of i.i.d. Bernoulli random variables. Observe that $\mathbb{P}(v_k > \ell) \leq q^\ell$ for any integers $\ell \leq k$ by (9), and $\mathbb{P}(v_k > \ell) = 0$ if $k < \ell$. Let $\{X_j\}_{j \in [L]}$ be a set of independent Bernoulli random variables with success probability q^ℓ . Then, since each v_k is independent, for any integers $\ell, L \geqslant 1$ and $i \in [n]$:

$$|\{k \in [i, i+L] : v_k > \ell\}| \leq \sum_{j=1}^{L} X_j.$$

If $\ell > Lq^\ell$ and $\ell >$ e, then taking $\delta = \ell/(Lq^\ell) - 1 > 0$ in Lemma 7 we have

$$\mathbb{P}\left(\sum_{j=1}^{L} X_j > \ell\right) \leqslant \left(\frac{e^{\delta}}{(1+\delta)^{1+\delta}}\right)^{Lq^{\ell}} \leqslant \left(\frac{e^{\frac{\ell}{Lq^{\ell}}-1}}{\left(\frac{\ell}{Lq^{\ell}}\right)^{\frac{\ell}{Lq^{\ell}}}}\right)^{Lq^{\ell}} \leqslant \left(\frac{eLq^{\ell}}{\ell}\right)^{\ell} \leqslant e^{\log(Lq^{\ell})\cdot\ell}. \quad (66)$$

We now break into two cases making different choices for ℓ depending on the value of q. These cases roughly correspond to the two bounds in the lemma. First, note that by (2) we have

$$1 - q \leqslant \log(1/q) \leqslant (1 - q)/q. \tag{67}$$

Case (i) $\left[1/n^2 \leqslant q \leqslant 1 - \frac{(\log \log n)^2}{\log n}\right]$: Set $\ell = 10 \cdot \sqrt{\log(n)/\log(1/q)}$, and so by (67)

$$\ell = 10 \cdot \sqrt{\frac{\log n}{\log(1/q)}} \leqslant 10 \cdot \sqrt{\frac{\log n}{1-q}} \leqslant 10 \cdot \sqrt{\frac{\log^2 n}{(\log \log n)^2}} = \frac{10 \log n}{\log \log n}.$$
 (68)

The lower bound on q in this case ensures that $\ell \geqslant 10 \cdot \sqrt{\frac{\log n}{\log(n^2)}} > e$, and (67) gives

$$L \cdot q^{\ell} = L \cdot e^{-10\sqrt{\log(1/q)\log n}} \leqslant L \cdot e^{-10\sqrt{(1-q)\log n}} \leqslant L \cdot e^{-10\sqrt{\frac{(\log\log n)^2}{\log n} \cdot \log n}} \leqslant (\log n)^2 \cdot \frac{1}{(\log n)^{10}},$$

for large n. Thus, for large n, $Lq^{\ell} < \ell$ and so (66) and the bound on ℓ from (68) give

$$\mathbb{P}\left(\sum_{j=1}^{L} X_j > \ell\right) \leqslant \left(\frac{eLq^{\ell}}{\ell}\right)^{\ell} \leqslant L^{\ell} \cdot q^{\ell^2} \leqslant \left(\log^2 n\right)^{\frac{10\log n}{\log\log n}} \cdot e^{-100\log n} \leqslant n^{-80}.$$

Case (ii) $\left[q \ge 1 - \frac{(\log \log n)^2}{\log n}\right]$: Set $\ell = 5\log(1-q)/\log q = 5\log\left(\frac{1}{1-q}\right)/\log(1/q)$. Now,

$$L \cdot q^{\ell} = \frac{100}{1 - q} \left(\frac{1}{1 - q} + \log n \right) \cdot (1 - q)^{5}$$

$$\leq 100 \left(\frac{(\log \log n)^{2}}{\log n} \right)^{3} + 100(\log n) \left(\frac{(\log \log n)^{2}}{\log n} \right)^{4}$$

$$< \frac{1}{\log n},$$

and thus $\log(Lq^{\ell}) < -\log\log n$. Also by (67) we have

$$\ell = \frac{5\log\frac{1}{1-q}}{\log(1/q)} \geqslant \frac{q \cdot 5\log\frac{1}{1-q}}{1-q} \geqslant \frac{5q}{1-q} \cdot \log\frac{\log n}{(\log\log n)^2}.$$

Hence, for suitably large n, we have

$$\ell \geqslant 5\left(1 - \frac{(\log\log n)^2}{\log n}\right) \frac{\log n}{(\log\log n)^2} \cdot (\log\log n - 2\log\log\log n) \geqslant \frac{4\log n}{\log\log n} > e.$$

To conclude this case, by (66) and the above, for suitably large n we have

$$\mathbb{P}\left(\sum_{j=1}^{L} X_j > \ell\right) \leqslant e^{\log(Lq^{\ell}) \cdot \ell} \leqslant e^{-\log(\log n) \cdot \frac{4 \log n}{\log \log n}} = n^{-4}.$$

It follows from Cases (i) and (ii) that for large n (and the appropriate $\ell := \ell(q)$) we have

$$\mathbb{P}\left(\bigcup_{i\in[n]}\mathcal{B}(i,\ell,L)^c\right)\leqslant n\cdot\mathbb{P}\left(\sum_{i=1}^LX_i>\ell\right)\leqslant n\cdot n^{-4}=n^{-3},$$

by the union bound. Thus, again for suitably large n, combining this with (65) gives

$$\mathbb{P}(W(\ell, L)^c) \leq n^{-7} + o(n^{-9}) + n^{-3} = \mathcal{O}(n^{-3}).$$

Thus by the claim and Cases (i) and (ii) we have $\mathbb{P}(\mathsf{cw}(\mathcal{P}(n,q)) > 4(\ell+1)) = \mathcal{O}(n^{-3})$ if $q \geqslant n^{-2}$. All that remains is to relate the cases to the statement of the lemma.

For the first statement of this lemma, the range $0 \leqslant q \leqslant 1 - (\log \log n)^2 / \log n$ is covered by Case (i) and the fact that $\mathbb{P}(\mathsf{cw}(\mathcal{P}(n,q)) \geqslant 2) = \mathcal{O}(n^{-3})$ if $q \leqslant 1/n^2$ from (64). We now let $q = 1 - c(\log \log n)^2 / \log n$, where $0 < c \leqslant 1$ is fixed. Then, for suitably large n, we have $\log \frac{1}{1-q} = \log \left(\frac{\log n}{c(\log \log n)^2} \right) \leqslant \log \log n$. Thus, by (67), we have

$$\sqrt{\frac{\log n}{\log(1/q)}} = \frac{\sqrt{\log(1/q)\log n}}{\log(1/q)} \geqslant \frac{\sqrt{(1-q)\log n}}{\log(1/q)} = \frac{\sqrt{c(\log\log n)^2}}{\log(1/q)} \geqslant \sqrt{c} \cdot \frac{\log\frac{1}{1-q}}{\log(1/q)},$$

and so if we set $C := C(c) = 5/\sqrt{c}$ then the first statement follows from Case (ii). The second statement follows from Case (ii) as $\frac{5}{\log q} \log(1-q) \leqslant \frac{5}{1-q} \log\left(\frac{1}{1-q}\right)$ by (67).

7 Conclusion & Open Problems

In this paper we introduced the $\mathcal{P}(n,q)$ model, found a sharp threshold for the existence of cut vertices, determined the treewidth up to a log factor, and proved bounds on the vertex expansion and diameter for restricted values of q. Roughly speaking, the tangled path has three regimes: path-like when $0 \leq q \leq 1 - \Theta(1/\log n)$, intermediate where $1 - \Theta(1/\log n) \leq q \leq 1 - \Theta(1/n)$, and expander if $1 - \Theta(1/n) \leq q \leq 1$. In the path-like regime there are no long edges and lots of independence, and in the expander regime one should be able to couple to the uniform case. The intermediate regime seems to be the trickiest and most interesting to analyze.

There are a wealth of open problems for $\mathcal{P}(n,q)$ as one could study the effect of q on almost any graph property/index of interest for sparse graphs. One fundamental problem is to determine the number of edges in $\mathcal{P}(n,q)$, recalling that we disregard multi-edges. This deceptively non-trivial problem is related to clustering of consecutive numbers in Mallows permutations [45]. It would also be nice to close the gap for treewidth by obtaining tight bounds for all q.

Theorem 2 proves that $q = 1 - \pi^2/(6 \log n)$ is a sharp threshold for containing a single vertex whose removal separates the graph into two macroscopic components. A key open problem is to determine if there is a notion of monotone property in the setting of tangled paths which guarantees the existence of a threshold (or even a sharp threshold). One candidate feature (for a property to be monotone with respect to) is the number of inversions in the permutation generating $\mathcal{P}(n,q)$. However, one issue with parameterizing by the number of inversions is the fact that the tangled paths generated by $\sigma = (\sigma_1, \ldots, \sigma_n)$ and its reverse $\sigma^R = (\sigma_n, \ldots, \sigma_1)$ are isomorphic, but the number of inversions may differ greatly as $\text{Inv}(\sigma^R) = \binom{n}{2} - \text{Inv}(\sigma)$.

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