

Convergence of the Quantile Admission Process with Veto Power

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Abstract

The *quantile admission process with veto power* is a stochastic processes suggested by Alon, Feldman, Mansour, Oren and Tennenholtz as a model for the evolution of an exclusive social group.

The model itself consists of a growing multiset of real numbers, representing the opinions of the members of the club. On each round two new candidates, holding i.i.d. μ -distributed opinions, apply for admission to the club. The one whose opinion is minimal is then admitted if the percentage of current members closer in their opinion to his is at least r . Otherwise neither of the candidates is admitted.

We show that for any μ and r , the empirical distribution of opinions in the club converges to a limit distribution. We further analyse this limit, show that it may be non-deterministic and provide conditions under which it is deterministic.

The results rely on a recent work of the authors relating tail probabilities of mean and maximum of any pair of unbounded i.i.d. random variables, and on a coupling of the evolution of the empirical r -quantile of the club with a random walk in a changing environment.

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1 Introduction

Consider the following stochastic model for increasing sets $\mathcal{S} = \{\mathcal{S}_t\}_{t \in \mathbb{N}_0}$. Let μ be an arbitrary probability distribution on \mathbb{R} , let $r \in (0, 1)$ be the *quantile* parameter of the model, and let $(X_t, Y_t)_{t \in \mathbb{N}_0}$ be a collection of pairs of i.i.d. μ -distributed random variables. For every $t \in \mathbb{N}_0$ denote the set $\mathcal{S}_t^+ := \{s \in \mathcal{S}_t : |s - \min(X_t, Y_t)| \leq |s - \max(X_t, Y_t)|\}$. Initialise the model with $\mathcal{S}_0 = \emptyset$ and for every time step $t \in \mathbb{N}$ set

$$\mathcal{S}_t = \begin{cases} \mathcal{S}_{t-1} \cup \{\min(X_t, Y_t)\}, & |\mathcal{S}_{t-1}^+| \geq r|\mathcal{S}_{t-1}| \\ \mathcal{S}_{t-1}, & \text{otherwise.} \end{cases}$$

In addition, define ν_t^* , the empirical distribution of opinions at step t , by $\frac{1}{|\mathcal{S}_t|} \sum_{s \in \mathcal{S}_t} \delta_s$ where δ_s is Dirac's delta measure at s .

This model, called the *quantile admission process with veto power*, was suggested by Alon, Feldman, Mansour, Oren and Tennenholtz in [1] as a model for the evolution of certain exclusive social clubs, that is, clubs whose present members take part in an admission procedure for screening and selecting

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new members. In our particular scenario, the distinction between candidates is represented by a one-dimensional parameter in \mathbb{R} (e.g. from political left to right). The selection itself involves two elements, a *voting system* and the founder's *veto power*. At every time step two random candidates from the general population are considered for admission to the club. Each club member then votes for the candidate whose opinion is closest to his (abstaining in case of a draw). Finally, if a candidate holding the minimal opinion received at least r -fraction of the votes (including abstainers) then such a candidate is admitted. Otherwise, both candidates are rejected.

Motivation for studying the model is further discussed in Section 1.3. We remark that the fine details of this admission procedure (such as how to handle draws) appear to be immaterial for the results. The only important consideration in selecting these is avoiding the possibility of a deadlock under which no additional members will ever be admitted.

The authors of [1] studied this model with μ which is uniform on $[0, 1]$ for arbitrary $r \neq \frac{1}{2}$. They showed that under these assumptions, ν_t^* converges to a deterministic distribution and obtained bounds on its empirical r -quantile. In addition, they showed that this process demonstrates a counter-intuitive behaviour when r is greater than $1/2$, when the empirical distribution of the club weakly converges to an atom at 1, the supremum of opinions in the population.

Here we show that ν_t^* converges weakly for any distribution μ , and analyse its form. In addition we provide necessary and sufficient conditions under which this limit distribution is deterministic, and provide several examples. In particular we recover the results of [1] and extend them to the case $r = \frac{1}{2}$, for which the limit is non-deterministic. Part of our interest in the process stems from its relations with *random walks in changing environment* and from the new probabilistic tool-set that was used handle it. For a survey of this see Section 1.4.

1.1 Results

Throughout we denote by X and Y a pair of independent, μ -distributed random variables. Denote by μ_{\min} and μ_{\max} the infimum and supremum of the support of μ , respectively, and write for $m < \mu_{\max}$:

$$\begin{aligned} \rho(m) = \rho_{\mu,r}(m) &:= \frac{\mu((m, \infty))^2}{\mu * \mu([2m, \infty))} - (1-r) \\ &= \mathbb{P}(X > m, Y > m \mid X + Y \geq 2m) - (1-r). \end{aligned} \tag{1}$$

Our main result is the following.

Theorem 1. *For any probability distribution μ on \mathbb{R}_+ , the sequence ν_t^* converges weakly to a (possibly random) distribution ν_∞ on \mathbb{R}_+ , almost surely.*

Moreover, the limit distribution ν_∞ satisfies one of the following forms:

(i) $\nu_\infty((s, \infty)) = \mathbb{P}(X > s, Y > s \mid X + Y \geq 2s)$ for $m < \mu_{\max}$ which satisfies

$$\rho(m) \leq 0 \leq \lim_{q \rightarrow m^-} \rho(q),$$

(ii) $\nu_\infty((s, \infty)) = \mathbb{P}(X > s, Y > s \mid X + Y > 2s)$ for $m < \mu_{\max}$ which satisfies

$$\rho(m) \leq 0 = \lim_{q \rightarrow m^+} \rho(q),$$

(iii) $\nu_\infty = \delta_{\mu_{\max}}$, and $0 \leq \lim_{q \rightarrow \mu_{\max}} \rho(q)$.

In case that μ is continuous, an explicit form of the possible limiting measures of the admission process ν_∞ is given by the following corollary.

Corollary 1.1. *Assume that μ is a continuous measure on \mathbb{R}_+ . Then, almost surely, either $\nu_\infty = \delta_{\mu_{\max}}$ or $d\nu_\infty(x) \propto \mu([x, \infty) \cap [2m - x, \infty)) d\mu(x)$ where m satisfies $\rho_{\mu, r}(m) = 0$.*

Observe that the empirical distribution of the club members' opinions in the veto-power process demonstrates a lighter tail near infinity, in comparison with the distribution of the general population (even though, as will become evident in Section 1.2 below, the mean of the empirical distribution of the club members' opinions is often larger than the mean of μ).

Corollary 1.2. *If $\mu_{\max} = \infty$ and $\nu_\infty = \lim_{t \rightarrow \infty} \nu_t^*$ then, almost surely, $\lim_{s \rightarrow \infty} \frac{\log \nu_\infty(s, \infty)}{\log \mu(s, \infty)} = 2$.*

However, ν_∞ need not be deterministic.

Proposition 1.3. *For any $r \in (0, 1)$, there exists a measure μ for which there are infinitely many possible limiting distributions for $\lim_{t \rightarrow \infty} \nu_t^*$, each occurring with positive probability.*

We further provide a necessary and sufficient condition for having a deterministic ν_∞ for all $r \in (0, 1)$. This extends [1, Theorem 3.5] which provides a weaker sufficient condition.

Theorem 2. *Let μ be a given probability measure. Then ν_∞ is deterministic for all $r \in (0, 1)$ if and only if $\rho_{\mu, 1}(m)$ is strictly monotone on the support of μ .*

Finally we provide the following criterion for verifying the monotonicity of ρ .

Proposition 1.4. *Let $g(x) = -\log \mu((x, \infty))$ for $x \in [\mu_{\min}, \mu_{\max}]$. If g is thrice differentiable with $g'' \geq 0$ and $g''' \leq 0$ on (μ_{\min}, μ_{\max}) , then $\rho_{\mu, 1}(m)$ is strictly monotone decreasing.*

1.2 Examples

Using these results, it is possible to compute the limit ν_∞ for many examples of μ and r . We state here a few examples of interest, omitting calculations which are straightforward. Throughout, we fix $r \in (0, 1)$.

Uniform distribution (extending [1, Section 4]). Let μ be the uniform distribution on $[0, 1]$. The function $\rho(m)$ from (1) is computed to be

$$\rho(m) = \begin{cases} \frac{(1-m)^2}{1-2m^2} - (1-r), & 0 \leq m < \frac{1}{2} \\ \frac{1}{2} - (1-r), & \frac{1}{2} \leq m \leq 1. \end{cases}$$

Using Theorem 1 we deduce different behaviour in the following cases. The first two cases recover the results of [1], whereas the case $r = \frac{1}{2}$ is new.

- $r > \frac{1}{2}$: a.s., the limit is $\nu_\infty = \delta_1$.
- $r < \frac{1}{2}$: a.s., the limit ν_∞ is an absolutely continuous measure with density proportional to

$$d\nu_\infty \propto \begin{cases} x - (2m - 1), & 0 \leq x \leq m, \\ 1 - x, & m \leq x \leq 1, \end{cases}$$

where $m \in [0, \frac{1}{2}]$ is a solution to $\frac{(1-m)^2}{1-2m^2} = 1 - r$, namely $m = \frac{1-\sqrt{1-3r+2r^2}}{3-2r}$.

Observe that as $r \rightarrow \frac{1}{2}$, the density of the limiting measure ν_∞ approaches the centred triangle function $\min(x, 1-x)$. This may be viewed as a perfect balance between the bias caused by the veto power and the bias caused by the quorum-admission condition.

- $r = \frac{1}{2}$: the limit ν_∞ is an absolutely continuous measure with density proportional to

$$d\nu_\infty \propto \begin{cases} x - (2M - 1), & 2M - 1 \leq x \leq M, \\ 1 - x, & M \leq x \leq 1, \end{cases}$$

where M is a random variable supported on $[\frac{1}{2}, 1]$. Moreover, by Lemma 2.12 below, $\mathbb{P}(M \in I) > 0$ for any $I \subseteq [\frac{1}{2}, 1]$. In particular, the limit is not deterministic (as can be verified by Theorem 2).

Normal distribution. Let $\mu \sim \mathcal{N}(0, 1)$. In this case $\frac{1}{2}\mu * \mu \sim \mathcal{N}(0, 1)$, and we obtain from (1) by direct computation that

$$\rho_{\mu,r}(m) = \frac{\mu((m, \infty))^2}{\mu([m, \infty))} - (1 - r) = \mu((m, \infty)) - (1 - r).$$

Therefore, for any $r \in [0, 1]$ the function $m \mapsto \rho_{\mu,r}(m)$ is strictly monotone on \mathbb{R} , and we obtain by Theorem 2 that ν_∞ is unique. By Theorem 1, the limiting measure ν_∞ has density proportional to

$$d\nu_\infty \propto \begin{cases} e^{-x^2/2} \int_{2m-x}^\infty e^{-t^2/2} dt, & x \leq m, \\ e^{-x^2/2} \int_x^\infty e^{-t^2/2} dt, & x > m, \end{cases}$$

where m is the unique solution to $\frac{1}{\sqrt{2\pi}} \int_m^\infty e^{-t^2/2} dt = 1 - r$.

Exponential distribution. Let $\mu \sim \text{Exp}(1)$. Then ν_t^* will almost surely converge to the unique measure ν_∞ with density

$$\frac{1}{m + \frac{1}{2}} \cdot \begin{cases} 1, & x \leq m, \\ e^{-2(x-m)}, & x > m, \end{cases}$$

where $m = \frac{r}{2(1-r)}$. Observe that when r tends to 1, m tends to infinity, demonstrating a drift to the right much like the uniform case.

Compressed exponential distributions. Consider the measure $\mu([x, \infty)) = e^{-x^\alpha} \mathbb{I}(x \geq 0)$ for $\alpha \in [1, 2]$. By Proposition 1.4 and Theorem 2, for all r the limit ν_∞ is deterministic. By Corollary 1.1 it has density which is proportional to

$$\alpha x^{\alpha-1} e^{-x^\alpha} \cdot \begin{cases} e^{-(2m-x)^\alpha}, & x \leq m \\ e^{-x^\alpha}, & x > m, \end{cases}$$

where m is the unique solution to

$$e^{2m^\alpha} \left(\int_0^{2m} \alpha x^{\alpha-1} e^{-x^\alpha - (2m-x)^\alpha} dx + e^{-(2m)^\alpha} \right) = \frac{1}{1-r}.$$

For $\alpha = 2$ this reduces to $\int_0^{2m} 2x e^{-2(x-m)^2} dx + e^{-2m^2} = \frac{1}{1-r}$, so that m is of the order $\frac{1}{1-r}$.

1.3 Background and motivation

The study of evolving social groups originates in sociology and economics, where analysis of such processes is expected to improve our understanding of human social behaviour, allowing us to anticipate the implications of various factors involved in the evolution of clubs, corporations, organizations and societies. At the same time, the models suggested for the evolution of social groups turned out to be of independent interest for the development of probability theory.

Sociological interest. Most of the research conducted on evolving social groups revolves around *decentralised* social groups, in which local laws are used to govern the admission into the group. These include social networks, business relations and the set of individuals holding a certain opinion. These social groups are characterised by the fact that admission of a new member into the group could be initiated by each of a certain class of members without any interaction with the rest of the members. The evolution of such groups is somewhat reminiscent of condensation behaviours in statistical mechanics. Indeed, such models were recently given the title “Statistical Physics of Social Dynamics” by Castellano, Fortunato and Loreto [6]. In particular, a few models for opinion formation and formation of social groups were suggested and analysed, see e.g. [11] and references within. Like our model, these models represent opinions by numbers, or vectors in Euclidian space [6, Chapter III], and association is governed by homophily (i.e., proximity of opinion, see [12]).

In [1], however, Alon et al. investigated models for the evolution of *centralised* social groups, where admission is governed by a global procedure, such as a voting process or a screening committee. Examples for such groups are ample and diverse, ranging from secret societies, to members of the national academy of sciences and to the inner circle of a martial art association. They include communes and condominiums, justices of the supreme court, associate professors of a math department and nations in the united nations. The evolution of the composition of such groups could vary greatly depending on the nature of the admission process. One purpose of the authors was to study the different behaviours of models in which the admission is relative, i.e., based on the comparison of several candidates. This they do both for growth models like the one discussed here and for models of population with fixed size. In particular they explore the difference between a voting model, where the candidate who received the highest number is admitted, and a consensus model, where all members must agree on the most suitable candidate. As an intermediate model they suggested the r -quantile admission process, where a candidate must obtain a quorum of a certain fraction of the votes in order to be admitted. Such admission models are common for the selection of justices and governmental committees.

One property of the relative consensus model is its radicalisation behaviour. In a population holding a uniform distribution of opinion in $[-1, 1]$, for example, it almost surely converges to atomic measures at ± 1 . This behaviour is explained by the fact that the only case in which new members are admitted is when both candidates are extremists in the same direction, in which case the milder among the two is admitted. To better understand this radicalisation effect and its interaction with r in the r -quantile admission process, the authors put it against a body holding a veto power who allows only admittance of the member whose opinion is minimal. Procedures with similar effects are used in practice by conservative and professional organisations such as the inner-circle of an organization for preserving a traditional craft. In these cases the opinion parameter represents a certain professional standard and the veto is usually exercised by supervising committee. In [1], it is shown that if the distribution of opinions is uniform, whenever $r > \frac{1}{2}$, the radicalization effect of the quorum dominates the veto and the process radicalises to an atom at 1, making the veto an ineffective method for preventing the group’s opinions from radicalising upwards.

In this work we provide a rigorous analysis of the r -quantile admission process with veto power, which results in a better understanding of the competition between the drift upwards caused by r , the high percentage of support required for admission, and μ , the natural distribution of the population. We show that the opinion profile of the club always converges weakly and that while the drift induced by the quorum pressure persists for any distribution, it is never sufficiently strong to make the empirical r -quantile opinion of the group tend to infinity. In addition, our tools appear to be useful in the analysis of other quantile driven processes and we hope that they will find applications for other rules of admission

and in opinion spaces of higher dimensions. In particular we believe them to be relevant for studying the r -quantile admission process without veto power. See Problem 2 below.

Mathematical interest. The process we analyse is an example of a *random walk in changing environment* (RWCE), that is, a stochastic process for which in every step the environment which determines the transition probabilities between possible states changes (in our case, the set of possible values of the r -quantile is effected by the history and the opinions of the current members of the group). RWCEs were recently studied by Amir, Benjamini, Gurel-Gurevich and Kozma [2], by Avin, Koucký and Lotker [3], by Dembo, Huang and Sidoravicius [7, 8], and by Redig and Völlering [13] to name but a few. It includes examples which were investigated for their own sake, such as reinforced random walk, bridge-burning random walk, loop-erased random walk, and self-avoiding walk with bond repulsion (see references within [2]). Typically, one is interested in properties such as convergence and recurrence vs. transience of the process albeit the changing environment. As pointed out by several examples in [2], these questions are not at all trivial.

While this paper carries out careful analysis of a certain family of RWCE (characterised by r and the distribution of opinions in the population μ), we believe that it may possibly be generalised to a larger family of RWCE that goes beyond quantile driven processes. One neat variant which could be addressed using our methods is a process in which at the ticks of a Poisson clock a point is added to the real line according to some absolutely continuous distribution supported on \mathbb{R} . A walker makes a simple random walk on these points at the ticks of an independent Poisson clock, moving from each point to one of its immediate neighbours. The probability that the random walk at point x moves to its right neighbour depends on a continuous $f(x) : \mathbb{R} \rightarrow [0, 1]$. Our methods could be used to show that such a walk always converges either to a point x satisfying $f(x) = \frac{1}{2}$, to ∞ if $\lim_{x \rightarrow \infty} f_x \geq \frac{1}{2}$ or to $-\infty$ if $\lim_{x \rightarrow -\infty} f_x \leq \frac{1}{2}$.

In addition, this work provides an application to a general inequality relating the average of two i.i.d. random variables to their minimum [9], demonstrating how it could be applied to control the recurrence of certain random processes.

1.4 Outline and highlights

In this section we survey the tools and methods used to obtain our results, providing a rough outline of key steps in our proof. To make this outline more accessible we restrict ourselves here to the case of continuous μ , avoiding the difficulties that arise from handling measures with an atomic component.

The first step in proving the convergence of ν_t^* is its reduction in Section 2 to the convergence of two probabilities: $\mathbb{P}(X + Y \geq 2m)$ and $\mathbb{P}(X \geq m)$ for X, Y i.i.d. μ -distributed. To analyse the behaviour of these probabilities we make a time change to the process, examining it at times in which a new member is admitted, thus defining the empirical q -quantile process m_k . We then provide four key lemmata (see Section 2), showing that m_k is bounded and classifying regions which it can visit only finitely many times. These lemmata are derived by studying the evolution of m_k via a coupling with y_k , a certain random walk in changing environment, presented in Section 3. To define y_k we regard the members of \mathcal{S} as the points $\{1, 2, \dots, k\}$ translated by some common integer shift Δ_k . Under this map, we let $y_k = rk + \Delta_k$ be the location of the empirical r -quantile after k members have been admitted. The translation Δ_k is chosen so that $\mathbb{E}(y_{k+1} - y_k) = \rho(m_k)$ and the steps of y_k are regular. Using continuity properties of ρ together with classical properties of drifting random walks, we show in Section 4 that if $\rho(x) \neq 0$ then there exists an interval I containing x which m_k can visit only finitely many times. Moreover, we show that if $\rho(x) > 0$ then m_k has positive probability to eventually stay above x , while

if $\rho(x) < 0$ then m_k has positive probability to eventually stay below x . To conclude our proof we require two more ingredients. The first is an argument, which essentially shows that an unbiased simple random walk on such an evolving lattice visits every point only finitely many times. This is used to show convergence of $\mathbb{P}(X + Y \geq 2m_k)$ and $\mathbb{P}(X \geq m_k)$ if m_k ends up in a region I with $\mu(I) > 0$ and $\rho(x) = 0$ for all $x \in I$. Finally to show that m_k is bounded we use the following general comparison inequality established in [9, Theorem 1].

Theorem A. *Let X, Y be independent random variables on \mathbb{R}_+ , which are not compactly supported. Then:*

$$\liminf_{m \rightarrow \infty} \frac{\mathbb{P}(\min(X, Y) > m)}{\mathbb{P}\left(\frac{X+Y}{2} \geq m\right)} = 0.$$

The quotient here and in (1) are the same, and both represent the probability that the opinion of the next admitted member is above or equal to the empirical r -quantile m . Using this we show that there are sufficiently many points with $\rho(x) < 0$ on \mathbb{R} so that eventually m_k will stay bounded below one of them. Finally we use compactness arguments together with all of these observations to show convergence of m_k to a non-empty interval J satisfying $\mu(J) = 0$ and $\rho(x) = 0$ for all $x \in J$.

Finally, in Section 5 we utilise elementary tools to analyse the example mentioned in Proposition 1.3 and prove our criterion for uniqueness given in Proposition 1.4.

1.5 Open Problems

We conclude the introduction with two open problems.

Problem 1. *Does the r -quantile admission process with veto power converge for any distribution μ over \mathbb{R}^d (including $d = 1$), where the founder may hold any opinion in \mathbb{R}^d ? What about other metric spaces?*

Problem 2. *Does the r -quantile admission process without veto power converge for all μ, r ? Does the drift to the extreme phenomenon persist in this model?*

It is also of interest to fully classify the distribution of ν_∞ in cases for which it is not deterministic, as this would complete the analysis of the r -quantile admission process with veto power on \mathbb{R} .

2 From convergence of measures to convergence of quantiles

In this section we reduce Theorems 1 and 2 to statements concerning with the evolution of the empirical r quantile of the *quantile admission process with veto power*. All theorems are reduced to a couple of propositions concerning this process, which we further reduce to several technical statements. Tools for proving these statements are provided in Section 3 and their proofs are given in Section 4.

2.1 Notation

The following notation is used throughout. For $A, B \subseteq \mathbb{R}$, denote $A + B = \{a + b : a \in A, b \in B\}$. We write $a > B$ if $a > b$ for every $b \in B$ and $A > B$ if $a > B$ for every $a \in A$. We follow the convention that the infimum of an empty set is infinity. We employ the abbreviations *i.o.* for *infinitely often*, *a.e.* for *almost everywhere* and *a.s.* for *almost surely*.

Throughout, we fix the probability measure μ and the quantile r , and let X and Y be two independent μ distributed random variables. We define the events

$$\begin{aligned}
D_m &:= \left\{ \frac{X+Y}{2} \geq m \right\} & D_{m+} &:= \left\{ \frac{X+Y}{2} > m \right\} \\
F_m &:= \mathbb{P}(D_m) = \mu * \mu([2m, \infty)) & F_{m+} &:= \mathbb{P}(D_{m+}) = \mu * \mu((2m, \infty)).
\end{aligned}$$

We consider our process mostly in times when a member is admitted to the club. To this end, we define a random sequence $\{x_k\}_{k \in \mathbb{N}_0}$ of the opinions of club members by order of admission. We further denote the multiset $S_k = \{x_0, \dots, x_{k-1}\}$ and define the lower r -quantile of the set S_k by

$$m_k = \min \left\{ x \in \mathbb{R}_+ : |\{j < k : x_j \leq x\}| \geq rk \right\} = \sup \left\{ x \in \mathbb{R}_+ : |\{j < k : x_j < x\}| < rk \right\}. \quad (2)$$

Observe that the distribution of x_k is that of $\min(X, Y)$ conditioned on $D_{m_k} = \{\frac{X+Y}{2} \geq m_k\}$.

We denote \mathcal{F}_k for the sigma algebra generated by x_0, \dots, x_k and write ν_k for the empirical distribution of opinions in $S_{k+1} = \{x_0, \dots, x_k\}$, given by $\frac{1}{|S_{k+1}|} \sum_{s \in S_{k+1}} \delta_s$. By the definition of ν_t , the measures ν_t and ν_t^* converge together and to the same limit. For this reason, the reader need not be alarmed by the visual similarity between S_k and \mathcal{S}_t (used in the introduction) as the latter will play no role in the remainder of the paper.

To simplify the treatment of ρ we introduce

$$\rho^+(m) = \rho_{\mu,r}^+(m) := \lim_{q \rightarrow m-} \rho_{\mu,r}(q). \quad (3)$$

Notice that if $\mu_{\max} < \infty$ then

$$\rho^+(m) = \frac{\mu([m, \infty))^2}{\mu * \mu([2m, \infty))} - (1-r). \quad (4)$$

We observe the following basic properties of ρ and ρ^+ .

Observation 2.1. ρ is lower-semi-continuous and ρ^+ is left-continuous.

$\rho(x) \leq \rho^+(x)$ for all x . Moreover $\rho(x) < \rho^+(x)$ if and only if $\mu(\{x\}) > 0$.

$\rho(x) \leq \lim_{t \rightarrow x+} \rho(t)$ for all x . Moreover $\rho(x) < \lim_{t \rightarrow x+} \rho(t)$ if and only if $\mu * \mu(\{2x\}) > 0$.

Observation 2.2. If $\mu(\{x\}) > 0$, then $\rho(x) < \lim_{y \rightarrow x+} \rho(y)$.

Observation 2.3. ρ and ρ^+ are monotone increasing on any interval I satisfying $\mu(I) = 0$.

Observation 2.4. $\lim_{q \rightarrow -\infty} \rho(q) = \lim_{q \rightarrow -\infty} \rho^+(q) = r$.

2.2 Reduction of Theorems 1 and 2

We begin by reducing Theorem 1 to the following proposition, whose proof we postpone to Section 2.4.

Proposition 2.5. Almost surely there exists a finite $m \in [0, \mu_{\max}]$ such that the sequence F_{m_k} converges to a limit F_ℓ , for $\ell \in \{m, m+\}$. Moreover,

$$\begin{aligned}
&\text{if } \ell = m < \mu_{\max}, \text{ then } \rho(m) \leq 0 \leq \rho^+(m), \\
&\text{if } \ell = m+ < \mu_{\max}, \text{ then } \rho(m) \leq 0 = \lim_{q \rightarrow m+} \rho(q), \\
&\text{while if } m = \mu_{\max}, \text{ then } \rho^+(\mu_{\max}) \geq 0.
\end{aligned}$$

Proof of Theorem 1. For $s \in [0, \infty)$ denote $B_s = \{(x, y) : \min(x, y) > s\}$ and D_s as above. Observe that for any $s, m' \geq 0$ we have,

$$p_{s,k} := \mathbb{P}(x_k > s) = \mathbb{P}\left(\min(X, Y) > s \mid \frac{X+Y}{2} \geq m_k\right) = \frac{\mathbb{P}((X, Y) \in D_{m_k} \cap B_s)}{\mathbb{P}((X, Y) \in D_{m_k})}. \quad (5)$$

By Proposition 2.5 we obtain that $\lim_{k \rightarrow \infty} \mathbb{P}(D_{m_k}) = \lim_{k \rightarrow \infty} F_{m_k}$ almost surely exists and is of the form F_ℓ for $\ell \in \{m, m+\}$. Since $\{D_s\}_{s \in \mathbb{R}}$ are ordered by inclusion, we have $\mathbb{I}(D_{m_k}) \xrightarrow[k \rightarrow \infty]{\mu} \mathbb{I}(D_\ell)$. Hence for any measurable set $B \subset \mathbb{R}_+^2$ we have

$$\mathbb{I}(D_{m_k} \cap B) = \mathbb{I}(D_{m_k}) \mathbb{I}(B) \xrightarrow[k \rightarrow \infty]{\mu} \mathbb{I}(D_\ell) \mathbb{I}(B) = \mathbb{I}(D_\ell \cap B). \quad (6)$$

In particular the numerator of (5) converges as k tends to infinity, almost surely. To show convergence of right-hand-side of (5), let us consider separately the case $F_m = 0$ and the case $F_m > 0$.

In the case $F_m = 0$, we have $m_k \xrightarrow{\text{a.s.}} \inf\{s : \mathbb{P}((X+Y)/2 \geq s) = 0\} = \mu_{\max}$. By Proposition 2.5, m_k is almost surely bounded and hence $\mu_{\max} < \infty$. By definition, the distribution of x_k is supported on $[2m_k - \mu_{\max}, \mu_{\max}]$, so that ν_k converges weakly to $\delta_{\mu_{\max}}$. By the moreover part of Proposition 2.5, this can only occur if condition (iii) in Theorem 1 is satisfied.

In the case $F_m > 0$ we have, almost surely, for all $s \geq 0$,

$$\lim_{k \rightarrow \infty} p_{s,k} = \frac{\mathbb{P}((X, Y) \in D_\ell \cap B_s)}{\mathbb{P}((X, Y) \in D_\ell)}.$$

We conclude that x_k converges weakly, so that, by the law of large numbers ν_k converges weakly to a limiting distribution ν_∞ . If $\ell = m < \mu_{\max}$, then, by the moreover part of Proposition 2.5 the condition $\rho(m) \leq 0 \leq \rho^+(m)$ is satisfied, so that, considering the limit distribution implied by (5) and (6), we obtain that item (i) of Theorem 1 holds. Similarly, if $\ell = m+ < \mu_{\max}$, then by the moreover part of Proposition 2.5 the conditions $\rho(m) \leq 0 \leq \lim_{q \rightarrow m-} \rho(q)$ and $\lim_{q \rightarrow m+} \rho(q) = 0$ are satisfied, so that, using (5) and (6), item (ii) of Theorem 1 holds. Finally, if $m = \mu_{\max}$ then, by similar arguments, item (iii) of Theorem 1 holds. \square

Next we obtain the sufficiency of the criterion in Theorem 2 as a corollary of Theorem 1.

Proof of the sufficiency criterion in Theorem 2. Writing S_- and S_+ for the support of the negative and positive parts of ρ respectively, we obtain from the strict monotonicity of ρ that either $S_+ = \emptyset$, or there exists a unique m' such that $\sup S_+ = \inf S_- = m'$.

If S_+ is empty, then both conditions for items (i) and (ii) in Theorem 1 cannot hold for any m , and hence item (iii) of the theorem must hold, so that ν_t^* converges to $\delta_{\mu_{\max}}$, as required. Otherwise, $\sup S_+ = \inf S_- = m'$, so that, by Theorem 1, ν_t^* converges to ν_∞ such that $\nu_\infty((s, \infty))$ is proportional to either $\mathbb{P}(X > s, Y > s, X+Y \geq 2m)$ or $\mathbb{P}(X > s, Y > s, X+Y > 2m)$. Since, by definition, $\rho(m)$ is a translation of the quotient of two monotone decreasing functions, where the numerator is right continuous and the denominator is left continuous, it can be monotone decreasing itself only if the denominator is continuous, that is, $\mathbb{P}(X+Y \geq 2m) = \mathbb{P}(X+Y > 2m)$ for every m . Hence also $\mathbb{P}(X > s, Y > s, X+Y \geq 2m) = \mathbb{P}(X > s, Y > s, X+Y > 2m)$ so that ν_∞ is uniquely determined and explicit, as required. \square

In order to establish the necessity of the criterion in Theorem 2 we use the following proposition, whose proof we postpone to Section 2.5.

Proposition 2.6. *For any $x \in \mathbb{R}_+$, if $\rho(x) < 0$ then $\mathbb{P}(m_k \leq x \text{ a.e.}) > 0$, while if $\rho(x) > 0$ then $\mathbb{P}(m_k > x \text{ a.e.}) > 0$. Moreover, if $a < b$ are such that $\mu([a, b]) > 0$ and $\rho(x) = 0$ for all $x \in [a, b]$, then $\mathbb{P}(m_k \in [a, b] \text{ a.e.}) > 0$.*

Proof of the necessity of the criterion in Theorem 2. We begin by showing the first part of the theorem. Recall that $\rho_{\mu,1}(0) = 1$ and $\liminf_{q \rightarrow \mu_{\max}} \rho_{\mu,1}(q) = 0$ by Theorem A. Thus, under the assumption that $\rho_{\mu,1}(q)$ is not monotone, there exist $x_0 < x_1$ and $r \in [0, 1]$ such that $\rho_{\mu,1}(x_0) < 1 - r < \rho_{\mu,1}(x_1)$. Since $\rho_{\mu,r}(q) = \rho_{\mu,1}(q) - (1 - r)$ this means that $\rho_{\mu,r}(x_0) < 0 < \rho_{\mu,r}(x_1)$. By Proposition 2.6, there is a positive probability that $m_k \leq x_0$ for almost every k , in which case, by Theorem 1, ν_t converges to a measure ν_∞ of the form given in (i) or (ii) for some $m \leq x_0$. On the other hand, by Proposition 2.6 there is a positive probability that $m_k > x_1$ for almost every k , in which case, by Theorem 1, ν_t converges to a measure ν_∞ as in (i) or (ii) with $m \geq x_1$ or as in (iii). As a measure cannot satisfy both requirements, we conclude that ν_∞ is non-deterministic.

Next, we show the moreover part. Suppose that ρ vanishes on $[a, b]$ where $a < b$ and $\mu([a, b]) > 0$. By Observation 2.1, the interval $[a, b]$ contains no atoms of μ . Hence we may restrict ourselves to the case $b < \mu_{\max}$, as otherwise we can apply our arguments to a subinterval of positive measure whose upper end satisfies this property. Observing that F_s is monotone decreasing in s and that $F_s > F_{s'}$ whenever $\mu([s, s']) > 0$, we obtain the existence of $a < a' < b' < b$ such that $F_a > F_{a'} > F_{b'} > F_b$. From Proposition 2.5 we obtain that F_{m_k} converges to a limit F_ℓ . By the second part of Proposition 2.6 both the event $A = \{m_k \in [a, a'] \text{ a.e.}\}$ and the event $B = \{m_k \in [b', b] \text{ a.e.}\}$ have positive probability. In the former case we obtain $F_\ell \geq F_{a'}$ while in the latter $F_\ell \leq F_{b'}$. We conclude that the values of F_ℓ under the event A and under the event B must be almost surely distinct. Observing that in both cases $\nu_\infty([b, \infty)) = \mu([b, \mu_{\max}))^2 / F_\ell$, where the numerator is a non-zero constant, independent of F_ℓ , we conclude that $\nu_\infty([b, \infty))$ is not a constant random variable, as required. \square

Thus we are left with proving Propositions 2.5 and 2.6. In the next section we reduce these propositions to several technical statements.

2.3 Properties of the empirical quantile process

In this section we present several lemmata concerning with the empirical quantile process m_k which will be of use in the proof of Propositions 2.5 and 2.6. We begin by introducing several definitions.

Definition 2.7 (Barrier). *A point $x \in \mathbb{R}$ is called a barrier if there exists k_0 such that either $m_k \leq x$ for all $k \geq k_0$, or $m_k > x$ for all $k \geq k_0$.*

A point $x \in \mathbb{R}$ is called a right-barrier if it is a.s. a barrier and $\mathbb{P}(m_k \leq x \text{ a.e.}) > 0$. Similarly, $x \in \mathbb{R}$ is called a left-barrier if it is a.s. a barrier and $\mathbb{P}(m_k > x \text{ a.e.}) > 0$.

Definition 2.8 (Separator). *An interval I is called a separator if there exists k_0 such that either $m_k < I$ for all $k \geq k_0$, or $m_k > I$ for all $k \geq k_0$.*

An interval I is called a right-separator if it is a.s. a separator and $\mathbb{P}(\exists k_0 : m_k < I \text{ for all } k \geq k_0) > 0$. Similarly, I is called a left-separator if it is a.s. a separator and $\mathbb{P}(\exists k_0 : m_k > I \text{ for all } k \geq k_0) > 0$.

To prove our propositions, we require the following lemmata.

Lemma 2.9 (Boundedness). *m_k is almost surely bounded.*

Lemma 2.10 (Negative drift). *Let $x_0 \leq \mu_{\max}$.*

- (a) If $\rho^+(x_0) < 0$, then there exists $a < x_0$ such that $(a, x_0]$ is a right-separator.
- (b) If $\rho(x_0) < 0$, then x_0 is a right-barrier.
- (c) If $\lim_{x \rightarrow x_0+} \rho(x) < 0$, then there exists $b > x_0$ such that (x_0, b) is almost surely a separator.

Lemma 2.11 (Positive drift). *Let $x_0 < \mu_{\max}$.*

- (a) If $\rho(x_0) > 0$, then there exist $a < x_0 < b$ such that (a, b) is a left-separator.
- (b) If $\rho^+(x_0) > 0$, then x_0 is almost surely a barrier.
- (c) If $\lim_{x \rightarrow x_0+} \rho(x) > 0$, then there exists $b > x_0$ such that (x_0, b) is almost surely a separator.

Lemma 2.12 (Zero drift). *Let $[a, b]$ be an interval such that $\mu((a, b)) > 0$, and $\rho(x) = \rho^+(x) = 0$ for all $x \in [a, b]$. Then $[a, b]$ almost surely contains a barrier. Moreover, $\mathbb{P}(m_k \in [a, b] \text{ a.e.}) > 0$.*

From these we also derive the following corollary.

Corollary 2.13. *let $a \leq b$ such that $F_a < F_{b+}$, then $[a, b]$ a.s. contains a barrier.*

Proof. We consider three cases. If there exists $x \in [a, b]$ such that $\rho(x) < 0$, then, by Lemma 2.10b, the point x is a.s. a barrier. Similarly, if there exists $x \in [a, b]$ such that $\rho^+(x) > 0$, then, by Lemma 2.11b, the point x is a.s. a barrier. Since $\rho(x) \leq \rho^+(x)$ for all x (by Observation 2.1), we obtain in the remaining case, $\rho(x) = \rho^+(x) = 0$ for all $x \in [a, b]$. We observe that, by definition, F_x is monotone decreasing on any interval of μ -measure zero. Hence by the assumption $F_a < F_{b+}$ together with (3), we have $\mu[a, b] > 0$. By Observation 2.1 we have $\mu(\{a\}) = \mu(\{b\}) = 0$ so that $\mu(a, b) > 0$. Thus, by Lemma 2.12, the interval (a, b) almost surely contains a barrier. \square

2.4 Proof of Proposition 2.5

By Lemma 2.9, the sequence m_k is almost surely bounded. In addition, F_ℓ is left-continuous and monotone decreasing, so that if F_{m_k} converges, then its limit must be of the form F_ℓ for $\ell \in \{m, m+\}$ for a finite $m \in \mathbb{R}$.

Let $\{m_{k_i}\}_{i \in \mathbb{N}}$ and $\{m'_{k'_i}\}_{i \in \mathbb{N}}$ be two convergent subsequences of $\{m_k\}_{k \in \mathbb{N}}$ whose limits we denote by m and m' respectively, and assume without loss of generality that $m' \leq m$. To show convergence of F_{m_k} it would suffice to show that

$$\lim_{i \rightarrow \infty} F_{m'_{k'_i}} = \lim_{i \rightarrow \infty} F_{m_{k_i}}. \quad (7)$$

For any $\varepsilon > 0$, we have $m_k < m' + \varepsilon$ for infinitely many values of k and $m_k > m - \varepsilon$ for infinitely many values of k so that the interval (m', m) does not contain any barrier. By Corollary 2.13, this implies that $F_{m'} = F_m$, almost surely.

To derive convergence we consider two cases. If $F_m = F_{m+}$, then

$$F_m = F_{m+} \leq F_{m'+} \leq F_{m'} = F_m,$$

so that (7) holds with both sides equal to $F_m = F_{m+} = F_{m'} = F_{m'+}$. On the other hand, if $F_m \neq F_{m+}$, then m is a barrier by Corollary 2.13. Hence either $m_k \leq m$ for almost all k , or $m_k > m$ for almost all k and $m = m'$. In the latter case we immediately obtain (7) with both sides equal F_{m+} . In the former, since $F_m \leq F_{m'+} \leq F_{m'} = F_m$, we obtain that (7) holds with both sides equal to F_m .

Next, we establish the moreover part of the proposition. Let $\{m_{k_i}\}_{i \in \mathbb{N}}$ be a monotone convergent subsequence of $\{m_k\}_{k \in \mathbb{N}}$ and denote its limit by $\ell \in \{m, m+\}$, and observe that, by definition, there cannot be a separator (a, b) containing m .

First, observe that if $\mu_{\max} < \infty$ and $\rho^+(\mu_{\max}) < 0$ then by Lemma 2.10a there exists $a < \mu_{\max}$ such that $(a, \mu_{\max}]$ is a separator. Therefore if $m = \mu_{\max} < \infty$ then $\rho^+(\mu_{\max}) \geq 0$.

Henceforth we assume that ρ is defined at m . Denote $S^+ = \{x : \rho(x) > 0\}$. For each $x \in S^+$ write $I_x^+ := (a_x, b_x)$ for the separator guaranteed by Lemma 2.11a and observe that $m \notin (a_x, b_x)$ almost surely. The collection $\{I_x^+ : x \in S^+\}$ is an open cover of S^+ . Since \mathbb{R} is second-countable, this collection has a countable subcover. Since m is almost surely not in any of the sets of this subcover, we deduce that $m \notin S^+$ almost surely, that is,

$$\rho(m) \leq 0. \quad (8)$$

Similarly, write $S^- = \{x : \rho^+(x) < 0, \lim_{t \rightarrow m+} \rho(t) \neq 0\}$. For each $x \in S^-$ write $I_x^- := (a_x, b_x)$ where $(a_x, m]$ is the separator guaranteed by Lemma 2.10a, and (m, b_x) is the separator guaranteed by either Lemma 2.10c or Lemma 2.11c, depending on the sign of $\lim_{t \rightarrow m+} \rho(t)$. Observe that $m \notin (a_x, b_x)$ almost surely. As before, the collection of separators $\{I_x^- : x \in S^-\}$ is an open cover of S^- so that $m \notin S^-$ almost surely. We deduce that

$$\text{either } \rho^+(m) \geq 0 \text{ or } \lim_{t \rightarrow m+} \rho(t) = 0. \quad (9)$$

Next, denote by A the set x such that $2x$ is an atom of $\mu * \mu$. Observe that for all $x \in A$ we have $F_x \neq F_{x+}$. For any $x \in A$ which satisfies $\rho^+(x) < 0$, we have, by Lemma 2.10a, that there exists $a < m$ such that $(a, x]$ is a separator so that m_{k_i} can only converge to x from above almost surely, i.e., if $m = x$ then $F_\ell = F_{m+} \neq F_m$ almost surely. Similarly, for any $x \in A$ which satisfies $\lim_{t \rightarrow x+} \rho(t) \neq 0$ we have, by Lemma 2.10c and Lemma 2.11c, that almost surely there exists $b > x$ such that (x, b) is a separator. Thus, almost surely, m_{k_i} can only converge to x from below, so that if $m = x$ then $F_\ell = F_m \neq F_{m+}$ almost surely. Since A is countable we deduce that in case that $m \in A$, the proposition is satisfied.

Finally for any $x \notin A \cup S^+ \cup S^-$, by Observation 2.1 and the fact that A contains the atoms of μ , we have $\rho(x) = \rho^+(x) = \lim_{t \rightarrow x+} \rho(t)$. By combining (8) and (9) we deduce that in case $m \notin A \cup S^+ \cup S^-$ we have $\rho(m) = \rho^+(m) = \lim_{t \rightarrow m+} \rho(t) = 0$. Since in this case both $F_m = F_{m+}$, the proposition is satisfied. \square .

2.5 Proof of Proposition 2.6

The case $\rho(x) < 0$ is immediate from Lemma 2.10b, the case $\rho(x) > 0$ of the first part is immediate from Lemma 2.11a. The moreover part is immediate from Lemma 2.12. \square .

3 Preliminaries

In this section we establish the probabilistic infrastructure required to prove lemmata 2.9-2.12. In Section 3.1 we introduce facts about general random walks with drift. In Section 3.2 we construct a coupling of the quantile process with a random walk in changing environment and introduce relevant notation for its analysis. In Section 3.3 we present a simple claim about continuous probability measures.

3.1 On general random walks

Let $\{A_k\}_{k \geq k_0}$ be a sequence of events, adapted to a filtration $\{\mathcal{F}_k\}$. Denote the stopping time $T_{\{A_k\}} := \min\{k \geq k_0 : A_k \text{ occurred}\}$. The first result we recall is Azuma's inequality [4] concerning martingales.

Lemma 3.1 (Azuma). *Let X_k be a random process started at $X_0 = 0$. Assume that*

- (X_k is a martingale) for all k we have $\mathbb{E}(X_{k+1} \mid X_0, \dots, X_k) = X_k$.
- (uniformly bounded steps) almost surely $|X_{k+1} - X_k| \leq 1$ for all $k \in \mathbb{N}$.

Then for every $\alpha > 0$ we have $\mathbb{P}(T_{\{X_k \geq \alpha\}} < n) \leq \exp(-\frac{\alpha^2}{2n})$.

We also recall Hoeffding's inequality for i.i.d. binomial random variables.

Lemma 3.2 (Hoeffding). *If $X \sim \text{Bin}(n, p)$, then for any $\varepsilon > 0$ we have $\mathbb{P}(X - pn \geq \varepsilon n) \leq e^{-2\varepsilon^2 n}$ and $\mathbb{P}(X - pn \leq -\varepsilon n) \leq e^{-2\varepsilon^2 n}$.*

We need the following variation on the classical “gambler's ruin” problem (a.k.a. Cramér-Lundberg inequality). Its proof, being standard, is omitted (see [14] and [10, Ch. XIV.2] for similar results).

Lemma 3.3. *Let X_k be a random process started at $X_0 = 0$ with the following properties:*

- *there exists $\eta > 0$ such that for all $k \in \mathbb{N}$ we have: $\mathbb{E}(X_{k+1} \mid X_0, \dots, X_k) \leq X_k - \eta$.*
- *there exists $M > 0$ such that almost surely $|X_{k+1} - X_k| \leq M$ for all $k \in \mathbb{N}$.*

Then there exists $q = q_{M,\eta} \in (0, 1)$ such that for all $\ell > 0$ we have:

$$\mathbb{P}(T_{\{X_k \geq \ell\}} < \infty \mid X_0 = 0) < q^\ell.$$

We use this lemma to obtain the following bound.

Lemma 3.4. *Let $\ell_0 \in \mathbb{N}$ and let $\{\ell_i\}_{i \in \mathbb{N}}$ be a sequence of random variables taking values in \mathbb{N}_0 , adapted to a filtration $\{\mathcal{F}_0^{(i)}\}_{i \in \mathbb{N}}$ (i.e., $\ell^{(1)}, \ell^{(i)}$ are $\mathcal{F}_0^{(i)}$ -measurable). Further let $\{X_k^{(i)}\}_{i \in \mathbb{N}, k \in \mathbb{N}_0}$ be a collection of random processes on \mathbb{R} , started at $X_0^{(i)} = 0$, which are adapted to a filtration $\{\mathcal{F}_1^{(i)}\}_{i \in \mathbb{N}}$ (i.e., $X^{(1)}, \dots, X^{(i)}$ are $\mathcal{F}_1^{(i)}$ -measurable), and assume that for all $i, k \in \mathbb{N}$ $|X_k^{(i)} - X_{k-1}^{(i)}| \leq 1$ almost surely, and assume that $\mathcal{F}_0^{(i)} \subseteq \mathcal{F}_1^{(i)} \subseteq \mathcal{F}_0^{(i+1)}$ for all $i \in \mathbb{N}$. Finally, let τ_i be a stopping time for $\{X_k^{(i)}\}_{k \in \mathbb{N}}$.*

Assume that the following conditions hold for some $\eta, p > 0$:

1. *For all $k < \tau_i$ we almost surely have $\mathbb{E}(X_{k+1}^{(i)} \mid X_0^{(i)}, \dots, X_k^{(i)}, \mathcal{F}_0^{(i)}) \leq X_k^{(i)} - \eta$,*
2. *We almost surely have $\mathbb{P}(\ell_i > 0 \mid \mathcal{F}_1^{(i-1)}) > p$.*

Then, there exists $q(\eta, p) \in (0, 1)$ such that

$$\mathbb{P}\left(\exists i \in \mathbb{N}, k < \tau_i : X_k^{(i)} \geq \sum_{j=0}^i \ell_j\right) \leq q^{\ell_0}.$$

Proof. Denote $E_{j,L} = \{\forall k < \tau_j : X_k^{(j)} < L\}$ and $F_{j,L} = \{\ell_j = L\}$. For $\{L_i\} = \{L_i\}_{i \in \mathbb{N}}$, a sequence of non-negative integers, denote $\bar{L}_j = \sum_{i=0}^j L_i$, $\bar{E}_{n,\{L_i\}} = \bigcap_{i=0}^n E_{i,\bar{L}_i}$ and $\bar{F}_{n,\{L_i\}} = \bigcap_{i=0}^n F_{i,L_i}$. We compute:

$$\mathbb{P}\left(\forall i \in \mathbb{N}, k < \tau_i : X_k^{(i)} < \sum_{j=0}^i \ell_j\right) = \sum_{\{L_i\}_{i \in \mathbb{N}}} \prod_{j=0}^{\infty} \mathbb{P}(F_{j,L_j} \mid \bar{F}_{j-1,\{L_i\}}, \bar{E}_{j-1,\{L_i\}}) \mathbb{P}(E_{j,L_j} \mid \bar{F}_{j,\{L_i\}}, \bar{E}_{j-1,\{L_i\}}) \quad (10)$$

We now bound from below each term in this product. Observe that, for each j , the process $(X^{(j)} \mid \mathcal{F}_0^{(j)})$ satisfies the conditions of Lemma 3.3 almost surely, so that there exists $w = w(\eta) \in (0, 1)$ such that

$$\mathbb{P}(E_{j,L_j} \mid \bar{F}_{j,\{L_i\}}, \bar{E}_{j-1,\{L_i\}}) \geq 1 - w^L \quad (11)$$

for each $j, L \in \mathbb{N}$. Let $\{B_i\}_{i \in \mathbb{N}}$ be a sequence of i.i.d. Bernoulli(p) random variables independent from everything else and write $\bar{B}_j = \sum_{i=1}^j B_i$. By our assumption, $\mathbb{P}(\ell_i > 0 \mid \mathcal{F}_1^{(i-1)}) \geq \mathbb{P}(B_j = 1)$ almost surely. Thus, for any given sequence of numbers $\{L_i\}_{i \in \mathbb{N}}$ we have:

$$\mathbb{P}(F_{j, L_j} \mid \bar{F}_{j-1, \{L_i\}}, \bar{E}_{j-1, \{L_i\}}) \geq \mathbb{P}(B_j = \mathbb{I}(L_i > 0)). \quad (12)$$

Using (11) and the stochastic domination provided by (12), we obtain a bound on the RHS of (10) in terms of the random sequence $\{B_i\}$:

$$\begin{aligned} 1 - \mathbb{P}\left(\exists i \in \mathbb{N}, k < \tau_i : X_k^{(i)} \geq \sum_{j=0}^i \ell_j\right) &\geq \sum_{\{b_i\} \in \{0,1\}^{\mathbb{N}}} \prod_{j=0}^{\infty} \mathbb{P}(B_j = b_i) (1 - w^{\bar{B}_j}) \\ &\geq \sum_{i=\ell_0}^{\infty} \sum_{n=1}^{\infty} p(1-p)^{n-1} (1-w^i)^n \\ &= p \sum_{i=\ell_0}^{\infty} \sum_{n=0}^{\infty} (1-p)^n (1-w^i)^n \\ &= \frac{p}{p + w_0^{\ell_0} - pw^{\ell_0}} = 1 - \frac{w^{\ell_0} - pw^{\ell_0}}{p + w^{\ell_0} - pw^{\ell_0}} \\ &\geq 1 - (1-p) \min\left(\frac{w^{\ell_0}}{p}, 1\right). \end{aligned}$$

Finally we let s be such that $w^s = p^2$ and observe that

$$(1-p) \min\left(\frac{w^{\ell_0}}{p}, 1\right) \leq \max\left(\sqrt{w}, \sqrt[s]{1-p}\right)^{\ell_0}.$$

Setting $q = \max(\sqrt{w}, \sqrt[s]{1-p})$, the proposition follows. \square

3.2 The quantile random walks

We have seen that the empirical distribution of opinions depends entirely on the quantile random process m_k . However, this process defies direct analysis as it has irregular steps and high dependence on the past. To tame it, we introduce a pair of transformations of m_k , which we refer to as *the quantile random walks*. These are defined as follows:

$$\begin{aligned} y_k &= rk - |\{j < k : x_j \leq m_j\}|, \\ y_k^+ &= rk - |\{j < k : x_j < m_j\}|. \end{aligned} \quad (13)$$

Observe that

$$\left\{ \begin{array}{l} \mathbb{P}(y_{k+1} = y_k + r) = \mathbb{P}(\min(X, Y) > m_k \mid \frac{X+Y}{2} \geq m_k), \\ \mathbb{P}(y_{k+1} = y_k + r - 1) = \mathbb{P}(\min(X, Y) \leq m_k \mid \frac{X+Y}{2} \geq m_k), \\ \mathbb{P}(y_{k+1}^+ = y_k^+ + r) = \mathbb{P}(\min(X, Y) \geq m_k \mid \frac{X+Y}{2} \geq m_k), \\ \mathbb{P}(y_{k+1}^+ = y_k^+ + r - 1) = \mathbb{P}(\min(X, Y) < m_k \mid \frac{X+Y}{2} \geq m_k). \end{array} \right. \quad (14)$$

The drifts of the walks y, y^+ are

$$\begin{aligned} \rho(x) &= \mathbb{E}[(y_{k+1} - y_k) \mid m_k = x, \mathcal{F}_{k-1}], \\ \rho^+(x) &= \mathbb{E}[(y_{k+1}^+ - y_k^+) \mid m_k = x, \mathcal{F}_{k-1}], \end{aligned} \quad (15)$$

where ρ and ρ^+ are those given in (1) and (3), respectively. We remark that the need for introducing the $+$ variant of each notation arises from the fact that our theorems are stated in full generality, allowing μ to have an atomic part. In case that μ is a continuous measure – both notions almost surely coincide.

We further denote

$$\begin{aligned}\psi_k(x) &= |\{j < k : x_j < x\}| - |\{j < k : x_j < m_j\}|, \\ \psi_k^+(x) &= |\{j < k : x_j \leq x\}| - |\{j < k : x_j \leq m_j\}|.\end{aligned}$$

Observe that, by the definition of the empirical quantile m_k in (2),

$$m_k \leq x \iff y_k \leq \psi_k^+(x), \quad \text{and} \quad m_k \geq x \iff y_k^+ > \psi_k(x). \quad (16)$$

Also notice that

$$\begin{cases} \psi_{k+1}(x) = \psi_k(x) + \mathbb{I}(x_k < x) - \mathbb{I}(x_k < m_k), \\ \psi_{k+1}^+(x) = \psi_k^+(x) + \mathbb{I}(x_k \leq x) - \mathbb{I}(x_k \leq m_k), \end{cases} \quad (17)$$

so that

$$m_k \geq x \Rightarrow \psi_{k+1}(x) \leq \psi_k(x), \quad m_k \leq x \Rightarrow \psi_{k+1}^+(x) \geq \psi_k^+(x), \quad (18)$$

and for any $a < b$ both $(\psi_k^+(b) - \psi_k^+(a))$ and $(\psi_k(b) - \psi_k(a))$ are monotone increasing. The evolution of m_k , ψ_k and y_k for a particular sequence of members $\{x_k\}_{k=0}^5$ is illustrated in figure 1.

For given $a < b$, write

$$I_k^{a,b} := (\psi_k(a) - \Delta_k, \psi_k^+(b)], \quad J_k^{a,b} := I_k^{a,b} + \Delta_k, \quad \Delta_k := y_k^+ - y_k \geq 0. \quad (19)$$

In what follows, we fix such $a < b$ and write $I_k = I_k^{a,b}$ and $J_k = J_k^{a,b}$ for short. We turn to prove a few short claims concerning these sets.

Claim 3.5. $y_k^+ \in J_k \iff y_k \in I_k \iff a \leq m_k \leq b$.

Proof. Using (16), we have $y_k^+ \leq \psi_k^+(b) + \Delta_k \iff y_k \leq \psi_k^+(b) \iff m_k \leq b$. Similarly, we have $y_k > \psi_k(a) - \Delta_k \iff y_k^+ > \psi_k(a) \iff m_k \geq a$. \square

Claim 3.6. $|I_{k+1}| - |I_k| = |J_{k+1}| - |J_k| = \mathbb{I}\{a \leq x_k \leq b\}$.

In particular $|I_k|$ and $|J_k|$ form the same integer-valued non-decreasing sequence, started at $|I_0| = 0$.

Proof. By (17), we have:

$$\begin{aligned}|I_{k+1}| - |I_k| &= \psi_{k+1}^+(b) - \psi_k^+(b) - (\psi_{k+1}(a) - \psi_k(a)) + \Delta_{k+1} - \Delta_k \\ &= \mathbb{I}\{x_k \leq b\} - \mathbb{I}\{x_k \leq m_k\} - (\mathbb{I}\{x_k < a\} - \mathbb{I}\{x_k < m_k\}) \\ &\quad + \mathbb{I}\{x_k \leq m_k\} - \mathbb{I}\{x_k < m_k\} \\ &= \mathbb{I}\{x_k \leq b\} - \mathbb{I}\{x_k < a\} = \mathbb{I}\{a \leq x_k \leq b\}.\end{aligned}$$

Which establishes the claim for I_k , and hence for J_k , as, by (19), $J_k = I_k + \Delta$. \square

Proposition 3.7. Suppose that $\mu([a, b]) > 0$. Then, almost surely, either $m_k > b$ for all sufficiently large k , or there exists $\ell \in \mathbb{N}$ such that $x_\ell \in [a, b]$.

$r = \frac{1}{3}$	
$x_0 = 0$	$m_0 = 0$
$x_1 = 5$	$m_1 = 0$
$x_2 = 2$	$m_2 = 0$
$x_3 = 3$	$m_3 = 2$
$x_4 = 2$	$m_4 = 2$
$x_5 = 1$	$m_5 = 1$

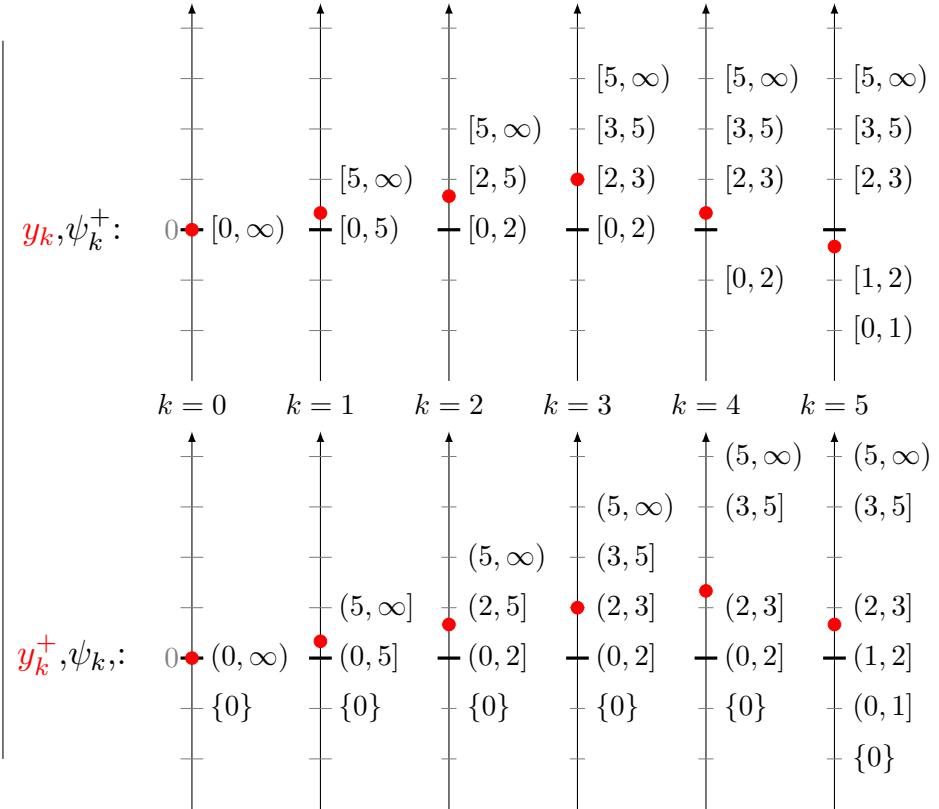


Figure 1: The evolution of ψ_k, ψ_k^+, y_k and y_k^+ over the first five steps of a sample of the process x_k . In this particular example $r = \frac{1}{3}$, and the first six admitted members hold opinions 0, 5, 2, 3, 2 and 1. The locations of y_k (above) and y_k^+ (below) are depicted by a small disk, and the intervals which are mapped by ψ_k^+ (above) and ψ_k (below) to each integer are written next to it. Observe that m_k could be computed by rounding down either y_k (or y_k^+) to the nearest integer, then rounding up (down) to the nearest element of the image, and then looking on the infimum of the corresponding preimage interval (in accordance with (16)). Also observe how y_k and y_k^+ evolve in exactly the same way as long as $x_k \neq m_k$ (see (13)), and ψ_k^+, ψ_k evolve in the same way except interval endpoints, as long as the admitted members have distinct opinions (see (17)).

Proof. We assume $\mathbb{P}(m_k \leq a) \geq \mathbb{P}(x_0, x_1, \dots, x_{k-1} \leq a) > 0$, as otherwise the claim is straightforward. Hence,

$$\begin{aligned}\mathbb{P}(x_k \in [a, b] \mid \mathcal{F}_{k-1}, m_k \leq a) &\geq \mathbb{P}(\min(X, Y) \in [a, b] \mid \frac{X+Y}{2} \geq a) \\ &\geq \mathbb{P}(X, Y \in [a, b]) = \mu([a, b])^2 > 0.\end{aligned}\quad (20)$$

By the Borel-Cantelli lemma, this implies that almost surely one of the following holds:

- there exists ℓ such that $x_\ell \in [a, b]$.
- $x_k \notin [a, b]$ for all k and there exists ℓ_0 such that for all $\ell > \ell_0$ we have $m_\ell > a$.

From Claim 3.6 it follows that in the former case $|I_\ell| = 1$ for some ℓ , while in the latter case $m_k > b$ for all $k > \ell_0$. The claim follows. \square

In the following proposition we capture the idea that the probability that m_k crosses an interval of negative drift decays exponentially in the number of club members whose opinion lies in that interval. Recall the notation $\nu_k(I) = |\{j \leq k : x_j \in I\}|$.

Proposition 3.8. *Let $[a, b]$ be an interval of positive measure- μ such that $\rho(x) < -\varepsilon < 0$ for all $x \in [a, b]$. Then there exists $q = q\left(\varepsilon, \frac{\mu([a, b])}{\mu([a, \infty))}\right) \in (0, 1)$ such that for all $\ell, t_0 \in \mathbb{N}$:*

$$\mathbb{P}\left(\exists k > t_0 : m_k \geq b \mid \mathcal{F}_{t_0-1}, m_{t_0} \leq a, \nu_{t_0-1}([a, b]) \geq \ell\right) < q^\ell. \quad (21)$$

Proof. Denote $I_k = I_k^{a,b}$ as in (19). In light of (16), Claim 3.5 and Claim 3.6 it would suffice to show that, for all $\ell, t_0 \in \mathbb{N}$,

$$\mathbb{P}\left(\exists k > t_0 : y_k > \psi_k^+(b) \mid \mathcal{F}_{t_0-1}, y_{t_0} < I_{t_0}, |I_{t_0}| \geq \ell\right) < q^\ell. \quad (22)$$

Define inductively for $i > 0$

$$\begin{aligned}t_i &= \min\{k > t_{i-1} : m_{k-1} < a, m_k \geq a\}, \\ s_i &= \min\{k \geq t_i : a \leq m_k \leq b, m_{k+1} \notin [a, b]\},\end{aligned}$$

By Claim 3.5, t_i are times in which the walk y_k entered the interval I_k from below while s_i are times in which y_k is about to exit the interval in its next step. Denote $p := \frac{\mathbb{P}(X \in [a, b])}{\mathbb{P}(X \in [a, \infty))}$ and observe that $p > 0$. By Claim 3.6 and the definition of the quantile process, we almost surely have,

$$\begin{aligned}\mathbb{P}(|I_{t_i}| > |I_{t_{i-1}}| \mid \mathcal{F}_{t_{i-1}}) &\geq \mathbb{P}(x_t \in [a, b] \mid \mathcal{F}_{t-1}, m_{t-1} \leq a, m_t \geq a) \\ &\geq \mathbb{P}(\min(X, Y) \in [a, b] \mid \frac{X+Y}{2} \geq a) \geq p.\end{aligned}\quad (23)$$

Write $i_\infty = \min\{i : t_i = \infty\}$. For each $i < i_\infty$, and $k \in [t_i, s_i]$ write $X_k^{(i)} = y_k - y_{t_i}$ and observe that this is a random walk started at $X_0^{(i)} = 0$ and satisfying $X_{k+1}^{(i)} - X_k^{(i)} = y_{k+1} - y_k$ for $k \in [t_i, s_i]$. By our premise and (15) we have $\mathbb{E}(X_{k+1}^{(i)} \mid X_0^{(i)}, \dots, X_k^{(i)}, \mathcal{F}_{t_{i-1}}) \leq X_k^{(i)} - \varepsilon$ for all $k \in \mathbb{N}_0$. Further observe that $\{X_j^{(i)}\}_{j \in \mathbb{N}, i \in [k]}$ are \mathcal{F}_{s_i} measurable. By (14) we have $|X_{k+1}^{(i)} - X_k^{(i)}| \leq 1$ for any i and k . Denote $\ell_j = |I_{t_j}| - |I_{t_{j-1}}|$ for $j \geq 1$ and $\ell_0 = |I_{t_0}|$. Consequently, for $i \in \mathbb{N}_0$ we have $L_i := \sum_{j \leq i} \ell_j = |I_{t_i}|$. By Claim 3.6, $\ell_j \geq 0$ for all j and by (23) we have $\mathbb{P}(\ell_j > 0 \mid \mathcal{F}_{t_{j-1}}) > p$ for all j . Since ℓ_j is F_{t_j-1}

measurable, and since $t_{i+1} \geq s_i + 1$ we conclude that the random processes $\{X_k^{(i)}\}$ and the process $\{\ell_j\}$ obey the conditions of Lemma 3.4 with $\eta = \varepsilon$, $\mathcal{F}_0^{(i)} = \mathcal{F}_{t_{i-1}}$ and $\mathcal{F}_1^{(i)} = \mathcal{F}_{s_i}$. We deduce that

$$\mathbb{P}\left(\exists i, k \in \mathbb{N} \text{ s.t. } X_k^{(i)} \geq L_i\right) \leq q^\ell,$$

for some $q \in (0, 1)$ which depends only on ε and p .

In order to obtain (22) it is enough to show that

$$\forall k \in [t_i, s_i] : \quad \{X_k^{(i)} \geq L_i\} \supseteq \{y_k > \psi_k^+(b)\}. \quad (24)$$

Using the fact that $\{L_i\}$ is integer-valued together with (19), we have:

$$\{X_k^{(i)} \geq L_i\} = \{X_k^{(i)} > L_i - 1\} = \{y_k > y_{t_i} + |I_{t_i}| - 1\} = \{y_k > \psi_{t_i}^+(b) - \psi_{t_i}(a) + y_{t_i}^+ - 1\},$$

so that in order to obtain (24) it would suffice to show that

$$\psi_k^+(b) \geq \psi_{t_i}^+(b) - \psi_{t_i}(a) + y_{t_i}^+ - 1. \quad (25)$$

Recalling that $m_j \leq b$ for all $j \in [t_i, s_i]$ and using (18) we have $\psi_{j+1}^+(b) \geq \psi_j^+(b)$ for all such j , and thus

$$\psi_k^+(b) \geq \psi_{t_i}^+(b). \quad (26)$$

Next, recalling the definition of t_i , we have $m_{t_i-1} < a$, so that by (16) we obtain $y_{t_i-1}^+ \leq \psi_{t_i-1}(a)$. By (2) we have $x_{t_i-1} > a > m_{t_i-1}$ and hence, by (17), we deduce that $\psi_{t_i}(a) = \psi_{t_i-1}(a)$. Putting these observations together we obtain

$$y_{t_i}^+ = y_{t_i-1}^+ + r \leq \psi_{t_i-1}(a) + r = \psi_{t_i}(a) + r < \psi_{t_i}(a) + 1. \quad (27)$$

Inequalities (26) and (27) imply (25), concluding the proof. \square

Proposition 3.9. *Let $[b, a]$ be an interval with $\mu([b, a]) > 0$, such that $\rho^+(x) > \varepsilon > 0$ for all $x \in [b, a]$. Then there exists $q = q(\varepsilon, \mu([b, a]))$ such that for all $\ell, t_0 \in \mathbb{N}_0$:*

$$\mathbb{P}\left(\exists k > t_0 : m_k \leq b \mid \mathcal{F}_{t_0-1}, m_{t_0} \geq a, \nu_{t_0-1}([b, a]) \geq \ell\right) < q^\ell. \quad (28)$$

Proof. Under appropriate changes, such as replacing ρ with ρ^+ and the intervals I_k with J_k (defined in (19)), most of the proof is identical to that of Proposition 3.8. The only significant difference is in the proof of a counterpart of (23), namely:

$$\mathbb{P}(|J_t| > |J_{t-1}| \mid \mathcal{F}_{t-1}, m_{t-1} > a, m_t \leq a) \geq \frac{\mu([b, a])}{\mu((-\infty, a])} \geq \mu([b, a]). \quad (29)$$

Thus, we provide here only the proof of (29). To this end, we observe, using (2), that the event $\{m_{t-1} > a, m_t \leq a\}$ occurs if the median of x_0, \dots, x_{t-2} is above a (i.e. $\nu_{t-2}((-\infty, a]) \geq r(t-1)$), the next accepted member's opinion lies in $(-\infty, a]$ (i.e. $x_{t-1} \leq a$), and this causes the new median m_t to be less than a (i.e. $\nu_{t-2}((-\infty, a]) + 1 < rt$). We thus have:

$$\begin{aligned} & \mathbb{P}(|J_t| > |J_{t-1}| \mid \mathcal{F}_{t-1}, m_{t-1} > a, m_t \leq a) \\ &= \mathbb{P}(x_{t-1} \in [b, a] \mid \mathcal{F}_{t-1}, m_{t-1} > a, m_t \leq a) \\ &= \mathbb{P}(x_{t-1} \in [b, a] \mid m_{t-1}, r(t-1) \leq \nu_{t-2}((-\infty, a]) < rt - 1, x_{t-1} \leq a), \\ &= \mathbb{P}(\min(X, Y) \in [b, a] \mid m_{t-1}, m_{t-1} > a, \frac{X+Y}{2} \geq m_{t-1}, \min(X, Y) \leq a) \end{aligned}$$

The first equality follows from Claim 3.6. In the last equality we have used the fact that the distribution of x_{t-1} is equal to that of $\min(X, Y)$ conditioned on $\frac{X+Y}{2} \geq m_{t-1}$, where X and Y are i.i.d. μ -distributed random variables. We then observe that for any fixed values $m > a$ and $y \geq m$ we have

$$\begin{aligned} \mathbb{P}(\min(X, Y) \in [b, a] \mid m_{t-1} = m > a, \frac{X+Y}{2} \geq m, \min(X, Y) \in [0, a], Y = y) \\ = \mathbb{P}(X \in [b, a] \mid X \in [2m - y, a]) \\ \geq \mathbb{P}(X \in [b, a] \mid X \in [0, a]) = \frac{\mu([b, a])}{\mu((-\infty, a])}. \end{aligned}$$

Therefore (29) holds, as required. \square

We conclude this part with the following simple observation about the continuity of the process m_k , following directly from (2) and from the fact that members are admitted one-by-one.

Observation 3.10. *Let $i < j < \ell$ and suppose that $m_j < x_i < m_\ell$. Then there exists $k \in (j, \ell)$ such that $m_k = x_i$.*

3.3 A property of continuous measures

We shall use the following observation about continuous measures.

Lemma 3.11. *Let μ be probability measure with no atoms in $[a, b]$ which satisfies $\mu([a, b]) > 0$. Then there exist two intervals $I_1, I_2 \subseteq [a, b]$, both of positive measure, such that $\frac{I_1 + I_2}{2} > I_1$.*

Proof. Let $a' = \inf\{d \geq a : \forall \epsilon > 0 \mu((d, d + \epsilon)) > 0\}$ and $b' = \sup\{d \leq b : \forall \epsilon > 0 \mu((d - \epsilon, d)) > 0\}$. Since $\mu([a, b]) > 0$ these infimum and supremum are taken over non-empty sets and are therefore finite. By the continuity of μ on $[a, b]$ we have $a' < b'$. Fix $\delta = \frac{b' - a'}{4}$ and set $I_1 = (a', a' + \delta)$ and $I_2 = (b' - \delta, b')$. Then $\mu(I_j) > 0$ for $j = 1, 2$ and $\inf(I_1) + \inf(I_2) = a' + b' - \delta = 2(a' + \delta) + \frac{(b' - a')}{4} > 2(a' + \delta) \geq 2 \sup(I_1)$. \square

4 Proofs of key lemmata

4.1 Proof of Lemma 2.10: negative drift

Proof outline. The proofs of the three parts of the lemma are rather similar. All three rely on the observation that to the left of every point at which $\rho(m_k)$, the drift of y_k , is negative, one can identify an interval I of positive measure where the drift is bounded away from zero from below uniformly. Our purpose is to use this property to show that if m_k is below I for sufficiently many steps, then it will never leave it from above. This is shown in Lemma 4.1 below. To show Lemma 4.1, we divide the process m_k into excursions into I entering it from below. We show that the probability of m_k ever exiting I from above in a given excursion is less than 1, and that it decays exponentially in the number of club members whose opinion lies in I at the start of the excursion (using Proposition 3.8). We further show that m_k will exit I from below after sufficiently many visits to I (using Lemma 3.3 on drifting random walks, and our choice of I), and that each time that a new excursion starts, there is a positive probability, bounded away from zero, that a new member with opinion in I will be admitted to the club. We then argue that the probability of ever exiting I from above decays to 0 exponentially fast as

the number of members in I accumulates, so that, by the Borel-Cantelli Lemma, the probability that infinitely many such crossings will occur must be zero.

Our key auxiliary lemma is the following.

Lemma 4.1. *let $\varepsilon > 0$, $\alpha \leq \beta \leq \gamma$ and $I \in \{(\beta, \gamma], (\beta, \gamma)\}$, such that $\mu([\alpha, \beta]) > 0$, $\mu((\alpha, \beta] \cup I) \leq \frac{\varepsilon}{2}$ and $\rho(x) \leq -\varepsilon$ for all $x \in [\alpha, \beta] \cup I$. Then γ is a right-barrier, and if $I \neq \emptyset$ then I is a right-separator.*

We present the proof of Lemma 2.10 and then turn to prove Lemma 4.1.

Proof of Lemma 2.10. **Part (a).** Let x_0 be such that $\rho^+(x_0) < 0$ and write $\varepsilon = -\frac{\rho^+(x_0)}{2}$. Let

$$a := \inf \left\{ x : \mu(x, x_0) \leq \frac{\varepsilon}{2}, \rho^+(x) \leq -\varepsilon \right\}.$$

Using Observations 2.3 and 2.4, we deduce that $a > -\infty$ and $\mu([a, x_0)) > 0$. Using the fact that $\rho \leq \rho^+$ (Observation 2.1), we deduce that $\rho(x) \leq -\varepsilon$ for all $x \in [a, x_0)$. As $\mu([a, x_0)) > 0$, there exists $b \in [a, x_0)$ such that $\mu([a, b]) > 0$. Hence $a \leq b < x_0$ and ε satisfy the conditions of Lemma 4.1. We conclude that $(b, x_0]$ is a right-separator, as required.

Part (b). Let x_0 be such that $\rho(x_0) < 0$. If $\rho^+(x_0) < 0$ then this part follows from part (a). Otherwise, we have $\rho^+(x_0) \geq 0$ so that, by Observation 2.1, we have $\mu(\{x_0\}) > 0$. Hence the conditions of Lemma 4.1 are satisfied with $\alpha = \beta = \gamma = x_0$ and $\varepsilon = -\rho(x_0)$. We conclude that x_0 is a right-barrier, as required.

Part (c). Let x_0 be such that $\lim_{x \rightarrow x_0+} \rho(x) < 0$ and write $\varepsilon = -\lim_{x \rightarrow x_0+} \rho(x)/2$. Write

$$\begin{aligned} b &= \sup \left\{ x : \mu(x_0, x) \leq \frac{\varepsilon}{4}, \rho(x) \leq -\varepsilon \right\}, \\ a &= \inf \left\{ x : \mu(x, b) \leq \frac{\varepsilon}{2}, \rho(x) \leq -\varepsilon \right\}. \end{aligned}$$

Observe that $a \leq x_0 < b$. By lower-semicontinuity of ρ (Observation 2.1), we have $\rho(x) \leq -\varepsilon$ for all $x \in [a, b]$. By Observation 2.3, we see that $\mu([a, x_0)) > 0$. Hence the conditions of Lemma 4.1 are satisfied with $\alpha = a$, $\beta = x_0$, $\gamma = b$ and ε . We deduce that $(x_0, b]$ is a right-separator, so that, by definition, (x_0, b) is also a right-separator – as required. \square

Proof of Lemma 4.1. Denote the events $E = \{m_k > I \text{ a.e.}\}$ and $T = \{m_k < I \text{ a.e.}\}$ (here E is associated with *escaping* while T is associated with being *trapped*). Our purpose is to show that $\mathbb{P}(E \cup T) = 1$.

We define a sequence of discrete stopping times $\{t_i\}_{i \in \mathbb{N}_0}$, setting $t_0 = \inf\{t \in \mathbb{N} : x_t \in [\alpha, \beta]\}$ and defining for $i > 0$,

$$t_i = \begin{cases} \inf\{t > t_{i-1} : m_t \leq \alpha, m_{t-1} > \alpha\} & t_{i-1} < \infty, \\ \infty & t_{i-1} = \infty, \end{cases}$$

these are subsequent entry times of m_k into the set $(-\infty, \alpha]$. From Proposition 3.7 we obtain that, almost surely, $\{t_0 < \infty\} \cup E$ holds.

Next, we use induction to show that

$$\mathbb{P}(\{\forall i \in \mathbb{N}_0 : t_i < \infty\} \cup E \cup T) = 1. \quad (30)$$

Indeed, assume that $t_i < \infty$ and let us show that either $t_{i+1} < \infty$, E or T holds. To this end, define a sequence of stopping times $\{s_j^i\}_{j \in \mathbb{N}_0}$ by setting $s_0^i = \inf\{s > t_i : m_s \in (\alpha, \beta] \cup I\}$ and

$$s_j^i = \begin{cases} \inf\{s_{j-1}^i < s < t_{i+1} : m_s \in (\alpha, \beta] \cup I, m_{s-1} \notin (\alpha, \beta] \cup I\} & s_{j-1}^i < \infty, \\ \infty & s_{j-1}^i = \infty. \end{cases}$$

These are subsequent entry times of m_k to the set $(\alpha, \beta] \cup I$ between times t_i and t_{i+1} . To show (30), it would suffice to show that $\mathbb{P}(\{\forall j \in \mathbb{N}_0 : s_j^i < \infty\}) = 0$. Define $Z_k := y_k - \psi_k^+(\alpha)$ and observe that, by (16), $t_{i+1} = \inf\{k > s_0^i : Z_k \leq 0\}$. Writing $A_k = \{m_k \in (\alpha, \beta] \cup I\}$, we compute

$$\begin{aligned}\mathbb{E}(Z_{k+1} - Z_k \mid \mathcal{F}_{k-1}, A_k) &= \mathbb{E}(y_{k+1} - y_k \mid \mathcal{F}_{k-1}, A_k) - \mathbb{E}(\psi_{k+1}^+(\alpha) - \psi_k^+(\alpha) \mid \mathcal{F}_{k-1}, A_k) \\ &\leq -\varepsilon + \mathbb{P}(x_k \leq m_k \mid \mathcal{F}_{k-1}, A_k) - \mathbb{P}(x_k \leq \alpha \mid \mathcal{F}_{k-1}, A_k) \\ &\leq -\varepsilon + \mathbb{P}(x_k \in (\alpha, \beta] \cup I \mid \mathcal{F}_{k-1}, A_k) \leq -\varepsilon + \mu((\alpha, \beta] \cup I) \leq -\frac{\varepsilon}{2},\end{aligned}$$

where the transition between the first line and the second uses (15), (17) and our assumption on ρ and the last inequality uses our assumption that $\mu((\alpha, \beta] \cup I) \leq \frac{\varepsilon}{2}$. Write

$$\tau_j^i = \inf\{k > s_j^i : m_k > \gamma\} = \inf\{k > s_j^i : Z_k > \psi_k^+(\gamma) - \psi_k^+(\alpha)\}$$

and observe that $\{t_{i+1} < \tau_j^i\} \subseteq \{s_{j+1}^i = \infty\}$. Notice that the process $\{Z_k : k \in [s_j^i, \min(\tau_j^i, t_{i+1})]\}$ has bounded steps and drift at most $-\varepsilon$, so by applying Lemma 3.3, there exists $q > 0$ such that

$$\begin{aligned}\mathbb{P}(s_{j+1}^i = \infty \mid \mathcal{F}_{s_j^i-1}, s_j^i < \infty) &\geq \mathbb{P}(t_{i+1} < \tau_j^i \mid \mathcal{F}_{s_j^i-1}, s_j^i < \infty) \\ &\geq \mathbb{P}(T_{\{Z_k \geq 1\}} \mid Z_0 \leq 0) > q.\end{aligned}$$

Thus, for all $j \in \mathbb{N}_0$ such that $\mathbb{P}(s_j^i < \infty) > 0$ we have $\mathbb{P}(s_{j+1}^i < \infty \mid s_j^i < \infty) < 1 - q$. Using the conditional second Borel-Cantelli lemma [5] we obtain $\mathbb{P}(\{\forall j \in \mathbb{N}_0 : s_j^i < \infty\}) = 0$, and establish (30).

Next, applying Proposition 3.8 with $[a, b] := [\alpha, \beta]$, $t_0 := t_i$, and ε , we obtain the existence of $q < 1$ such that for all $i \in \mathbb{N}_0$,

$$\mathbb{P}(T^c \mid \mathcal{F}_{t_i-1}, t_i < \infty) < \mathbb{P}(\exists k > t_i : m_k \geq \beta \mid \mathcal{F}_{t_i-1}, m_{t_i} \leq \alpha, \nu_{t_i-1}([\alpha, \beta]) \geq 1) < q. \quad (31)$$

Denote $L_j = |\{i < j : \exists t \in (t_i, t_{i+1}), m_t > \beta\}|$ for $j \in \mathbb{N} \cup \{\infty\}$. By (31) for all $i \in \mathbb{N}$ we obtain, $\mathbb{P}(L_\infty > L_i \mid \mathcal{F}_{t_i-1}, t_i < \infty) < q$. Hence $\mathbb{P}(\{L_\infty > k\} \cap \{\forall i \in \mathbb{N}_0 : t_i < \infty\}) < q^k$ for any $k \in \mathbb{N}$, so that $\mathbb{P}(T^c \cap \{\forall i \in \mathbb{N}_0 : t_i < \infty\}) = 0$. Combined with (30) it follows that $\mathbb{P}(E \cup T) = 1$, so that by definition, if $I \neq \emptyset$ then $(\beta, \gamma]$ is a separator, while if $I = \emptyset$, then γ is a barrier. Moreover, from (31) we obtain $\mathbb{P}(T) > 0$, so that if $I \neq \emptyset$, then $(\beta, \gamma]$ is a right-separator, while if $I = \emptyset$, then γ is a right-barrier. \square

4.2 Proof of Lemma 2.11: positive drift

The proof is analogous to that of Lemma 2.10. Firstly, we introduce a counterpart of Lemma 4.1.

Lemma 4.2. *let $\varepsilon > 0$ and $\gamma \leq \beta \leq \alpha$ be such that $\mu((\gamma, \alpha)) \leq \frac{\varepsilon}{2}$, $\mu([\beta, \alpha]) > 0$ and $\rho^+(x) \geq \varepsilon$ for all $x \in I \cup [\beta, \alpha]$ for $I \in \{[\gamma, \beta], (\gamma, \beta)\}$. Then $\mathbb{P}(m_k > I \text{ a.e.}) > 0$ and if $I \neq \emptyset$ then I is almost surely a separator, while if $I = \emptyset$ then, almost surely, either $m_k \geq \gamma$ almost everywhere or $m_k < \gamma$ almost everywhere.*

Mutatis mutandis, the proof of Lemma 4.2 is the same as that of Lemma 4.1. A few noteworthy alterations are the following. Define $T = \{m_k > I \text{ a.e.}\}$ and $E = \{m_k < I \text{ a.e.}\}$ and $Z_k := \psi_k(\alpha) - y_k^+$. In addition, Proposition 3.9 plays in the proof the role previously given to Proposition 3.8.

Proof of Lemma 2.11. Part (a). Let x_0 be such that $\rho(x_0) > 0$ and denote $\varepsilon := \frac{\rho(x_0)}{2}$. Write

$$a = \sup \left\{ x : \mu(x_0, x) \leq \frac{\varepsilon}{2}, \rho(x) \geq \varepsilon \right\}. \quad (32)$$

Using Observation 2.3 and the fact that $x_0 < \mu_{\max}$ we deduce that $\mu((x_0, a]) > 0$. Using the fact that $\rho \leq \rho^+$ (Observation 2.1) and that $\rho^+(x) = \lim_{t \rightarrow x^-} \rho(t)$, we deduce that $\rho^+(x) \geq \varepsilon$ for all $x \in [c, a]$ for some $c < x_0$. As $\mu((x_0, a]) > 0$, we can choose $b \in (x_0, a]$ such that $\mu([b, a]) > 0$. Hence $c < b \leq a$ and ε satisfy the conditions of Lemma 4.2. We conclude that (c, b) is a left-separator containing x_0 , as required.

Part (b). Let x_0 be such that $\rho^+(x_0) > 0$. If $\rho(x_0) > 0$ then this part follows from part (a). Otherwise, we have $\rho(x_0) \leq 0$. If $\rho(x_0) < 0$, then by Lemma 2.10b, x_0 is a right-barrier and we are done. In the remaining $\rho(x_0) = 0$ case, by Observation 2.1, we have $\mu(\{x_0\}) > 0$. Hence the conditions of Lemma 4.2 are satisfied with $\alpha = \beta = \gamma = x_0$ and $\varepsilon = \rho^+(x_0)$. We conclude that either $m_k \geq x_0$ almost everywhere, or $m_k < x_0$ almost everywhere. By Observation 2.2, we have $\lim_{x \rightarrow x_0^+} \rho(x) > 0$. Hence there exists $b > x_0$ such that $\rho^+(x) \geq \varepsilon > 0$ for all $x \in [x_0, b]$ and some $\varepsilon > 0$. By Lemma 4.2, $(x_0, b]$ is a separator and hence x_0 is a barrier.

Part (c). Let x_0 be such that $\lim_{x \rightarrow x_0^+} \rho(x) > 0$ and denote $\varepsilon := \lim_{x \rightarrow x_0^+} \rho(x)/2$. We repeat the arguments of Part a, setting a as in (32) and $b \in (x_0, a]$ so that $\mu([b, a]) > 0$. This way we obtain that $\rho^+(x) \geq \varepsilon$ for all $x \in (x_0, a]$. Hence $x_0 < b \leq a$ and ε satisfy the conditions of Lemma 4.2 and we deduce that, almost surely, $(x_0, b]$ is a separator, as required. \square

4.3 Proof of Lemma 2.12: zero drift

Proof outline. Our purpose is to show that if a process visits a positive measure interval I with $\rho(I) = 0$, sufficiently many times, then it will eventually find itself entrapped within. To do so we identify inside I a sub-interval I_1 , which we in turn sub-divide into a countable sequence of nested shells with outer shells thinner than inner ones. We then show that after every visit of m_k to the innermost shells, there is a probability bounded away from zero that it will never leave the interval. To do so we construct an event under which this happens, and bound its probability from below. On this event, by the time at which the walk breaks out from the i -th shell, many elements are added to the $i+1$ -th shell, so that it takes significantly more time for the walk to cross it. By showing that this sequence of times converges to infinity we conclude that the walk never breaks out of I_1 .

Let $I = [a, b]$ be such that $\mu(I) > 0$ and $\rho^+(x) = \rho(x) = 0$ for all $x \in I$. By Observation 2.1 this implies that μ has no atoms in $[a, b]$. Hence, by Lemma 3.11, there exist intervals $I_1, I_2 \subseteq I$, such that $\mu(I_j) > 0$ ($j = 1, 2$) and $\frac{I_1 + I_2}{2} > I_1$. We further divide the interval $I_1 = [\alpha, \beta]$ as follows. Let $c \in I_1$ be the minimal point satisfying $\mu([\alpha, c]) = \mu([c, \beta])$. Define two sequences $c = r_0 < r_1 < r_2 < \dots < \beta$ and $c = \ell_0 > \ell_1 > \ell_2 > \dots > \alpha$ such that $\mu([r_j, r_{j+1}]) = \mu([\ell_{j+1}, \ell_j]) = 2^{-(j+2)}\mu(I_1)$. Fix a parameter $n_0 \in \mathbb{N}$. For any $t > 0$ and $j \in \mathbb{N}$ let

$$N_j(t) = \min \left(\nu_{t-1}([\ell_{j+2}, \ell_{j+1}]), \nu_{t-1}([r_{j+1}, r_{j+2}]) \right).$$

Fix $t_0, n_0 \in \mathbb{N}$ and denote the event $B_0 = \{m_{t_0} \in [\ell_1, r_1], N_0(t_0) \geq n_0\}$. Our proof relies on the following lemma, whose proof appears later in this section.

Lemma 4.3. *There exists n_0 such that if $t_0 > \frac{n_0}{r(1-r)} + 5$, then $\mathbb{P}(B_0) > 0$ and*

$$\mathbb{P}(\forall k > t_0 : m_k \in I_1 \mid \mathcal{F}_{t_0-1}, B_0) > \frac{1}{3}.$$

We turn to establish Lemma 2.12.

Proof of Lemma 2.12. The fact that $\mathbb{P}(m_k \in I \text{ a.e.}) > 0$ is immediately implied by Lemma 4.3. We are left with showing that almost surely, I contains a barrier.

Let $F := \{m_k < \ell_1 \text{ i.o.}\} \cap \{m_k > r_1 \text{ i.o.}\}$. If F^c holds then either r_1 or ℓ_1 is a barrier and we are done. Thus it remains to show that

$$\mathbb{P}(F \cap \{I \text{ does not contain a barrier}\}) = 0. \quad (33)$$

If $\mathbb{P}(F) = 0$ we are done. Otherwise, let n_0 be as in Lemma 4.3. For $j \in \mathbb{N}$ define

$$\begin{aligned} s_0 &= \min\{s \in \mathbb{N} : m_s \in [\ell_1, r_1], N_0(s) > n_0\}, \\ s_j &= \min\{s > s_{j-1} : m_s \in [\ell_1, r_1], m_{s-1} \notin [\ell_1, r_1]\}. \end{aligned}$$

Under F , we have $s_j < \infty$ for all j (using Proposition 3.7 with $[a, b] = [\ell_1, r_1]$ and Observation 3.10). By Lemma 4.3 we have $\mathbb{P}(\{\forall k \geq s_j : m_k \in I_1 \mid \mathcal{F}_{s_{j-1}}\} \cap F) > \frac{1}{3}$, for all $j \in \mathbb{N}$. Thus by the conditional Borel-Cantelli Lemma [5], $\mathbb{P}(\{m_k \in I_1 \text{ a.e.}\} \cap F) = 1$, which implies that under F , every point of $I_2 \subset I$ is a barrier, (33) follows. \square

It remains to prove Lemma 4.3. For $j \in \mathbb{N}$, define inductively

$$t_j := \inf\{k > t_{j-1} : m_k \notin [\ell_j, r_j]\}, \quad N_j := \begin{cases} N_j(t_j), & t_j < \infty, \\ \infty, & t_j = \infty. \end{cases}$$

Let $\delta = 2^{-3}\mu(I_1)\mu(I_2)$ and denote (for $j \in \mathbb{N}$) the events:

$$A_j = \{t_{j+1} - t_j \geq N_j^{1.5}\} \cup \{t_j = \infty\}, \quad B_j = \{N_{j+1} \geq \delta 2^{-j} N_j^{1.5}\} \cup \{t_j = \infty\}.$$

Notice that both A_j and B_j are $\mathcal{F}_{t_{j+1}-1}$ -measurable. In proving Lemma 4.3 we rely on three key properties, whose proofs we provide at the end of this section.

Claim 4.4. For each $j \in \mathbb{N}$:

$$\mathbb{P}(A_j^c \mid \mathcal{F}_{t_{j-1}}) \leq e^{-\sqrt{N_j}/2}.$$

Claim 4.5. There exists a constant $C > 0$ such that for all $j \in \mathbb{N}$,

$$\mathbb{P}(B_j \mid \mathcal{F}_{t_{j-1}}, A_j) \geq 1 - 2e^{-C2^{-2j}N_j^{1.5}}.$$

Claim 4.6. For all $Q > 0$ and $t_0 > 0$ there exists $n_0 = n_0(Q)$ such that

$$\bigcap_{j \in \mathbb{N}_0} B_j \subseteq \bigcap_{j \in \mathbb{N}} \{N_j \geq e^{Q(1.1)^j}\}.$$

Proof of Lemma 4.3. First we show that for any n_0 and $t_0 > \frac{n_0}{r(1-r)} + 5$ we have $\mathbb{P}(B_0) > 0$. Let $K_1, K_2, K_3 \subset \mathbb{N}$ be disjoint sets of integers such that $\{0, 1, \dots, t_0 - 1\} = K_1 \cup K_2 \cup K_3$, $|K_1| = \lceil \frac{n_0}{1-r} \rceil$, $|K_2| = \lceil \frac{n_0}{r} \rceil$, and $|K_3| \geq 3$. Let (X_k, Y_k) denote the opinions of the independent candidates at stage k . Define

$$\begin{aligned} G_0 &:= \{\forall k \in K_1 : X_k \in [\ell_2, \ell_1]\} \cap \{\forall k \in K_2 : X_k \in [r_1, r_2]\} \\ &\quad \cap \{\forall k \in K_3 : X_k \in [\ell_1, r_1]\} \cap \{\forall k \leq t_0 : Y_k \in I_2\}. \end{aligned}$$

Under G_0 we have $N_0(t_0) \geq \min\left(\frac{n_0}{r}, \frac{n_0}{1-r}\right) \geq n_0$. Now notice that under G_0 we have

$$\nu_{t_0-1}([\ell_2, \ell_1]) = |K_1| < r(|K_1| + |K_2| + |K_3|) = rt_0 \text{ and } \nu_{t_0-1}([\ell_2, r_1]) = |K_1| + |K_2| > r(|K_1| + |K_2| + |K_3|) = rt_0,$$

which implies that $m_{t_0} \in [\ell_1, r_1]$. We deduce that $G_0 \subseteq B_0$. Since $\{X_k\}$ and $\{Y_k\}$ are all independent it follows that $\mathbb{P}(G_0) \geq \mathbb{P}(B_0) > 0$.

Let $Q > 0$ be a large parameter, to be chosen later. By Claim 4.6, there exists $n_0 = n_0(Q)$ so that

$$\bigcap_{j \in \mathbb{N}_0} B_j \subseteq \bigcap_{j \in \mathbb{N}} \{N_j \geq e^{Q(1.1)^j}\}. \quad (34)$$

Let $t_0 > \frac{n_0}{r(1-r)} + 5$, so that $\mathbb{P}(B_0) > 0$ by the arguments above. Using (34) and the bounds from Claims 4.4 and 4.5 we have:

$$\begin{aligned} \mathbb{P}\left(\forall j : N_j \geq e^{Q(1.1)^j} \mid \mathcal{F}_{t_0-1}, B_0\right) &\geq \mathbb{P}\left(\bigcap_{j \in \mathbb{N}} B_j \bigcap_{j \in \mathbb{N}} A_j \mid \mathcal{F}_{t_0-1}, B_0\right) \\ &= \prod_{j \in \mathbb{N}} \mathbb{P}(B_j \mid \mathcal{F}_{t_0-1}, B_0, \dots, B_{j-1}, A_1, \dots, A_j) \prod_{j \in \mathbb{N}} \mathbb{P}(A_j \mid \mathcal{F}_{t_0-1}, B_0, \dots, B_{j-1}, A_1, \dots, A_{j-1}) \\ &\geq \prod_{j=1}^{\infty} \left(1 - 2e^{-C2^{-2j}N_j^{1.5}}\right) \left(1 - e^{-\sqrt{N_j}/2}\right) \geq \prod_{j=1}^{\infty} \left(1 - 2e^{-C4^{-j}e^{1.5Q(1.1)^j}}\right) \left(1 - e^{-\frac{1}{2}e^{\frac{1}{2}Q(1.1)^j}}\right) \\ &\geq 1 - 2 \sum_{j=1}^{\infty} e^{-C4^{-j}e^{1.5Q(1.1)^j}} - \sum_{j=1}^{\infty} e^{-\frac{1}{2}e^{\frac{1}{2}Q(1.1)^j}} \end{aligned}$$

We may choose Q sufficiently large to have both $Ce^{1.5Q(1.1)^j} \geq 8^j$ and $\frac{1}{2}e^{\frac{1}{2}Q(1.1)^j} \geq 2^j$ for all $j \in \mathbb{N}$. For such Q we obtain

$$\mathbb{P}(\forall j : N_j > e^{Q(1.1)^j} \mid B_0) \geq 1 - 2 \sum_{j=1}^{\infty} e^{-C4^{-j}e^{1.5Q(1.1)^j}} - \sum_{j=1}^{\infty} e^{-\frac{1}{2}e^{\frac{1}{2}Q(1.1)^j}} \geq 1 - 3 \sum_{j=1}^{\infty} e^{-2^j} \geq \frac{1}{3}.$$

Since $t_{j+1} \geq N_j$, we have $\{\lim_{j \rightarrow \infty} t_j = \infty\} \supseteq \{\forall j : N_j \geq e^{Q(1.1)^j}\}$. Thus

$$\mathbb{P}(m_k \in I_1 \forall k \geq t_0 \mid \mathcal{F}_{t_0-1}, B_0) \geq \mathbb{P}(\lim_{j \rightarrow \infty} t_j = \infty \mid \mathcal{F}_{t_0-1}, B_0) \geq \frac{1}{3},$$

as required. \square

At last, we prove Claims 4.4, 4.5 and 4.6.

Proof of Claim 4.4. We observe that, since $\rho^+ \equiv 0$ on I_1 , the process $\{y_k^+ - y_{t_j}^+ : k \in [t_j, t_{j+1}]\}$ is a martingale Z_k whose steps are uniformly bounded by 1, started at 0 and stopped when $k = t_{j+1}$, that is, when $m_k \notin [\ell_j, r_j]$. Observe that at this stopping time $|Z_{t_{j+1}}| = |y_{t_{j+1}}^+ - y_{t_j}^+| \geq N_j$. We may thus apply Azuma's inequality (Lemma 3.1) to obtain

$$\mathbb{P}(t_{j+1} - t_j < N_j^{1.5}, t_j < \infty \mid \mathcal{F}_{t_j-1}) \leq \mathbb{P}(T_{\{X_k > N_j\}} < N_j^{1.5} \mid \mathcal{F}_{t_j-1}) \leq \exp\left(-N_j^2/(2N_j^{1.5})\right) = e^{-\sqrt{N_j}/2}.$$

\square

Proof of Claim 4.5. We begin by noticing that for every $j, k \geq 0$ we have

$$\begin{aligned} \mathbb{P}(x_k \in [r_{j+1}, r_{j+2}] \mid m_k \in I_1, \mathcal{F}_{k-1}) &\geq \mathbb{P}(\{X \in [r_{j+1}, r_{j+2}], Y \in I_2\} \cup \{Y \in [r_{j+1}, r_{j+2}], X \in I_2\}) \\ &= 2\mu([r_{j+1}, r_{j+2}])\mu(I_2) = 2^{-(j+2)}\mu(I_1)\mu(I_2). \end{aligned} \quad (35)$$

The same holds when $[r_j, r_{j+1}]$ is replaced by $[\ell_{j+1}, \ell_j]$.

Now, assume that $t_j < \infty$ and denote $N'_{j+1} := \nu_{t_{j+1}-1}([r_{j+1}, r_{j+2}])$ and $N''_{j+1} := \nu_{t_{j+1}-1}([\ell_{j+2}, \ell_{j+1}])$ so that $N_{j+1} = \min(N'_{j+1}, N''_{j+1})$. Let $Z \sim \text{Bin}(n, p)$ be a binomially distributed random variable with $n = N_j^{1.5}$ and $p = 2^{-j-2}\mu(I_1)\mu(I_2)$. By (35), the random variables N'_{j+1} and N''_{j+1} , conditioned on $t_{j+1} - t_j \geq N_j^{1.5}$ and \mathcal{F}_{t_j-1} , stochastically dominate Z . Thus,

$$\mathbb{P}(N'_{j+1} > 2^{-j-3}\mu(I_1)\mu(I_2)N_j^{1.5} \mid \mathcal{F}_{t_j-1}, t_{j+1} - t_j \geq N_j^{1.5}) \geq \mathbb{P}\left(Z > \frac{pn}{2}\right),$$

and the same bound holds when N'_{j+1} is replaced by N''_{j+1} . Since $N_{j+1} = \min(N'_{j+1}, N''_{j+1})$ we may apply a union bound to obtain

$$\begin{aligned} & \mathbb{P}(B_j \mid \mathcal{F}_{t_j-1}, t_j < \infty, A_j) \\ & \geq 1 - \mathbb{P}(N'_{j+1} \leq \delta 2^{-j} N_j^{1.5} \mid \mathcal{F}_{t_j-1}, A_j) - \mathbb{P}(N''_{j+1} \leq \delta 2^{-j} N_j^{1.5} \mid \mathcal{F}_{t_j-1}, A_j) \\ & \geq 1 - 2\mathbb{P}(Z > \frac{pn}{2}) \\ & \geq 1 - 2e^{-p^2n/2} = 1 - 2e^{-C2^{-2j}N_j^{1.5}}, \end{aligned}$$

where in the last line we applied Hoeffding's bound (Lemma 3.2) with $\varepsilon = p/2$, and $C = 2^{-5}\mu(I_1)^2\mu(I_2)^2$. \square

Proof of Claim 4.6. Setting $X_j = \log N_j$ for $j \in \mathbb{N}_0$, we have:

$$\bigcap_{j \in \mathbb{N}_0} B_j \subseteq \bigcap_{j \in \mathbb{N}_0} \{N_{j+1} \geq \delta 2^{-j} N_j^{1.5}\} \cap \{N_0 \geq n_0\} \subseteq \bigcap_{j \in \mathbb{N}_0} \left\{X_{j+1} \geq \frac{3}{2}X_j - j \log 2 + \log \delta\right\} \cap \{X_0 \geq \log n_0\}.$$

Using induction one sees that, for any $Q > 0$, under the right-most event, $X_j \geq Q(1.1)^j$ for all $j \geq 1$, provided that n_0 is sufficiently large. The claim follows. \square

4.4 Proof of Lemma 2.9: boundedness

We may assume that μ is not compactly supported, since otherwise the lemma is straightforward. By Observation 2.4, there exists M such that $\rho^+(m) > \frac{r}{2}$ for all $m \in (-\infty, M]$ and $\mu((-\infty, M]) \in (0, \frac{r}{4})$. Hence, by Lemma 4.2 applied with $\gamma = -\infty$, $\beta = -\infty$ and $\alpha = M$, the sequence of quantiles m_k almost surely enters $(-\infty, M)$ only a finite number of times, so that m_k is almost surely bounded from below.

By Theorem A and the definition of ρ^+ , $\liminf_{m \rightarrow \infty} \rho^+(m) = -(1-r)$, so that there exists an increasing sequence $b_j \rightarrow \infty$ such that, for all $j \in \mathbb{N}$,

$$\rho^+(b_j) < -\frac{7}{8}(1-r), \tag{36}$$

and $\mu([b_1, \infty)) < \frac{1}{4}$. Let a_j be a sequence be such that $a_j \leq b_j < a_{j+1}$ and

$$\frac{1}{4} \leq \frac{\mu([a_j, b_j])}{\mu([a_j, \infty))}, \quad \frac{\mu((a_j, b_j))}{\mu((a_j, \infty))} \leq \frac{1}{2}. \tag{37}$$

Then for each $x \in [a_j, b_j]$ we have (using (36) and the right side of (37)):

$$\frac{\mu((x, \infty))^2}{(\mu * \mu)([2x, \infty))} = \frac{[\mu((x, b)) + \mu([b, \infty)))]^2}{(\mu * \mu)([2x, \infty))} \leq \frac{4\mu([b_j, \infty))^2}{(\mu * \mu)([2b_j, \infty))} = 4(\rho^+(b_j) + 1 - r) < \frac{1-r}{2}.$$

Therefore,

$$\forall j \in \mathbb{N} : \rho|_{[a_j, b_j]} < -\frac{1-r}{2}. \quad (38)$$

Let $t_0 = n_0 = 1$. For each $j \in \mathbb{N}$ define inductively

$$t_j := \inf\{k > t_{j-1} : m_k > b_{n_{j-1}}\}, \quad k_j = \min\{k \in \mathbb{N} : m_{t_j} < b_k\}.$$

Our goal is to prove that, almost surely, there exists a finite j with $t_j = \infty$. Define for $j \in \mathbb{N}$:

$$s_j := \inf\{k \in \mathbb{N} : x_k \geq a_{n_j}\}.$$

We have $s_j \leq t_j$. By (37) we have

$$\mathbb{P}(x_{s_j} \in [a_{n_j}, b_{n_j}] \mid s_j < \infty, \mathcal{F}_{s_j-1}) \geq \frac{1}{16},$$

which yields

$$\mathbb{P}(\nu_{t_j-1}([a_{n_j}, b_{n_j}]) \geq 1 \mid t_j < \infty, \mathcal{F}_{t_j-1}) \geq \frac{1}{16}.$$

By Proposition 3.8 applied with $a = a_{n_j}$, $b = b_{n_j}$, $\varepsilon = \frac{1-r}{2}$ we have

$$\mathbb{P}(t_j = \infty \mid t_{j-1} < \infty, \mathcal{F}_{t_j-1}, \nu_{t_j-1}([a_{n_j}, b_{n_j}]) \geq 1) > 1 - q.$$

Here q depends only on the negative upper bound on $\rho|_{[a_{n_j}, b_{n_j}]}$, which, by (38) is $-\frac{1-r}{2}$ and thus uniform in j , and the lower bound on $\frac{\mu([a_{n_j}, b_{n_j}])}{\mu([a_{n_j}, \infty))}$, which by the left part of (37) is also uniform in j . We conclude that for all j ,

$$\mathbb{P}(t_j = \infty \mid t_{j-1} < \infty, \mathcal{F}_{t_j-1}) \geq \frac{1}{16}(1 - q) > 0.$$

Hence we may apply the conditional Borel-Cantelli lemma, and obtain that almost surely there exists a finite j for which $t_j = \infty$, as required.

5 Uniqueness of the limit

5.1 Proof of Proposition 1.3: an example of non-uniqueness

Fix a parameter $p \in (0, 1)$. Let μ be the atomic probability measure defined by

$$\mu = \sum_{\ell \in \mathbb{N}_0} (1-p)p^\ell \delta_{1-2^{-\ell}},$$

where δ_x is the Dirac measure at x . Observe that for every $a \in \{1 - 2^{-\ell}\}$:

$$\mathbb{P}\left(\frac{X+Y}{2} \geq a\right) = \mathbb{P}(X \geq a)\mathbb{P}(Y \geq a), \quad (39)$$

where X, Y are i.i.d. μ -distributed random variables. Let $r \in (0, 1)$ be such that $r < 1 - p^2$. At every atom $a = 1 - 2^{-\ell}$ we have

$$\begin{aligned} \rho^+(a) &= \frac{\mathbb{P}(X \geq a)^2}{\mathbb{P}\left(\frac{X+Y}{2} \geq a\right)} - 1 + r = r > 0, \\ \rho(a) &= \frac{\mathbb{P}(X > a)^2}{\mathbb{P}\left(\frac{X+Y}{2} \geq a\right)} - 1 + r = \left(\frac{\mathbb{P}(X > a)}{\mathbb{P}(X \geq a)}\right)^2 - 1 + r = p^2 - 1 + r < 0. \end{aligned}$$

Thus, by Proposition 3.8 applied with $[a, b] = \{a\}$, which is of positive measure and negative drift, for any $k_0 \in \mathbb{N}$ we have

$$\mathbb{P}(m_k \leq a \text{ for all } k \geq k_0 \mid \mathcal{F}_{k_0-1}, m_{k_0} \leq a, \nu_{k_0-1}(\{a\}) \geq 1) > 0.$$

This implies that the stopping time $\tau := \min\{t \in \mathbb{N} : m_t = a\}$ satisfies

$$\mathbb{P}(m_k \leq a \text{ for all } k \geq \tau \mid \mathcal{F}_{\tau-1}, \tau < \infty) > 0.$$

We note that $\mathbb{P}(\tau < \infty) > 0$ (e.g. $\tau = 1$ under the event that both candidates in the first round of applications hold the opinion a , which is an event of positive probability). In addition, by (39), $\mathbb{P}(\exists k > \tau : m_k < a \mid \mathcal{F}_{\tau-1}, \tau < \infty) = 0$, so that $\mathbb{P}(m_k = a \text{ for all } k \geq \tau \mid \mathcal{F}_{\tau-1}, \tau < \infty) > 0$. Hence

$$\mathbb{P}(m_k = a \text{ a.e.}) > 0.$$

Thus we have shown that $\mathbb{P}(\nu_\infty = \delta_a) > 0$ for every atom a of μ – and hence that ν_∞ is supported on an infinite number of limit distributions, as required.

5.2 Proof of Proposition 1.4: a condition for uniqueness

Our proof uses ideas from [9]. Write $\mathbb{P}(X \geq m) = e^{-g(m)}$. By the hypothesis, $g'' \geq 0$ and $g''' \leq 0$ everywhere in $[\mu_{\min}, \mu_{\max}]$. We note that, in particular, g' must be a concave, non-decreasing, positive function on $[\mu_{\min}, \mu_{\max}]$. As these conditions cannot hold when $\mu_{\min} = -\infty$, we may assume, without loss of generality, that $\mu_{\min} = 0$. By (1) we have $\rho(m) = \frac{1}{\Psi(m)} + r - 1$ where

$$\Psi(m) = \frac{\mathbb{P}\left(\frac{X+Y}{2} \geq m\right)}{\mathbb{P}(X \geq m)^2}.$$

We need to prove that $\Psi(m)$ is strictly monotone increasing in m . We have:

$$\mathbb{P}(X + Y \geq 2m) = \int_0^\infty \mathbb{P}(Y \geq 2m - x) d\mu(x) = \int_0^\infty e^{-g(2m-x)} d\mu(x)$$

However $d\mu = -d(e^{-g}) = e^{-g}g'$, so that:

$$\begin{aligned} \Psi(m) &= \frac{\int_0^\infty e^{-g(2m-x)-g(x)} g'(x) dx}{e^{-2g(m)}} \\ &= \int_0^\infty e^{2g(m)-g(x)-g(2m-x)} g'(x) dx \\ &= \int_0^m e^{2g(m)-g(m+x)-g(m-x)} (g'(m+x) + g'(m-x)) dx + e^{2g(m)-g(2m)} \end{aligned}$$

Differentiation with respect to m yields,

$$\begin{aligned} \Psi'(m) &= \int_0^m e^{2g(m)-g(m+x)-g(m-x)} \\ &\quad \times \left((2g'(m) - g'(m+x) - g'(m-x)) (g'(m+x) + g'(m-x)) + g''(m+x) + g''(m-x) \right) dx \\ &\quad + e^{2g(m)-g(2m)} (2g'(m) + 2g'(2m)), \end{aligned}$$

by the premises that $g' \geq 0$ and $g'' \geq 0$ on the support of μ . Moreover, for $x < m$ we have $2g'(m) - g'(m+x) - g'(m-x) = -2xg'''(c_{m,x}) > 0$ (for some $c_{m,x} \in (m-x, m+x)$). Lastly, since $\mu_{\min} = 0$, we have $g'(0) > 0$. We conclude that $\Psi'(m) \geq 4e^{2g(m)-g(2m)}g'(0) > 0$ for all $m \in [\mu_{\min}, \mu_{\max}]$, so that Ψ is strictly monotone on the support of μ , as required.

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