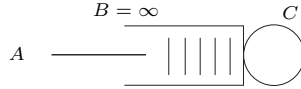


LDP for queue size

Large Deviations and Queues—Damon Wischik

Consider a queue with constant service rate C , with buffer size $B = \infty$, and with arrival process $A = (\dots, A_{-1}, A_0)$ where the A_t are independent and identically distributed. Recall that the queue size Q is given by $Q = q(A)$ where $q(a) = \sup_{t \geq 0} a(-t, 0] - Ct$ and $a(-t, 0] = a_{-t+1} + \dots + a_0$.



Let $\Lambda(\theta) = \log \mathbb{E} e^{\theta A_0}$ and $\Lambda^*(x) = \sup_{\theta \in \mathbb{R}} \theta x - \Lambda(\theta)$. Assume that $\Lambda(\theta)$ is finite for all θ (and thus that $\mathbb{E} A_0$ is finite, and $\Lambda(\theta)$ is infinitely differentiable for all θ).

Theorem 1 *If $\mathbb{E} A_0 < C$ then, for $q > 0$,*

$$\lim_{l \rightarrow \infty} \frac{1}{l} \log \mathbb{P} \left(\frac{Q}{l} > q \right) = -q \sup \{ \theta > 0 : \Lambda(\theta) < \theta C \} \quad (1)$$

$$= - \inf_{t > 0} t \Lambda^*(C + q/t) \quad (2)$$

$$= - \inf_{t > 0} \sup_{\theta \geq 0} \theta(q + Ct) - t \Lambda(\theta) \quad (3)$$

(The limit, infimum and supremum are taken over $l \in \mathbb{R}$, $t \in \mathbb{R}$ and $\theta \in \mathbb{R}$.)

We will split the proof into three parts: the large deviations upper bound

$$\limsup l^{-1} \log \mathbb{P}(Q > lq) \leq (1), \quad (4)$$

the large deviations lower bound

$$\liminf l^{-1} \log \mathbb{P}(Q > lq) \geq (2), \quad (5)$$

and finally (1) = (2) = (3).

Proof of LD upper bound. Write out the probability we wish to estimate, and then use the Chernoff bound. For any $\theta > 0$ such that $\Lambda(\theta) < \theta C$,

$$\begin{aligned} \mathbb{P}(Q > lq) &= \mathbb{P}(\sup_{t \geq 0} A(-t, 0] - Ct > lq) \\ &= \mathbb{P}(A(-t, 0] - Ct > lq \text{ for some } t \geq 0) \\ &\leq \sum_{t \geq 0} \mathbb{P}(A(-t, 0] - Ct \geq lq) \\ &\leq \sum_{t \geq 0} e^{-\theta lq} e^{t\{\Lambda(\theta) - \theta C\}} \quad \text{by Chernoff's bound, since } \theta > 0 \\ &= e^{-\theta lq} \frac{e^{\Lambda(\theta) - \theta C}}{1 - e^{\Lambda(\theta) - \theta C}} \quad \text{the series is summable, since } \Lambda(\theta) < \theta C \end{aligned}$$

and so $\limsup l^{-1} \log \mathbb{P}(Q > lq) \leq -\theta q$. Take the infimum over all such θ to prove the result (4).

Note that if no such θ existed then the supremum would be $-\infty$, by convention, and so the bound would be trivial. But such a θ does exist, because $\Lambda(\theta)$ is finite in a neighbourhood of $\theta = 0$, hence differentiable at $\theta = 0$, and we've assumed that $\Lambda'(0) = \mathbb{E}A_0 < C$; therefore $\Lambda(\theta) < \theta C$ for θ sufficiently small. \square

Proof of LD lower bound. Pick any $u > 0$, $u \in \mathbb{R}$. We will find a lower bound for $\mathbb{P}(Q > lq)$ by estimating the probability that the queue reaches level lq in time lu using Cramér's theorem:

$$\begin{aligned}
\liminf_{l \rightarrow \infty} \frac{1}{l} \log \mathbb{P}(Q > lq) & \tag{6} \\
&= \liminf_{l \rightarrow \infty} \frac{1}{l} \log \mathbb{P}\left(\sup_v A(-v, 0] - Cv > lq\right) \\
&= \liminf_{l \rightarrow \infty} \frac{1}{l} \log \mathbb{P}(A(-v, 0] - Cv > lq \text{ for some } v) \\
&\geq \liminf_{l \rightarrow \infty} \frac{1}{l} \log \mathbb{P}\left(A(-[lu], 0] > lq + C[lu]\right) \quad \text{by choosing } v = [lu] \\
&\geq \liminf_{l \rightarrow \infty} \frac{u}{[lu] - 1} \log \mathbb{P}\left(A(-[lu], 0] > \frac{[lu]}{u}q + C[lu]\right) \quad \text{by bounds}^1 \text{ for } [lu] \\
&= u \liminf_{n \rightarrow \infty} \frac{1}{n-1} \log \mathbb{P}\left(\frac{1}{n}A(-n, 0] > C + \frac{q}{u}\right) \quad \text{where } n = [lu] \\
&\geq u \liminf_{n \rightarrow \infty} \frac{1+\varepsilon}{n} \log \mathbb{P}\left(\frac{1}{n}A(-n, 0] > C + \frac{q}{u}\right) \quad \text{for any } \varepsilon > 0 \\
(6) &\geq -u \liminf_{n \rightarrow \infty} \frac{1}{n} \log \mathbb{P}\left(\frac{1}{n}A(-n, 0] > C + \frac{q}{u}\right) \quad \text{since } \varepsilon > 0 \text{ arbitrary} \\
&\geq -u \inf_{x > C+q/u} \Lambda^*(x) \quad \text{by Cramér's theorem} \\
&= -u\Lambda^*(C + q/u+) \quad \text{since } \Lambda^*(x) \text{ is increasing for } x \geq \mathbb{E}A_0 \\
&\quad \text{where by } f(x+) \text{ we mean } \lim_{y \downarrow x} f(y) \\
(6) &\geq -\inf_{u > 0} u\Lambda^*(C + q/u+) \quad \text{since } u > 0 \text{ arbitrary} \\
&\geq -(t + \delta)\Lambda^*(C + q/(t + \delta)+) \quad \text{choosing } u = t + \delta, \delta > 0 \\
&\geq -(t + \delta)\Lambda^*(C + q/t) \quad \text{since } q/(t + \delta)+ < q/t \text{ and } \Lambda^* \text{ is increasing} \\
(6) &\geq -t\Lambda^*(C + q/t) \quad \text{since } \delta > 0 \text{ arbitrary} \\
(6) &\geq -\inf_{t > 0} t\Lambda^*(C + q/t) \quad \text{since } t > 0 \text{ arbitrary}
\end{aligned}$$

This completes the proof. \square

Equality of rate functions. First, (2)=(3): Expand Λ^* , and use the fact that the supremum over θ in $\Lambda^*(x) = \sup_{\theta \in \mathbb{R}} \theta x - \Lambda(\theta)$ can be taken over $\theta \geq 0$ for $x \geq \mathbb{E}A_0$, as we saw in the proof of Cramér's theorem.

Second, (3) \geq (1): For any $\theta > 0$ with $\Lambda(\theta) < \theta C$,

$$\theta(q + Ct) - t\Lambda(\theta) = \theta q + t(\theta C - \Lambda(\theta)) \geq \theta q$$

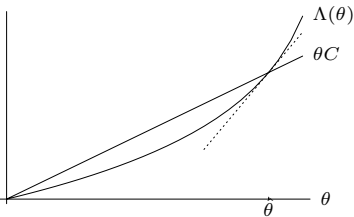
¹Recall that $[x] - 1 < x \leq [x]$, so $l \leq [lu]/u$ and $1/l < u/([lu] - 1)$.

Taking the supremum over such θ ,

$$\begin{aligned} & \sup_{\theta > 0: \Lambda(\theta) < \theta C} \theta(q + Ct) - t\Lambda(\theta) \geq \sup_{\theta > 0: \Lambda(\theta) < \theta C} \theta q \\ \implies & \sup_{\theta \geq 0} \theta(q + Ct) - t\Lambda(\theta) \geq q \sup\{\theta > 0 : \Lambda(\theta) < \theta C\}. \end{aligned}$$

Now take the infimum over $t > 0$.

Finally, (3) \leq (1): Let $\hat{\theta} = \sup\{\theta > 0 : \Lambda(\theta) < \theta C\}$. (The set is non-empty, by our remark in the proof of the LD upper bound.) If $\hat{\theta} = \infty$, we are done. Otherwise, using the fact that Λ is convex and differentiable, it must be that $\Lambda(\hat{\theta}) = \hat{\theta}C$ and $\Lambda'(\hat{\theta}) > C$.



Now consider the supporting tangent to $\Lambda(\theta)$ at $\hat{\theta}$: by convexity, $\Lambda(\theta) \geq \hat{\theta}C + \Lambda'(\hat{\theta})(\theta - \hat{\theta})$, and so

$$\begin{aligned} (3) &= \inf_{t > 0} \sup_{\theta \geq 0} \theta(q + Ct) - t\Lambda(\theta) \\ &\leq \inf_{t > 0} \sup_{\theta \geq 0} \theta(q + Ct) - t(\hat{\theta}C + \Lambda'(\hat{\theta})(\theta - \hat{\theta})) \quad \text{from supporting tangent at } \hat{\theta} \\ &= \inf_{t > 0} \sup_{\theta \geq 0} \theta(q - t(\Lambda'(\hat{\theta}) - C)) + \hat{\theta}t(\Lambda'(\hat{\theta}) - C) \quad \text{gathering } \theta \text{ terms} \\ &= \inf_{t > 0} \begin{cases} \infty & \text{if } t < q/(\Lambda'(\hat{\theta}) - C) \\ \hat{\theta}t(\Lambda'(\hat{\theta}) - C) & \text{else} \end{cases} \quad \text{performing the } \theta\text{-optimization} \\ &= \hat{\theta}q \quad \text{performing the } t\text{-optimization} \\ &= (1). \end{aligned}$$

This completes the proof. □