

# Bounding Stochastic Safety: Leveraging Freedman's Inequality with Discrete-Time Control Barrier Functions, Extended Proofs

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## I. PROOF FOR LEMMA 1

### A. Ville's Proof

Proof of Ville's inequality [1], Lemma ??.

*Proof.* Fix  $\lambda > 0$  and define the stopping time  $\tau \triangleq \{\inf_{k \in \mathbb{N}} W_k > \lambda\}$  with  $\tau = +\infty$  if  $W_k \leq \lambda$  for all time. Since  $W_k$  is a nonnegative supermartingale, the stopped process  $W_{k \wedge \tau}$  is also a nonnegative supermartingale and

$$\mathbb{E}[W_{k \wedge \tau}] \leq \mathbb{E}[W_0] \text{ and } \liminf_{k \rightarrow \infty} \mathbb{E}[W_{k \wedge \tau}] \leq \mathbb{E}[W_0]. \quad (1)$$

We can further bound this in the case that  $\tau$  is finite:

$$\mathbb{E}[W_0] \geq \liminf_{k \rightarrow \infty} \mathbb{E}[W_{k \wedge \tau} \mathbb{1}_{\{\tau < \infty\}}] \quad (2)$$

$$\geq \mathbb{E}[\liminf_{k \rightarrow \infty} W_{k \wedge \tau} \mathbb{1}_{\{\tau < \infty\}}] \quad (3)$$

$$> \mathbb{E}[\lambda \mathbb{1}_{\tau < \infty}] = \lambda \mathbb{P}\{\tau < \infty\} = \lambda \mathbb{P}\{\sup_{k \in \mathbb{N}} W_k > \lambda\}.$$

The first inequality is by the nonnegativity of  $W_k$  and the second inequality is by Fatou's Lemma [2]. Rearranging terms completes the proof.  $\square$

## II. PROOF FOR THEOREM 1 (RSS + c-MARTINGALES)

### A. RSS Proof

*Proof.* Let  $W_k = \alpha^{-k}(M - h(\mathbf{x}_k)) + M(\alpha^{-K} - \alpha^{-k})$ . This is a nonnegative supermartingale for  $k \leq K$ :

$$\mathbb{E}[W_{k+1} | \mathcal{F}_k] \quad (4)$$

$$= \alpha^{-(k+1)}(M - \mathbb{E}[h(\mathbf{x}_{k+1}) | \mathcal{F}_k]) + M(\alpha^{-K} - \alpha^{-(k+1)})$$

$$\leq \alpha^{-(k+1)}(M - \alpha h(\mathbf{x}_k)) + M(\alpha^{-K} - \alpha^{-(k+1)}) \quad (5)$$

$$= \alpha^{-k}(M - h(\mathbf{x}_k)) + M(\alpha^{-K} - \alpha^{-k}) = W_k. \quad (6)$$

Apply Ville's inequality ?? to  $W_k$  to find:

$$\mathbb{P}\left\{\sup_{k \leq K} W_k \leq \lambda\right\} \leq \frac{\mathbb{E}[W_0]}{\lambda}. \quad (7)$$

Note that  $\{\sup_{k \leq K} -h(\mathbf{x}_k) + \alpha^{-K}M > \lambda\} \subseteq \{\sup_{k \leq K} W_k > \lambda\}$ . Choose  $\lambda = \alpha M^{-K}$  to achieve:

$$P_u(K, \mathbf{x}_0) \leq \frac{M\alpha^{-K} - h(\mathbf{x}_0)}{M\alpha^{-K}} = 1 - \frac{h(\mathbf{x}_0)}{M} \alpha^K \quad (8)$$

$\square$

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### B. Steinhardt Proof

*Proof.* Let  $W_k \triangleq M - h(\mathbf{x}_k) + (K - k)c$ . This is a non-negative supermartingale for  $k \leq K$ :

$$W_k = M - h(\mathbf{x}_k) + (K - k)c \geq 0 \quad (9)$$

$$\mathbb{E}[W_{k+1} | \mathcal{F}_k] = M - \mathbb{E}[h(\mathbf{x}_{k+1}) | \mathcal{F}_k] + (K - k - 1)c \quad (10)$$

$$\leq M - h(\mathbf{x}_k) + c + (K - k - 1)c \quad (11)$$

$$= M - h(\mathbf{x}_k) + (K - k)c = W_k \quad (12)$$

Apply Ville's inequality ?? to  $W_k$  to find:

$$\mathbb{P}\left\{\sup_{k \leq K} W_k \leq \lambda\right\} \leq \frac{\mathbb{E}[W_0]}{\lambda} \quad (13)$$

Note that  $\{\sup_{k \leq K} M - h(\mathbf{x}_k) \geq \lambda\} \subseteq \{\sup_{k \leq K} W_k \leq \lambda\}$ . Choose  $\lambda = M$  to achieve:

$$P_u(K, \mathbf{x}_0) \leq \frac{M - h(\mathbf{x}_0) + Kc}{M} = 1 - \frac{h(\mathbf{x}_0) - Kc}{M}.$$

$\square$

## III. PROOF FOR THEOREM 4

### A. New c-martingale proof

Proof of Theorem ??.

*Proof.* Define the candidate function  $W_k = -h(\mathbf{x}_k) - kc + h(\mathbf{x}_0)$ . This satisfies  $W_0 = 0$  and is a supermartingale:

$$\begin{aligned} \mathbb{E}[W_{k+1} | \mathcal{F}_k] &= \mathbb{E}[-h(\mathbf{x}_{k+1}) | \mathcal{F}_k] - (k+1)c + h(\mathbf{x}_0) \\ &\leq -h(\mathbf{x}_k) - kc + h(\mathbf{x}_0) = W_k. \end{aligned} \quad (14)$$

Consider the Doob's decomposition of this supermartingale. The resulting martingale is:

$$M_k = W_k - \sum_{i=1}^k (\mathbb{E}[W_i | \mathcal{F}_{i-1}] - W_{i-1}) \quad (15)$$

The martingale difference is given by:

$$M_k - M_{k-1} = W_k - \mathbb{E}[W_k | \mathcal{F}_{k-1}] \quad (16)$$

$$= -h(\mathbf{x}_k) + \mathbb{E}[h(\mathbf{x}_k) | \mathcal{F}_{k-1}] \quad (17)$$

$$\leq 1 \quad (18)$$

Now we consider the inequalities:

$$\sup_{k \leq K} -h(\mathbf{x}_k) - Kc + h(\mathbf{x}_0) \leq \sup_{k \leq K} h(\mathbf{x}_k) - kc + h(\mathbf{x}_0)$$

$$\leq \sup_{k \leq K} W_k + \sum_{i=1}^k (\mathbb{E}[h(\mathbf{x}_i) | \mathcal{F}_{i-1}] - h(\mathbf{x}_i) + c) \quad (19)$$

$$= \sup_{k \leq K} W_k - \sum_{i=1}^k (\mathbb{E}[W_i | \mathcal{F}_{i-1}] - W_{i-1}) = \sup_{k \leq K} M_k. \quad (20)$$

Consider the case when  $\lambda = -Kc + h(\mathbf{x}_0)$ . Then the safety event satisfies the containment:

$$\left\{ \inf_{k \leq K} h(\mathbf{x}_k) < 0 \right\} \subseteq \left\{ \sup_{k \leq K} M_k \geq -Kc + h(\mathbf{x}_0) \right\}.$$

Applying the probability bound from Theorem ?? results in the probability bound:

$$P_u(K, \mathbf{x}_0) \leq H(-Kc + h(\mathbf{x}_0), \sigma\sqrt{K}) \quad (21)$$

□

#### IV. PROOF OF PROPOSITION 6

*Proof.* By combining the bounds (??) and (??):

$$h(\mathbf{x}_{k+1}) \geq \mathbb{E}[h(\mathbf{x}_{k+1}) | \mathcal{F}_k] - \delta \geq \alpha h(\mathbf{x}_k) - \delta \quad (22)$$

Thus, for  $K < \infty$ , we can lower bound  $h(\mathbf{x}_k) \geq \alpha^k h(\mathbf{x}_0) - \sum_{i=0}^{k-1} \alpha^i \delta$ . Furthermore, for all time,  $h(\mathbf{x}_k) \geq \frac{-\delta}{1-\alpha} \implies h(\mathbf{x}_{k+1}) \geq \frac{-\delta}{1-\alpha}$ , so  $\mathcal{C}_\delta$  is safe. □

#### V. PROOF OF COROLLARY 1

##### A. Proof of Corollary ??

*Proof.* First define  $\eta(\mathbf{x}) = \frac{h(\mathbf{x}) + \epsilon}{\delta}$  and  $W_k = -\alpha^{K-k} \eta(\mathbf{x}_k) + \alpha^K \eta(\mathbf{x}_0) + \sum_{i=1}^k \alpha^{K-i} \epsilon (1 - \alpha)$ . First note that  $h_k$  satisfies:

$$\mathbb{E}[\eta(\mathbf{x}_k) | \mathcal{F}_{k-1}] \geq \alpha(\eta_{k-1}) + \frac{\epsilon}{\delta}(1 - \alpha) \quad (23)$$

Note that  $W_k$  is a supermartingale:

$$\mathbb{E}[W_{k+1} | \mathcal{F}_k] \quad (24)$$

$$\begin{aligned} &= -\alpha^{K-(k+1)} \mathbb{E}[\eta_{k+1} | \mathcal{F}_k] + \alpha^K \eta_0 + \sum_{i=1}^{k+1} \alpha^{K-i} \epsilon (1 - \alpha) / \delta \\ &\leq -\alpha^{K-k} \eta_k + \alpha^K \eta_0 + \sum_{i=1}^k \alpha^{K-i} \epsilon (1 - \alpha) / \delta = W_k \end{aligned} \quad (25)$$

and  $W_0 = 0$ .

Take the Doob's decomposition. Then consider the martingale:  $M_k = W_0 + \sum_{i=1}^k (W_i - \mathbb{E}[W_i | \mathcal{F}_{i-1}])$ . Since the predictable quadratic variation  $\langle h(\mathbf{x}) \rangle_k$  is bounded, so is  $\langle M \rangle_k$ :

$$\langle M \rangle_k = \sum_{i=1}^k \mathbb{E}[(W_i - \mathbb{E}[W_i | \mathcal{F}_{i-1}])^2 | \mathcal{F}_{i-1}] \quad (26)$$

$$\begin{aligned} &= \sum_{i=1}^k \alpha^{2(K-i)} \mathbb{E}[(\eta(\mathbf{x}_i) - \mathbb{E}[\eta(\mathbf{x}_i) | \mathcal{F}_{i-1}])^2 | \mathcal{F}_{i-1}] \\ &\quad (27) \end{aligned}$$

$$\begin{aligned} &= \sum_{i=1}^k \frac{\alpha^{2(K-i)}}{\delta^2} \mathbb{E}[(h(\mathbf{x}_i) - \mathbb{E}[h(\mathbf{x}_i) | \mathcal{F}_{i-1}])^2 | \mathcal{F}_{i-1}] \\ &\quad (28) \end{aligned}$$

$$\leq \sum_{i=1}^k \frac{\alpha^{2(K-i)}}{\delta^2} \sigma^2 \leq \frac{K\sigma^2}{\delta^2} \quad (29)$$

Next we'll bound some stuff:

$$\min_{k \leq K} \eta(\mathbf{x}_k) < 0 \iff \min_{1 \leq k \leq K} \eta(\mathbf{x}_k) \leq 0 \quad (30)$$

$$\implies \max_{1 \leq k \leq K} W_k \geq \alpha^K \eta(\mathbf{x}_0) + \sum_{i=1}^k \alpha^{K-i} \frac{\epsilon}{\delta} (1 - \alpha) \quad (31)$$

$$\implies \max_{1 \leq k \leq K} W_k \geq \alpha^K \eta(\mathbf{x}_0) + \alpha^{K-1} \frac{\epsilon}{\delta} (1 - \alpha) \quad (32)$$

$$\implies \max_{1 \leq k \leq K} M_k \geq \alpha^K \eta(\mathbf{x}_0) + \alpha^{K-1} \frac{\epsilon}{\delta} (1 - \alpha) \quad (33)$$

Lower bound  $h$  as in Theorem ??,  $h(\mathbf{x}_k) \geq \alpha^K h(\mathbf{x}_0) + \sum_{i=0}^{K-1} \alpha^i \delta$  to achieve the indicator probability and then choose  $\lambda = \alpha^K \eta(\mathbf{x}_0) + \alpha^{K-1} \frac{\epsilon}{\delta} (1 - \alpha)$  and apply Theorem ?? to  $M_k$  to finalize the bound. □

#### REFERENCES

- [1] J. Ville, "Etude critique de la notion de collectif," 1939.
- [2] G. Grimmett and D. Stirzaker, *Probability and Random Processes*. Oxford University Press, July 2020.