## Algorithm 1 Adaptive-Quadrature

```
Input: A list xs and tolerance \tau \geq 0 N \leftarrow \operatorname{length}(xs).

if N > 2 then Q \leftarrow N \cdot (xs[0] + xs[-1])/2.

else return \operatorname{sum}(xs).

end if \varepsilon \leftarrow |Q - \operatorname{sum}(xs)|

if \varepsilon \geq \tau then c \leftarrow \lfloor N/2 \rfloor.

Q \leftarrow \operatorname{Adaptive-Quadrature}(xs[:c], \tau/2) + \operatorname{Adaptive-Quadrature}(xs[c:], \tau/2)

end if return Q.
```

# Formatting Instructions for RLC Submissions

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#### Abstract

Hello

#### 1 Introduction

In reinforcement learning, ...

## 2 Problem Setting

There are multiple ways to look and define this problem.

## 3 Algorithm

**Proposition 3.1.** Given tolerance  $\tau > 0$  and a real-valued list xs of size N, Algorithm 1 returns an approximate sum Q that satisfies  $|Q - sum(xs)| \le \tau$ .

*Proof.* This proof will follow by induction. Assume that N < 3. Then Algorithm 1 returns Q = sum(xs). Therefore  $|Q - \text{sum}(xs)| = |\text{sum}(xs) - \text{sum}(xs)| = 0 \le \tau$  for all  $\tau \ge 0$ .

Now assume that for a fixed N > 2 and  $\tau \ge 0$  it holds that  $|Q - \text{sum}(xs)| \le \tau$ .

#### References

Richard S. Sutton and Andrew G. Barto. Reinforcement Learning: An Introduction. The MIT Press, Cambridge, MA, 1998.

### A Experimental Details