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Table 1: Hidden-state inference Overall Decision Quality

	$DQ_o^{urn}$		$DQ_s^{urn}$	
	(1)	(2)	(3)	(4)
<i>Intercept</i>	2.208	1.785	2.572	2.627
	[2.143, 2.270]	[1.738, 1.834]	[2.452, 2.690]	[2.565, 2.691]
$ACC_u$	-0.015			
	[-0.016, -0.013]			
<i>urns</i>		-0.072		
		[-0.100, -0.044]		
<i>cols</i>		-0.090		
		[-0.117, -0.064]		
<i>ballSeq</i>		-0.325		
		[-0.354, -0.297]		
$CC_u$			-0.038	
			[-0.043, -0.033]	
$DV(seq = 2)$				-0.977
				[-1.041, -0.918]
$DV(seq = 3)$				-0.977
				[-1.039, -0.913]
$sd(pID)$	0.125	0.132	0.229	0.212
	[0.084, 0.170]	[0.094, 0.180]	[0.178, 0.295]	[0.176, 0.255]

# 1 Lab Research

## 1.1 Background and Objectives

Overconfidence is a prevalent cognitive bias that impairs decision-making quality (Achtziger & Alós-Ferrer, 2014). The potential role of AI advisors in mitigating this bias in cognitive tasks remains largely unexplored. This study aims to address the following questions:

## 1.2 Analysis of Hidden State Decision Quality

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## References

- Achtziger, A., & Alós-Ferrer, C. (2014). Fast or Rational? A Response-Times Study of Bayesian Updating. *Management Science*. <https://doi.org/10.1287/mnsc.2013.1793>