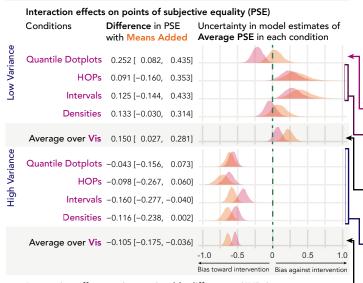
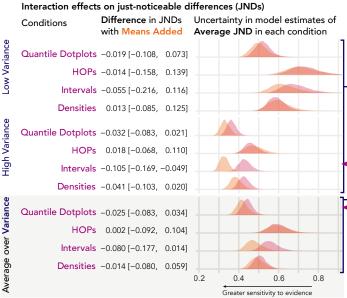
Interaction effects on linear in log odds slopes Uncertainty Difference in slope Uncertainty in model estimates of with Means Added Visualizations Average Slope in each condition Low Variance Quantile Dotplots -0.033 [-0.055, -0.013] HOPs -0.043 [-0.075, -0.013] Intervals -0.023 [-0.043, -0.003] Densities -0.025 [-0.048, -0.002] Average over Vis -0.031 [-0.044, -0.020] High Variance Quantile Dotplots -0.006 [-0.029, 0.018] HOPs -0.038 [-0.071, -0.004] Intervals 0.045 [0.024, 0.064] Densities 0.038 [0.014, 0.063] Average over Vis 0.010 [-0.003, 0.023] Average over Variance Quantile Dotplots -0.020 [-0.038, -0.001] HOPs -0.041 [-0.067, -0.015] Intervals 0.011 [-0.005, 0.027] Densities 0.007 [-0.013, 0.026] 0.2 0.3 0.4 0.5 0.6 0.7 Bias Toward Underestimation





4 RESULTS

4.1 Probability of Superiority Judgments

For each uncertainty visualization, adding means at low variance decreases LLO slopes. Recall that a slope of one corresponds to no bias, and a slope less than one indicates underestimation. When we average over uncertainty visualizations, adding means at low variance reduces LLO slopes for the average user, indicating a very small 0.8 percentage points increase in probability estimation error.

At high variance, the effect of adding means changes directions for different uncertainty visualizations. Adding means decreases LLO slopes for HOPs, whereas adding means increases LLO slopes for intervals and densities. Because differences in LLO slopes represent changes in the exponent of a power law relationship, these slope differences of similar magnitude indicate small reductions in error for probability of superiority judgments of about 1.5 and 1.0 percentage points for intervals and densities, repsectively, and a very small increase in estimation error of 0.3 percentage points for HOPs.

Users in all uncertainty visualziations underestimate effect size. When we **average over variance**, users show an average error of 8.6, 12.4, 14.0, and 14.8 percentage points in probability of superiority units for quantile dotplots, densities, HOPs, and intervals, respectively, each **without means** and in increasing order of bias. In this marginalization, **adding means** only has a reliable impact on LLO slopes for **HOPs**, but the difference is practically negligable.

4.2 Intervention Decisions

4.2.1 Points of Subjective Equality

For each uncertainty visualization, adding means at low variance increases PSE. This results in different effects depending on whether the encoding without means has a PSE below or above optimal. Recall that a PSE of zero describes a utility-optimal decision criterion, negative PSE indicates intervening too often, and positive PSE indicates not intervening enough. Users of quantile dotplots with no means have negative PSE which becomes unbiased when we add means. Users of HOPs and intervals with no means have positive-PSE, a bias which increases when we add means. Users of densities with no means have PSE near zero and become more biased when we add means. Only the effect for quantile dotplots is reliable. When we average over uncertainty visualizations, at low variance the average user may have a PSE 0.6 percentage points above utility-optimal with no means, and adding means increases this mild bias by about 1.7 percentage points in terms of the probability of winning.

At high variance, adding means decreases PSE. Since PSE for all uncertainty visualziations with no means is below optimal, adding means increases bias in PSE in all conditions, however, the effect is only reliable for intervals. When we average over uncertainty visualizations, at high variance the average user has a negative PSE 9.5 percentage points below utilityalwith no means, and adding means increases this bias by about 2.1 percentage points.

4.2.2 Just-Noticeable Differences

-At **low and high variance**, the effects of **adding means** on JNDs are mostly unreliable. Recall that smaller JNDs indicate that a user is sensitive to smaller differences in effect size for the purpose of decision-making. **Adding means** only has a reliable effect on JNDs for **intervals** at **high variance**, where it reduces JNDs by 1.2 percentage points in terms of the probability of winning.

When we average over variance, users of quantile dotplots with means have the smallest JNDs, and users of HOPs with or without means have the largest JNDs, a difference of about 1 percentage point in terms of the probability of winning. Quantile dotplots with or without means have reliably smaller JNDs than other conditions, with the exception of unreliable differences between quantile dotplots with no means and densities with or without means.