

Topic	Title	Link
Calibration	On Calibration of Modern Neural Networks	<a href="https://arxiv.org/abs/1706.04599">https://arxiv.org/abs/1706.04599</a>
Calibration	Improving Predictor Reliability with Selective Recalibration	<a href="https://arxiv.org/abs/2410.05407">https://arxiv.org/abs/2410.05407</a>
Calibration	Obtaining calibrated probability estimates from decision trees and naive Bayesian classifiers	<a href="https://dl.acm.org/doi/10.5555/645530.655658">https://dl.acm.org/doi/10.5555/645530.655658</a>
Calibration	Revisiting the Calibration of Modern Neural Networks	<a href="https://arxiv.org/abs/2106.07998">https://arxiv.org/abs/2106.07998</a>
Calibration	Calibrated Selective Classification	<a href="https://arxiv.org/abs/2208.12084">https://arxiv.org/abs/2208.12084</a>
Selective Prediction	SelectiveNet: A Deep Neural Network with an Integrated Reject Option	<a href="https://arxiv.org/abs/1901.09192">https://arxiv.org/abs/1901.09192</a>
Selective Prediction	Predict Responsibly: Improving Fairness and Accuracy by Learning to Defer	<a href="https://arxiv.org/abs/1711.06664">https://arxiv.org/abs/1711.06664</a>
DFUQ	A Gentle Introduction to Conformal Prediction and Distribution-Free Uncertainty Quantification	<a href="https://arxiv.org/abs/2107.07511">https://arxiv.org/abs/2107.07511</a>
DFUQ	Classification with Valid and Adaptive Coverage	<a href="https://arxiv.org/abs/2006.02544">https://arxiv.org/abs/2006.02544</a>
DFUQ	Learn Then Test	<a href="https://arxiv.org/abs/2110.01052">https://arxiv.org/abs/2110.01052</a>
DFUQ	Quantile Risk Control: A Flexible Framework for Bounding the Probability of High-Loss Predictions	<a href="https://arxiv.org/abs/2212.13629">https://arxiv.org/abs/2212.13629</a>
DFUQ	Distribution-Free Statistical Dispersion Control for Societal Applications	<a href="https://arxiv.org/abs/2309.13786">https://arxiv.org/abs/2309.13786</a>
Bayesian UQ	Evidential Deep Learning to Quantify Classification Uncertainty	<a href="https://arxiv.org/abs/1806.01768">https://arxiv.org/abs/1806.01768</a>
Bayesian UQ	Predictive Uncertainty Estimation via Prior Networks	<a href="https://arxiv.org/abs/1802.10501">https://arxiv.org/abs/1802.10501</a>
Bayesian UQ	Epistemic Neural Networks	<a href="https://arxiv.org/abs/2107.08924">https://arxiv.org/abs/2107.08924</a>
Bayesian UQ	Dropout as a Bayesian Approximation: Representing Model Uncertainty in Deep Learning	<a href="https://arxiv.org/abs/1506.02142">https://arxiv.org/abs/1506.02142</a>
Bayesian UQ	Simple and Scalable Predictive Uncertainty Estimation using Deep Ensembles	<a href="https://arxiv.org/abs/1612.01474">https://arxiv.org/abs/1612.01474</a>
LLMs	Uncertainty in Natural Language Generation: From Theory to Applications	<a href="https://arxiv.org/abs/2307.15703">https://arxiv.org/abs/2307.15703</a>
LLMs	Semantic Uncertainty: Linguistic Invariances for Uncertainty Estimation in Natural Language Generation	<a href="https://arxiv.org/abs/2302.09664">https://arxiv.org/abs/2302.09664</a>
LLMs	Shifting Attention to Relevance: Towards the Predictive Uncertainty Quantification of Free-Form Large Language Models	<a href="https://arxiv.org/abs/2307.01379">https://arxiv.org/abs/2307.01379</a>
LLMs	INSIDE: LLMs' Internal States Retain the Power of Hallucination Detection	<a href="https://arxiv.org/abs/2402.03744">https://arxiv.org/abs/2402.03744</a>
LLMs	Language Models (Mostly) Know What They Know	<a href="https://arxiv.org/abs/2207.05221">https://arxiv.org/abs/2207.05221</a>
LLMs	Just Ask for Calibration: Strategies for Eliciting Calibrated Confidence Scores from Language Models Fine-Tuned with Human	<a href="https://arxiv.org/abs/2305.14975">https://arxiv.org/abs/2305.14975</a>
LLMs	Navigating the Grey Area: How Expressions of Uncertainty and Overconfidence Affect Language Models	<a href="https://arxiv.org/abs/2302.13439">https://arxiv.org/abs/2302.13439</a>
LLMs	R-Tuning: Instructing Large Language Models to Say 'I Don't Know'	<a href="https://arxiv.org/abs/2311.09677">https://arxiv.org/abs/2311.09677</a>
LLMs	CLAM: Selective Clarification for Ambiguous Questions with Generative Language Models	<a href="https://arxiv.org/abs/2212.07769">https://arxiv.org/abs/2212.07769</a>
LLMs	Conformal Language Modeling	<a href="https://arxiv.org/abs/2306.10193">https://arxiv.org/abs/2306.10193</a>
LLMs	Language Models with Conformal Factuality Guarantees	<a href="https://arxiv.org/abs/2402.10978">https://arxiv.org/abs/2402.10978</a>
LLMs	Mitigating LLM Hallucinations via Conformal Abstention	<a href="https://arxiv.org/abs/2405.01563">https://arxiv.org/abs/2405.01563</a>
LLMs	Prompt Risk Control: A Rigorous Framework for Responsible Deployment of Large Language Models	<a href="https://arxiv.org/abs/2311.13628">https://arxiv.org/abs/2311.13628</a>
LLMs	To Believe or Not to Believe Your LLM	<a href="https://arxiv.org/abs/2406.02543">https://arxiv.org/abs/2406.02543</a>
LLMs	Fine-Tuning Language Models via Epistemic Neural Networks	<a href="https://arxiv.org/abs/2211.01568">https://arxiv.org/abs/2211.01568</a>
LLMs	Decomposing Uncertainty for LLMs through Input Clarification Ensembling	<a href="https://arxiv.org/abs/2311.08718">https://arxiv.org/abs/2311.08718</a>
LLMs	Adaptive Elicitation of Latent Information Using Natural Language	<a href="https://openreview.net/forum?id=63c2erbMoc">https://openreview.net/forum?id=63c2erbMoc</a>
UQ	To Rely or Not to Rely? Evaluating Interventions for Appropriate Reliance on Large Language Models	<a href="https://arxiv.org/abs/2412.15584">https://arxiv.org/abs/2412.15584</a>