

# Privacy Engineering Domains

## Data Scientist (including AI)

By the 2024-2025 [IAPP Privacy Engineering Section Advisory Board](#)

“I turn data into valuable insights that drive business strategies and decision-making. However, I often work with sensitive personal information, making privacy a crucial element in my role. I need to ensure that I’m balancing the utility of data with strong privacy practices to protect individuals’ rights and build trust in our data-driven solutions.”

- Data Scientist

<div>Tasks</div> <div></div>	<p><b>Data analysis and modeling:</b> Extract insights only using necessary, proportionate data, ensuring privacy compliance throughout analysis and modelling.</p> <p><b>Privacy-preserving techniques:</b> Apply privacy-enhancing technologies like differential privacy, anonymization, aggregation and federated learning to protect data.</p> <p><b>Privacy impact assessments:</b> Conduct assessments during the planning and design phases to evaluate potential privacy impacts and identify necessary mitigations.</p> <p><b>Govern data use and provenance:</b> Process data for its intended purpose, manage its lifecycle and track consent and provenance to ensure ethical reuse.</p> <p><b>Ensure fairness and protect sensitive data:</b> Identify and address bias risks in AI models and safeguard against unintended inference of sensitive data.</p> <p><b>Collaboration:</b> Work closely with privacy engineers, legal and compliance teams to align data activities with privacy policies and standards.</p>
<div>Professional profile</div> <div></div>	<p><b>Technical competencies:</b> Proficiency in statistical analysis, machine learning, data anonymization, encryption and data lifecycle management</p> <p><b>Areas of experience:</b> Programming, data science, algorithm development, artificial intelligence, data engineering and cloud-based analytics</p> <p><b>AI lifecycle experience:</b> Active across all stages: planning, design, training, evaluation, implementation, deployment, online learning, post-deployment training and maintenance</p> <p><b>Privacy tools:</b> Familiarity with privacy-preserving technologies, such as federated learning, homomorphic encryption and synthetic data generation</p> <p><b>Privacy certifications:</b> Certifications like the Certified Information Privacy Technologist or other data protection credentials to enhance privacy expertise</p>
<div>In the organization</div> <div></div>	<p><b>Reports to:</b> Chief data officer, head of AI or chief technology officer</p> <p><b>Cross-functional collaboration:</b> Works with: privacy engineers, UX designers, legal teams and product managers to ensure privacy is maintained throughout the AI development process</p> <p><b>Key stakeholders:</b> AI product, business operations, product development and marketing teams</p>
<div>Strategic drivers</div> <div></div>	<p><b>Privacy by design:</b> Embed privacy principles in every step of the data analysis process, from data collection to the deployment of models.</p> <p><b>Transparency and accountability:</b> Maintain transparency in data use and establish accountability mechanisms to uphold privacy commitments.</p> <p><b>Ethical data usage:</b> Ensure data models, including AI, are fair, transparent and respectful of individual privacy and societal norms.</p> <p><b>Regulatory adherence:</b> Stay compliant with evolving privacy laws and standards to avoid legal repercussions and enhance business reputation.</p>
<div>Tools and resources</div> <div></div>	<p><b>Privacy-preserving technologies:</b> Pretty Good Privacy, Privacy Preserving Machine Learning, TensorFlow Privacy, Diffprivlib and Microsoft SEAL for privacy-preserving techniques</p> <p><b>Guidance and standards:</b> ISO/TR 31700, NIST Privacy Framework and the European Union Agency for Cybersecurity guidelines for data protection best practices</p> <p><b>Privacy certifications:</b> Certified Information Privacy Technologist and other certifications to deepen privacy expertise</p>
<div>Getting it right means</div> <div></div>	<p><b>Effective data minimization:</b> Collect and only use necessary data to achieve project goals — for example, data required to train or run DataStage models.</p> <p><b>Successful integration of privacy-preserving technologies:</b> Effectively use techniques like differential privacy, federated learning, and secure multi-party computation to protect data.</p> <p><b>Transparency and accountability:</b> Ensure AI systems are explainable and their data usage is transparent to stakeholders and end-users.</p> <p><b>Trust and compliance:</b> Achieve high levels of user trust through transparent data practices and maintain a record free of privacy violations.</p> <p><b>High data utility:</b> Extract actionable insights from data without compromising privacy, ensuring that all analyses align with ethical standards and regulations.</p> <p><b>Bias mitigation and fairness:</b> Maintain fair and unbiased AI models and mechanisms that continuously monitor and correct and deviations.</p>