

# Development Tools - Exam Questions

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## Objective Questions (MCQ/MSQ) - 20 Questions

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1. Which VS Code shortcut opens the command palette?

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- A) `Ctrl+Shift+P`
- B) `Ctrl+P`
- C) `Ctrl+Shift+0`
- D) `Ctrl+K`

**Answer: A**

2. What does the `uv` tool primarily manage in Python projects?

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- A) Virtual environments only
- B) Package dependencies and virtual environments
- C) Code formatting
- D) Testing frameworks

**Answer: B**

3. Which `npx` command creates a new React application?

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- A) `npx react-app create`
- B) `npx create-react-app myapp`
- C) `npx new-react-app`
- D) `npx react create myapp`

**Answer: B**

4. In Git, which command shows the commit history?

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- A) `git history`
- B) `git log`
- C) `git show`
- D) `git commits`

**Answer: B**

## 5. Which CSS selector targets elements with a specific class?

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- A) `#classname`
- B) `.classname`
- C) `@classname`
- D) `*classname`

**Answer: B**

## 6. What does `JSON.parse()` do in JavaScript?

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- A) Converts object to JSON string
- B) Converts JSON string to object
- C) Validates JSON syntax
- D) Formats JSON output

**Answer: B**

## 7. Which bash command displays the current working directory?

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- A) `cwd`
- B) `pwd`
- C) `dir`
- D) `path`

**Answer: B**

## 8. In Excel, which function looks up values in a table?

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- A) `VLOOKUP()`
- B) `XLOOKUP()`

- C) `LOOKUP()`
- D) All of the above

**Answer: D**

### 9. Which SQLite command shows all tables in a database?

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- A) `.tables`
- B) `.show tables`
- C) `SHOW TABLES`
- D) `LIST TABLES`

**Answer: A**

### 10. What does GitHub Copilot primarily provide?

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- A) Version control
- B) AI-powered code suggestions
- C) Project management
- D) Code deployment

**Answer: B**

### 11. Which Unicode encoding is most commonly used for web content?

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- A) UTF-16
- B) UTF-32
- C) UTF-8
- D) ASCII

**Answer: C**

### 12. In browser DevTools, which panel shows network requests?

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- A) Console
- B) Elements

- C) Network
- D) Sources

**Answer: C**

### 13. Which Git command creates a new branch?

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- A) `git branch newbranch`
- B) `git checkout -b newbranch`
- C) `git create newbranch`
- D) Both A and B

**Answer: D**

### 14. What does the `llm` CLI tool primarily interface with?

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- A) Local databases
- B) Large Language Models
- C) Linux systems
- D) Log files

**Answer: B**

### 15. Which Excel function splits text into columns?

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- A) `SPLIT()`
- B) `TEXTSPLIT()`
- C) `SEPARATE()`
- D) `DIVIDE()`

**Answer: B**

### 16. In JSON, which data types are supported?

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- A) String, Number, Boolean
- B) Object, Array, null
- C) Both A and B
- D) Only strings and numbers

**Answer: C**

**17. Which bash operator redirects output to a file?**

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- A) |
- B) >
- C) <
- D) &

**Answer: B**

**18. What does `git clone` do?**

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- A) Creates a copy of a local repository
- B) Downloads a remote repository
- C) Duplicates a branch
- D) Copies files between directories

**Answer: B**

**19. Which VS Code extension is essential for Python development?**

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- A) Python
- B) Pylance
- C) Python Debugger
- D) All of the above

**Answer: D**

**20. In SQLite, which command imports CSV data?**

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- A) `.import file.csv table`
- B) `IMPORT CSV file.csv`
- C) `LOAD DATA file.csv`
- D) `.load file.csv`

**Answer: A**

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# Subjective/Scenario Questions - 20 Questions

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## 1. Development Environment Setup

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Design a complete development environment for a data science team working on Python projects with version control, dependency management, and code quality tools. What tools would you choose and how would you configure them?

**Answer:** Use **VS Code** with Python extensions, **uv** for dependency management and virtual environments, **Git** with conventional commits, **pre-commit hooks** for code quality (black, flake8, mypy), **GitHub** for collaboration, and **Docker** for environment consistency. Configure workspace settings, establish coding standards, and implement CI/CD pipelines.

## 2. Version Control Strategy

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Your team of 10 developers needs a Git workflow for a data science project with notebooks, datasets, and code. Design a branching strategy and collaboration workflow that handles both code and data versioning.

**Answer:** Implement **Git Flow** with feature branches, use **DVC** for data versioning, **nbstripout** for notebook cleaning, and **Git LFS** for large files. Establish branch protection rules, require pull request reviews, use semantic versioning, and implement automated testing. Create separate workflows for experimental notebooks vs. production code.

## 3. Code Quality Framework

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Establish a comprehensive code quality framework for a Python data science project. What tools, standards, and processes would you implement?

**Answer:** Implement **black** for formatting, **flake8** for linting, **mypy** for type checking, **pytest** for testing, **pre-commit** for automation. Establish coding standards, documentation requirements (docstrings), test coverage thresholds, and code review processes. Use **GitHub Actions** for CI/CD and quality gates.

## 4. Cross-Platform Development

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Design a development setup that works consistently across Windows, macOS, and Linux for a distributed team. What challenges would you address and how?

**Answer:** Use **Docker** for environment consistency, **VS Code** with Remote Development extensions, **uv** for cross-platform Python management, and **Git** with proper line ending configuration. Address path differences with **pathlib**, use environment variables for configuration, and provide setup scripts for each platform.

## 5. Database Integration Strategy

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Design a workflow for data scientists to work with multiple databases (SQLite, PostgreSQL, MongoDB) while maintaining code portability and version control. What tools and patterns would you use?

**Answer:** Use **SQLAlchemy** for database abstraction, **environment variables** for connection strings, **database migrations** with Alembic, and **Docker Compose** for local development. Implement connection pooling, query optimization, and data access patterns. Use **DBeaver** or similar tools for database exploration.

## 6. Automated Testing for Data Science

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Create a testing strategy for data science projects that includes data validation, model testing, and pipeline testing. What types of tests would you implement?

**Answer:** Implement **unit tests** for functions, **integration tests** for pipelines, **data validation tests** with Great Expectations, **model performance tests**, and **regression tests** for model outputs. Use **pytest** with fixtures, **property-based testing** with Hypothesis, and **continuous testing** in CI/CD pipelines.

## 7. Documentation and Knowledge Management

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Design a documentation system for a data science team that captures code documentation, data schemas, model specifications, and project knowledge. What tools and processes would you use?

**Answer:** Use **Sphinx** for code documentation, **MkDocs** for project documentation, **Jupyter notebooks** for analysis documentation, **data catalogs** for schema documentation, and **wikis** for knowledge sharing. Implement automated documentation generation, version control for docs, and regular review processes.

## 8. Performance Monitoring and Debugging

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Establish monitoring and debugging practices for data science applications. What tools would you use to identify performance bottlenecks and debug issues?

**Answer:** Use **profiling tools** (cProfile, line\_profiler), **memory monitoring** (memory\_profiler), **logging frameworks** (loguru, structlog), **APM tools** (New Relic, DataDog), and **debugging tools** (pdb, VS Code debugger). Implement performance benchmarking, error tracking, and alerting systems.

## 9. Dependency Management Strategy

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Design a dependency management strategy for a large data science project with multiple teams and environments. How would you handle version conflicts and ensure reproducibility?

**Answer:** Use **uv** for Python dependencies with lock files, **Docker** for system dependencies, **conda** for scientific packages when needed, and **pip-tools** for dependency resolution. Implement dependency scanning for security, regular updates with testing, and environment isolation strategies.

## 10. Code Review Process

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Establish a code review process specifically tailored for data science projects. What guidelines and tools would you implement?

**Answer:** Create **review checklists** for data science code, use **GitHub/GitLab** for pull requests, implement **automated checks** (linting, testing), establish **domain expertise requirements** for reviews, and create **templates** for different types of changes (models, data processing, analysis). Include data validation and reproducibility checks.



## 11. Security and Compliance

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Design security practices for a data science development environment handling sensitive data. What measures would you implement?

**Answer:** Implement **secrets management** (HashiCorp Vault, AWS Secrets Manager), **access controls** (RBAC, MFA), **data encryption** at rest and in transit, **audit logging**, **secure coding practices**, and **compliance frameworks** (SOC 2, GDPR). Use **security scanning** tools and regular security assessments.

## 12. Collaboration Tools Integration

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Design an integrated toolchain that connects development tools with communication and project management platforms. What integrations would you implement?

**Answer:** Integrate **Git** with **Slack/Teams** for notifications, **Jira/Asana** for issue tracking, **GitHub Actions** with project management tools, **VS Code** with collaboration extensions, and **Jupyter** with sharing platforms. Implement automated status updates and progress tracking.

## 13. Local Development Optimization

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Optimize local development environments for data scientists working with large datasets and computationally intensive tasks. What strategies would you implement?

**Answer:** Use **SSD storage** for fast I/O, **sufficient RAM** for in-memory processing, **GPU acceleration** for ML tasks, **local caching** strategies, **data sampling** for development, **incremental processing**, and **efficient development workflows**. Implement resource monitoring and optimization guidelines.

## 14. Remote Development Setup

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Design a remote development solution for data scientists who need access to powerful computing resources and large datasets. What architecture would you implement?

**Answer:** Use **cloud development environments** (GitHub Codespaces, AWS Cloud9), **remote VS Code** connections, **JupyterHub** deployments, **VPN access** to resources, **shared storage** solutions, and **container-based** development. Implement security controls and resource management.

## 15. Continuous Integration for Data Science

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Create a CI/CD pipeline specifically designed for data science projects. What stages and checks would you include?

**Answer:** Include **code quality checks**, **data validation**, **model testing**, **performance benchmarking**, **security scanning**, **documentation generation**, and **deployment automation**. Use **GitHub Actions/GitLab CI** with custom runners, implement **parallel execution**, and create **environment-specific** deployments.

## 16. Tool Standardization Strategy

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Establish tool standardization across a data science organization while allowing flexibility for specific use cases. How would you balance standardization with innovation?

**Answer:** Create **approved tool lists** with justification processes for exceptions, **standard configurations** with customization options, **training programs** for standard tools, **evaluation frameworks** for new tools, and **migration strategies** for tool changes. Implement governance processes and regular reviews.

## 17. Backup and Recovery

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Design backup and recovery strategies for development environments and code repositories. What would you include in your disaster recovery plan?

**Answer:** Implement **automated backups** of repositories, **distributed version control**, **cloud storage** for critical data, **environment recreation** scripts, **documentation backup**, and **recovery testing**. Use **multiple backup locations**, **versioned backups**, and **recovery time objectives**.

## 18. Onboarding Process

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Create an onboarding process for new data scientists joining your development environment. What training and setup procedures would you establish?

**Answer:** Develop **setup automation** scripts, **training materials** for tools and processes, **mentorship programs**, **hands-on exercises**, **documentation walkthroughs**, and **gradual responsibility increase**. Create **checklists**, **video tutorials**, and **feedback mechanisms** for continuous improvement.

## 19. Performance Benchmarking

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Establish performance benchmarking for development tools and processes. How would you measure and improve developer productivity?

**Answer:** Track **build times**, **test execution times**, **deployment frequency**, **code review turnaround**, **bug resolution time**, and **developer satisfaction**. Use **metrics dashboards**, **regular surveys**, **performance profiling**, and **optimization initiatives**. Implement **continuous improvement** processes.

## 20. Legacy System Integration

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Design strategies for integrating modern development tools with legacy systems and databases. What approaches would you use to modernize gradually?

**Answer:** Implement **API wrappers** for legacy systems, **data extraction** pipelines, **gradual migration** strategies, **compatibility layers**, **documentation** of legacy systems, and **risk assessment** for changes. Use **strangler fig pattern**, **feature toggles**, and **parallel running** during transitions.