DSA 610 Redesign, Lecture 13 Outline

#### **Lecture Outline: Data Destruction**

Duration: 50 minutes

#### **1. Introduction to Data Destruction (5 minutes)**

- **Objective:** Understand the importance and methods of data destruction.
- Content:
  - **Definition:** Data destruction refers to the process of eliminating data from storage devices so that it cannot be recovered or reconstructed.
  - **Purpose:** Protect sensitive information, comply with regulations, and free up storage resources.

#### 2. Reasons for Data Destruction (10 minutes)

- a. Security
  - **Prevent Unauthorized Access:** Ensure that sensitive information is not accessible after its lifecycle ends.
  - **Mitigate Data Breaches:** Reduce the risk of data theft and misuse by ensuring that data is fully destroyed.

#### b. Compliance

- **Regulatory Requirements:** Adhere to laws and regulations that mandate the secure destruction of data.
- Industry Standards: Meet standards for data protection and privacy (e.g., PCI DSS, HIPAA).

#### c. Resource Management

- Free Up Storage: Reclaim storage space by removing unnecessary data.
- **Optimize Performance:** Improve system performance by eliminating old or irrelevant data.

#### d. Risk Management

- Protect Confidentiality: Avoid potential liabilities and legal issues associated with data exposure.
- Reduce Legal Risks: Mitigate risks related to data retention beyond required periods.

#### 3. Mechanisms for Data Destruction (15 minutes)

#### a. Physical Destruction

- Shredding: Physically shred hard drives or other storage devices to prevent data recovery.
- **Crushing:** Use mechanical crushers to destroy storage devices.
- Incineration: Burn storage devices to ensure complete destruction.
- **b.** Logical Destruction
  - **Data Wiping:** Use software tools to overwrite data with random patterns, making it unrecoverable.
  - **Degaussing:** Apply a strong magnetic field to disrupt the magnetic storage medium, erasing data.
  - Formatting: Perform a full format on storage devices, though this method is less secure than others.

#### c. Secure Deletion Tools

- Software Tools: Utilize data wiping tools like DBAN (Darik's Boot and Nuke), Eraser, or Blancco.
- Verification: Use tools that verify data destruction by scanning the device for residual data.

#### 4. Regulations and Standards (10 minutes)

a. General Data Protection Regulation (GDPR)

- Article 17: Right to erasure (Right to be forgotten) requires organizations to securely delete personal data upon request.
- **Compliance:** Ensure that data destruction practices align with GDPR requirements for data protection and privacy.

# b. Health Insurance Portability and Accountability Act (HIPAA)

- **Data Disposal Requirements:** HIPAA mandates secure disposal of protected health information (PHI) to protect patient privacy.
- **Compliance:** Implement procedures for the secure destruction of medical records and electronic health information.

# c. Payment Card Industry Data Security Standard (PCI DSS)

- **Requirement 3:** Ensures that cardholder data is protected and securely deleted when no longer needed.
- **Compliance:** Adhere to standards for the secure destruction of payment card information.

# d. Federal Information Security Management Act (FISMA)

- **Requirement:** FISMA mandates secure destruction of federal data to protect sensitive information.
- **Compliance:** Follow guidelines for data destruction in federal agencies and contractors.

# 5. Best Practices for Data Destruction (5 minutes)

## a. Develop a Data Destruction Policy

- **Define Procedures:** Establish clear procedures for data destruction, including roles and responsibilities.
- Schedule Regular Destruction: Implement routine data destruction schedules to manage data lifecycle.

## **b.** Use Certified Providers

- **Third-Party Services:** Employ certified data destruction services for physical and logical destruction.
- **Certification:** Ensure providers have certifications like NAID AAA for secure data destruction.

# c. Document Destruction Processes

- Maintain Records: Keep records of data destruction activities, including methods used and dates.
- Audit Trails: Implement audit trails to verify compliance with destruction policies and regulations.

## d. Ensure Data is Unrecoverable

• Verify Destruction: Use verification tools and methods to ensure that data is fully destroyed and cannot be recovered.

## 6. Q&A and Discussion (5 minutes)

- **Objective:** Address questions and discuss practical considerations for implementing data destruction practices.
- Content:
  - **Q&A Session:** Open the floor for student questions.
  - **Discussion:** Explore real-world scenarios and challenges in data destruction, including case studies or examples.

## Key Takeaways

• **Data Destruction:** Essential for security, compliance, resource management, and risk management.

- **Mechanisms:** Includes physical and logical methods, with various tools and techniques available. •
- Regulations: Adhere to GDPR, HIPAA, PCI DSS, and FISMA requirements for secure data destruction.
- Best Practices: Develop policies, use certified providers, document processes, and ensure data is unrecoverable.

#### **Resources**:

Data Destruction: https://dataspan.com/blog/what-are-the-different-types-of-data-destruction-andwhich-one-should-you-use/

Data Destruction Standards: https://compucycle.com/what-are-current-data-destruction-standards/ Best Practices for Data Destruction: https://studentprivacy.ed.gov/resources/best-practices-datadestruction

Why Data Destruction: https://www.discoverdatascience.org/articles/data-destruction/ 10 Recent Cases of Data Theft: https://compucycle.com/recent-cases-of-data-threat-and-why-datadestruction-is-important/

Data Bearing Device Destruction: https://learn.microsoft.com/en-us/compliance/assurance/assurancedata-bearing-device-destruction

Methods and Techniques: https://www.bitraser.com/knowledge-series/data-destruction-methods-andtechniques.php?srsltid=AfmBOop18mNxwtrhjD1LbpLrNkDRtrO8k9t wwmJ36lKj7H6lN9y8xZo

GDPR: https://heydata.eu/en/magazine/data-destruction-according-to-the-gdpr

HIPAA: https://www.hhs.gov/hipaa/for-professionals/faq/disposal-of-protected-healthinformation/index.html

PCI DSS: https://www.bitraser.com/article/data-erasure-requirements-for-pci-dsscompliance.php?srsltid=AfmBOorGj5f14wva6h5YtaeEe5DIGCnrjcT HwsEy-4E6fPVQmZTrXAk FISMA: https://jatheon.com/blog/fisma-compliance-email-archiving/

# Lecture Outline: Operationalizing a Model in the Data Analysis Lifecycle

**Duration:** 50 minutes

# 1. Introduction to Operationalizing a Model (5 minutes)

- **Objective:** Understand what operationalizing a model entails and its role in the data analysis • lifecycle.
- Content:
  - 0 **Definition:** Operationalizing a model involves integrating a trained model into a production environment where it can be used for real-time decision-making and predictions.
  - Purpose: Ensure that the model is effectively deployed and maintained to deliver value 0 consistently.

## 2. Steps to Operationalize a Model (15 minutes)

## a. Model Deployment

- Integration: Embed the model into a production system, such as a web application, API, or batch • processing system.
- **Environment Setup:** Ensure that the deployment environment (hardware, software) matches the requirements of the model.

## b. Monitoring and Maintenance

Performance Monitoring: Track model performance metrics (accuracy, precision, recall) to ensure it continues to perform well.

- **Drift Detection:** Implement methods to detect concept drift or data drift that may affect model accuracy over time.
- c. Scaling and Optimization
  - Scalability: Ensure the model can handle increasing loads and volumes of data efficiently.
  - **Optimization:** Optimize the model and deployment system for performance, including reducing latency and computational costs.
- d. Security and Compliance
  - Data Security: Protect sensitive data and ensure secure data handling practices.
  - **Regulatory Compliance:** Ensure the model adheres to relevant regulations and industry standards for data privacy and security.

## 3. Considerations and Pitfalls (15 minutes)

#### a. Data Quality and Consistency

- **Data Changes:** Monitor for changes in data quality or distribution that could affect model performance.
- **Consistency:** Ensure that the input data fed into the model in production is consistent with the data used during training.

#### b. Model Drift and Retraining

- Concept Drift: Address shifts in data patterns that may require retraining or updating the model.
- **Retraining Frequency:** Define a strategy for periodic retraining or updating of the model to maintain performance.

#### c. Integration Challenges

- **Compatibility:** Ensure compatibility between the model and production systems, including software and hardware constraints.
- **Testing:** Thoroughly test the model in a staging environment before full deployment to catch potential issues.

## d. Resource Management

- **Computational Resources:** Manage the resources required for model inference, including memory and processing power.
- **Cost:** Monitor and control costs associated with running and maintaining the model in production.

## 4. Iterative Nature of the Data Analysis Lifecycle (10 minutes)

## a. The Iterative Process

- **Continuous Improvement:** Understand that operationalizing a model is part of a broader iterative process involving continuous monitoring, evaluation, and improvement.
- **Feedback Loop:** Establish feedback loops from the production environment to the model development phase to incorporate new insights and data.

## b. Lifecycle Stages

- **Data Collection:** Gather new data from the production environment for ongoing analysis and model enhancement.
- Feature Engineering: Continuously refine features based on new insights and data patterns.
- **Model Evaluation:** Regularly evaluate model performance and adjust as needed based on feedback and performance metrics.

## c. Real-World Example

• **Case Study:** Discuss a case study where iterative improvements and operationalization were successfully implemented, such as deploying a recommendation system or fraud detection model.

#### 5. Best Practices for Operationalizing a Model (5 minutes)

#### a. Documentation and Communication

- **Document Processes:** Maintain thorough documentation of the model deployment process, including configurations, dependencies, and troubleshooting steps.
- **Communicate:** Ensure clear communication between data scientists, engineers, and stakeholders regarding model performance and updates.

#### b. Collaboration

- **Cross-Functional Teams:** Work collaboratively with IT, operations, and business units to ensure seamless integration and alignment with business goals.
- **Feedback Mechanism:** Implement a mechanism for collecting feedback from end-users and stakeholders to continuously refine and improve the model.

#### 6. Q&A and Discussion (5 minutes)

- **Objective:** Address questions and discuss practical considerations for operationalizing a model in the data analysis lifecycle.
- Content:
  - **Q&A Session:** Open the floor for student questions.
  - **Discussion:** Explore challenges and solutions related to model deployment and iterative improvements.

#### Key Takeaways

- **Operationalizing a Model:** Involves deployment, monitoring, scaling, and maintaining models in a production environment.
- **Considerations:** Address data quality, model drift, integration challenges, and resource management.
- Iterative Nature: Understand the continuous cycle of data analysis, including feedback loops and iterative improvements.
- **Best Practices:** Document processes, collaborate with cross-functional teams, and implement feedback mechanisms.

#### **Resources**:

Operationalizing Your Model: <u>https://www.iguazio.com/glossary/operationalizing-machine-learning/</u> <u>https://www.bitstrapped.com/blog/how-to-operationalize-a-machine-learning-model</u> <u>https://www.subex.com/blog/demystifying-mlops-the-art-of-operationalizing-machine-learning/</u> Monitoring Machine Learning Models: <u>https://developer.nvidia.com/blog/a-guide-to-monitoring-machine-learning-models-in-production/</u>

5 Things to Consider: <u>https://tdwi.org/articles/2022/02/14/adv-all-operationalizing-your-machine-learning.aspx</u>

Compliance Considerations: <u>https://iapp.org/news/a/machine-learning-compliance-considerations</u> Model Drift: <u>https://domino.ai/data-science-dictionary/model-drift</u>

End-to-End: <u>https://adabhishekdabas.medium.com/ml-ops-operationalizing-a-machine-learning-model-end-to-end-89a273ed311c</u>

Literature Review: https://ieeexplore.ieee.org/document/9808768

Lecture Outline: Introduction to NLP with Examples in Python Duration: 50 minutes

#### 1. Introduction to Natural Language Processing (NLP) (5 minutes)

- **Objective:** Understand the basics of NLP and its applications.
- Content:
  - **Definition:** NLP is a field of AI that focuses on the interaction between computers and human language.
  - **Applications:** Text classification, sentiment analysis, machine translation, named entity recognition, and more.

#### 2. Regular Expressions (15 minutes)

#### a. Introduction to Regular Expressions

- **Definition:** Regular expressions (regex) are patterns used to match sequences of characters in text.
- **Usage:** Useful for text processing tasks like searching, extracting, and replacing text.

#### **b.** Basic Examples

• Pattern Matching: Find patterns like email addresses, phone numbers, or dates in text.

import re

```
# Example text
text = "Contact us at support@example.com or call 123-456-7890."
```

# Find email addresses
email\_pattern = r'\b[A-Za-z0-9.\_%+-]+@[A-Za-z0-9.-]+\.[A-Z|a-z]{2,}\b'
emails = re.findall(email\_pattern, text)
print("Email addresses:", emails)

# Find phone numbers
phone\_pattern = r'\b\d{3}-\d{3}-\d{4}\b'
phones = re.findall(phone\_pattern, text)
print("Phone numbers:", phones)

#### c. Advanced Examples

• Extracting Dates: Use regex to extract dates from text.

# Example text with dates text = "The project starts on 2024-08-15 and ends on 2024-12-31."

# Find dates in YYYY-MM-DD format date\_pattern = r'\b\d{4}-\d{2}-\d{2}\b' dates = re.findall(date\_pattern, text) print("Dates:", dates)

## 3. Creating Word Clouds (15 minutes)

## a. Introduction to Word Clouds

• **Definition:** A word cloud is a visual representation of word frequency, where the size of each word indicates its frequency in the text.

## b. Example: Generating a Word Cloud

from wordcloud import WordCloud

import matplotlib.pyplot as plt

# Example text

text = "Natural language processing is a fascinating field. NLP allows computers to understand and generate human language. Applications of NLP include text classification, sentiment analysis, and machine translation."

# Generate a word cloud
wordcloud = WordCloud(width=800, height=400, background\_color='white').generate(text)

# Display the word cloud plt.figure(figsize=(10, 5)) plt.imshow(wordcloud, interpolation='bilinear') plt.axis('off') plt.show()

#### c. Customizing Word Clouds

• Adjustments: Customize the appearance by changing the color scheme, adding a mask, or setting maximum words.

# Generate a word cloud with customization
wordcloud\_custom = WordCloud(width=800, height=400, background\_color='black', max\_words=100,
colormap='viridis').generate(text)

# Display the customized word cloud plt.figure(figsize=(10, 5)) plt.imshow(wordcloud\_custom, interpolation='bilinear') plt.axis('off') plt.show()

#### 4. General Introduction to NLP with Examples (10 minutes)

#### a. Text Preprocessing

- Tokenization: Splitting text into words or sentences.
- Example:

from nltk.tokenize import word\_tokenize, sent\_tokenize

text = "Natural language processing is a fascinating field. It allows computers to understand human language."

# Tokenize sentences
sentences = sent\_tokenize(text)
print("Sentences:", sentences)

# Tokenize words
words = word\_tokenize(text)
print("Words:", words)

#### **b.** Basic Sentiment Analysis

• Example using TextBlob:

from textblob import TextBlob

text = "I love programming in Python. It's such a powerful language."

# Create a TextBlob object blob = TextBlob(text)

# Analyze sentiment
sentiment = blob.sentiment
print("Sentiment:", sentiment)

#### c. Named Entity Recognition (NER)

• Example using spaCy:

import spacy

# Load the spaCy model nlp = spacy.load('en\_core\_web\_sm')

text = "Apple Inc. is planning to open a new office in New York."

# Process the text
doc = nlp(text)

# Extract named entities entities = [(ent.text, ent.label\_) for ent in doc.ents] print("Named Entities:", entities)

#### 5. Q&A and Discussion (5 minutes)

- **Objective:** Address questions and discuss practical considerations for using regex, word clouds, and NLP techniques.
- Content:
  - **Q&A Session:** Open the floor for student questions.
  - **Discussion:** Explore real-world applications and challenges in text processing and analysis.

#### Key Takeaways

- **Regular Expressions:** Useful for pattern matching and text processing tasks.
- Word Clouds: Visualize word frequency and text data insights.
- **NLP Techniques:** Basic preprocessing, sentiment analysis, and named entity recognition are foundational techniques in NLP.

#### **Resources**:

NLP: <u>https://www.geeksforgeeks.org/natural-language-processing-nlp-tutorial/</u> Regular Expressions: <u>https://www.w3schools.com/python/python\_regex.asp</u> Word Clouds: <u>https://www.datacamp.com/tutorial/wordcloud-python</u> Text Processing: <u>https://www.geeksforgeeks.org/text-preprocessing-in-python-set-1/</u> Tokenizing: <u>https://www.geeksforgeeks.org/nlp-how-tokenizing-text-sentence-words-works/</u> Sentiment Analysis: <u>https://www.datacamp.com/tutorial/text-analytics-beginners-nltk</u> NER: <u>https://www.wisecube.ai/blog/named-entity-recognition-ner-with-python/</u>