SHMTools Dataset Management

Overview

This notebook provides comprehensive documentation and management utilities for the SHMTools example datasets. It covers:

- 1. Dataset Overview: Physical descriptions and experimental setups
- 2. **Data Loading**: Standardized loading procedures and validation
- 3. **Dataset Exploration**: Structure analysis and visualization
- 4. **Usage Examples**: Common data access patterns for different examples
- 5. Integrity Validation: Automated checking of dataset completeness

This notebook serves as both documentation and a practical utility for working with SHMTools data.

Setup

Import required modules and setup the environment.

```
In [1]: # Setup path to find shmtools
        import sys
        from pathlib import Path
        # Handle different execution contexts
        current dir = Path.cwd()
        possible_paths = [
            current_dir, # From project root
            current_dir.parent, # From examples/
            current_dir.parent.parent, # From examples/notebooks/
            current_dir.parent.parent.parent, # From examples/notebooks/utilities/
        1
        for path in possible_paths:
            if (path / 'shmtools').exists():
                if str(path) not in sys.path:
                    sys.path.insert(0, str(path))
                print(f"Found shmtools at: {path}")
                break
        else:
            print("Warning: Could not find shmtools module")
        # Import libraries
        import numpy as np
        import matplotlib.pyplot as plt
        from typing import Dict, Any
```

```
# Import SHMTools data utilities
from shmtools.utils.data_loading import (
    load 3story data,
   load_cbm_data,
    load_active_sensing_data,
    load_sensor_diagnostic_data,
    load_modal_osp_data,
    get_available_datasets,
    check data availability,
    validate_dataset_integrity,
    print_dataset_summary,
    load_example_data,
    get data dir
# Setup plotting defaults
plt.style.use('default')
plt.rcParams['figure.figsize'] = (12, 8)
plt.rcParams['font.size'] = 10
print("Setup complete.")
```

Found shmtools at: /Users/eric/repo/shm/shmtools-python Setup complete.

```
/Users/eric/repo/shm/shmtools-python/shmtools/classification/nlpca.py:27: Us erWarning: TensorFlow not available. NLPCA functions will not work. Install TensorFlow: pip install tensorflow warnings.warn(
```

Dataset Availability Check

Check which datasets are currently available and validate their integrity.

```
In [2]: # Check dataset availability
        print("Dataset Availability:")
        print("=" * 60)
        check_data_availability()
      Dataset Availability:
      Data directory: /Users/eric/repo/shm/shmtools-python/examples/data
      Directory exists: True
      data3SS.mat
                                     ( 25.0 MB) - ✓ Available
      data CBM.mat
                                     ( 54.0 MB) - ✓ Available
      data_example_ActiveSense.mat ( 32.0 MB) - ✓ Available
      dataSensorDiagnostic.mat
                                   ( 0.1 MB) – ✓ Available
      data OSPExampleModal.mat ( 0.1 MB) - ✓ Available
      Available: 5/5 datasets (111.1 MB total)
In [3]: # Comprehensive dataset summary
        print_dataset_summary()
```

====
SHMTools Dataset Summary

```
data3SS.mat (25 MB) - ✓ VALID
Description: 3-story structure base excitation (primary dataset)
Examples: PCA, Mahalanobis, SVD, NLPCA, Factor Analysis
Structure:
  shape: (8192, 5, 170)
  fs: 2000.0
  num states: 17
  channels: ['Force', 'Ch2', 'Ch3', 'Ch4', 'Ch5']
data_CBM.mat (54 MB) - ✓ VALID
Description: Condition-based monitoring (rotating machinery)
Examples: CBM Bearing Analysis, CBM Gearbox Analysis
Structure:
 shape: (10240, 4, 384)
 fs: 2048.0
 fault_states: 6
 channels: ['Tachometer', 'Gearbox_Accel', 'Top_Bearing_Accel', 'Side_Beari
data example ActiveSense.mat (32 MB) - ✓ VALID
Description: Guided wave ultrasonic measurements
Examples: Active Sensing Feature Extraction
Structure:
variables: ['__header__', '__version__', '__globals__', 'waveformBase', 'w
aveformTest', 'borderStruct', 'sampleRate', 'sensorLayout', 'pairList', 'act
uationWaveform', 'damageLocation']
 estimated_size: 39686248
dataSensorDiagnostic.mat (0.06 MB) - ✓ VALID
_____
Description: Piezoelectric sensor impedance measurements
Examples: Sensor Diagnostics
Structure:
 variables: ['_header__', '__version__', '__globals__', 'sd_ex_broken', 's
 estimated_size: 73692
data OSPExampleModal.mat (0.05 MB) - ✓ VALID
Description: Modal analysis and optimal sensor placement
Examples: Modal Features, Optimal Sensor Placement
Structure:
  variables: ['__header__', '__version__', '__globals__', 'modeShapes', 'res
pDOF', 'nodeLayout', 'elements']
 estimated size: 90096
______
```

Summary: 5/5 valid, 5/5 available

Physical Description

The primary dataset contains measurements from a 3-story aluminum frame structure designed for structural health monitoring research at Los Alamos National Laboratory.

Physical Structure:

- Aluminum columns (17.7×2.5×0.6 cm) and plates (30.5×30.5×2.5 cm)
- 4-column frame design per floor (essentially 4-DOF system)
- Sliding rails constraining motion to x-direction only
- Suspended center column with adjustable bumper for damage simulation
- Base isolation using rigid foam

Instrumentation:

- Electrodynamic shaker for base excitation (band-limited random 20-150 Hz)
- Load cell measuring input force (2.2 mV/N sensitivity)
- 4 accelerometers at floor centerlines (1000 mV/g sensitivity)
- National Instruments PXI data acquisition system

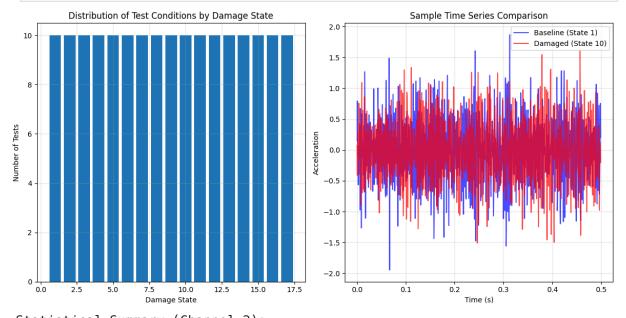
```
In [4]: # Load and examine 3-story structure data
        try:
            data_3story = load_3story_data()
            print("3-Story Structure Dataset:")
            print("=" * 50)
            print(f"Dataset shape: {data_3story['dataset'].shape}")
            print(f"Sampling frequency: {data 3story['fs']} Hz")
            print(f"Channels: {data 3story['channels']}")
            print(f"Total conditions: {len(data 3story['conditions'])}")
            print(f"Damage states: {len(data 3story['state descriptions'])}")
            print(f"Description: {data_3story['description']}")
            # Show damage state descriptions
            print("\nDamage State Descriptions:")
            print("-" * 50)
            for state, desc in data 3story['state descriptions'].items():
                print(f"State {state:2d}: {desc}")
        except FileNotFoundError as e:
            print(f"3-Story dataset not available: {e}")
            print("Please download data3SS.mat and place it in the data directory.")
```

```
3-Story Structure Dataset:
Dataset shape: (8192, 5, 170)
Sampling frequency: 2000.0 Hz
Channels: ['Force', 'Ch2', 'Ch3', 'Ch4', 'Ch5']
Total conditions: 170
Damage states: 17
Description: 3-story structure base excitation data (LANL)
Damage State Descriptions:
State 1: Undamaged - Baseline condition
State 2: Undamaged - Mass = 1.2 kg at the base
State 3: Undamaged - Mass = 1.2 kg on the 1st floor
State 4: Undamaged - 87.5% stiffness reduction in column 1BD
State 5: Undamaged - 87.5% stiffness reduction in columns 1AD and 1BD
State 6: Undamaged - 87.5% stiffness reduction in column 2BD
State 7: Undamaged - 87.5% stiffness reduction in columns 2AD and 2BD
State 8: Undamaged - 87.5% stiffness reduction in column 3BD
State 9: Undamaged - 87.5% stiffness reduction in columns 3AD and 3BD
State 10: Damaged - Gap = 0.20 mm
State 11: Damaged - Gap = 0.15 mm
State 12: Damaged - Gap = 0.13 mm
State 13: Damaged - Gap = 0.10 mm
State 14: Damaged - Gap = 0.05 mm
State 15: Damaged - Gap = 0.20 mm and mass = 1.2 kg at the base
State 16: Damaged - Gap = 0.20 mm and mass = 1.2 kg on the 1st floor
State 17: Damaged - Gap = 0.10 mm and mass = 1.2 kg on the 1st floor
```

Data Structure Analysis

```
In [5]: # Analyze data structure if available
        if 'data_3story' in locals():
            dataset = data 3story['dataset']
            damage_states = data_3story['damage_states']
            # Plot damage state distribution
            plt.figure(figsize=(12, 6))
            plt.subplot(1, 2, 1)
            unique_states, counts = np.unique(damage_states, return_counts=True)
            plt.bar(unique_states, counts)
            plt.xlabel('Damage State')
            plt.ylabel('Number of Tests')
            plt.title('Distribution of Test Conditions by Damage State')
            plt.grid(True, alpha=0.3)
            # Plot example time series from different states
            plt.subplot(1, 2, 2)
            t = np.arange(dataset.shape[0]) / data 3story['fs']
            # Plot baseline condition (state 1, test 1) - channel 2
            baseline_signal = dataset[:1000, 1, 0] # First 1000 points, channel 2,
            plt.plot(t[:1000], baseline_signal, 'b-', label='Baseline (State 1)', al
```

```
# Plot damaged condition (state 10, test 1) - channel 2
damage_signal = dataset[:1000, 1, 90] # First 1000 points, channel 2, d
plt.plot(t[:1000], damage_signal, 'r-', label='Damaged (State 10)', alph
plt.xlabel('Time (s)')
plt.ylabel('Acceleration')
plt.title('Sample Time Series Comparison')
plt.legend()
plt.grid(True, alpha=0.3)
plt.tight_layout()
plt.show()
# Statistical summary
print("\nStatistical Summary (Channel 2):")
print("-" * 40)
# Compare baseline vs damaged conditions
baseline_data = dataset[:, 1, :90].flatten() # All baseline conditions,
damaged_data = dataset[:, 1, 90:].flatten() # All damaged conditions,
print(f"Baseline - Mean: {np.mean(baseline_data):.4f}, Std: {np.std(base
print(f"Damaged - Mean: {np.mean(damaged_data):.4f}, Std: {np.std(damag
print(f"RMS Ratio (Damaged/Baseline): {np.std(damaged_data)/np.std(basel
```



Statistical Summary (Channel 2):

Baseline - Mean: -0.0039, Std: 0.5019 Damaged - Mean: -0.0039, Std: 0.4938 RMS Ratio (Damaged/Baseline): 0.984

Dataset 2: Condition-Based Monitoring (data_CBM.mat)

Physical Description

Rotating machinery vibration data collected from the SpectraQuest Magnum Machinery Fault Simulator for bearing and gearbox fault analysis.

Test Setup:

- Main shaft: 3/4" diameter steel, 28.5" center-to-center bearing support
- Gearbox: Hub City M2, 1.5:1 ratio, 18/27 teeth (pinion/gear)
- Belt drive: ~1:3.71 ratio, 13" span, 3.7 lbs tension
- Magnetic brake: 1.9 lbs-in torsional load
- Shaft speed: ~1000 rpm nominally constant

Fault Conditions:

- Ball bearing faults (roller spin)
- Gearbox worn tooth faults
- Baseline conditions with ball and fluid bearings

```
In [6]: # Load and examine CBM data
        try:
            data_cbm = load_cbm_data()
            print("Condition-Based Monitoring Dataset:")
            print("=" * 50)
            # Show available variables
            print("Available variables:")
            for key, value in data_cbm.items():
                if isinstance(value, np.ndarray):
                    print(f" {key}: {value.shape} ({value.dtype})")
                else:
                    print(f" {key}: {value}")
            # Show fault state descriptions
            if 'fault_states' in data_cbm:
                print("\nFault State Descriptions:")
                print("-" * 50)
                for state, desc in data cbm['fault states'].items():
                    print(f"State {state}: {desc}")
            # Show bearing fault frequencies if available
            shaft_freq = data_cbm['shaft_speed_rpm'] / 60.0 # Convert RPM to Hz
            print(f"\nBearing Fault Frequencies (Shaft = {shaft_freq:.1f} Hz):")
            print("-" * 50)
            print(f"Cage Speed: {3.048 * shaft_freq:.1f} Hz")
            print(f"Outer Race: {3.048 * shaft_freq:.1f} Hz")
            print(f"Inner Race: {4.95 * shaft_freq:.1f} Hz")
            print(f"Ball Spin: {1.992 * shaft_freq:.1f} Hz")
        except FileNotFoundError as e:
            print(f"CBM dataset not available: {e}")
            print("Please download data_CBM.mat and place it in the data directory."
```

```
Condition—Based Monitoring Dataset:
Available variables:
  __header__: b'MATLAB 5.0 MAT-file, Platform: PCWIN64, Created on: Fri Jul
26 13:27:52 2013'
 __version__: 1.0
  __globals__: []
  Fs: (1, 1) (uint16)
  damageStates: (384, 1) (uint8)
  dataset: (10240, 4, 384) (float32)
  stateList: (384, 1) (uint8)
  fs: 2048.0
  duration: 5.0
  shaft speed rpm: 1000.0
  channels: ['Tachometer', 'Gearbox_Accel', 'Top_Bearing_Accel', 'Side_Beari
ng Accel']
  fault_states: {1: 'Baseline 1 (ball bearings, healthy)', 2: 'Baseline 2 (b
all bearings, healthy)', 3: 'Ball roller spin fault', 4: 'Baseline 1 (fluid
bearings, healthy)', 5: 'Baseline 2 (fluid bearings, healthy)', 6: 'Gearbox
worn tooth fault'}
 description: SpectraQuest Magnum fault simulator CBM data (LANL)
Fault State Descriptions:
State 1: Baseline 1 (ball bearings, healthy)
State 2: Baseline 2 (ball bearings, healthy)
State 3: Ball roller spin fault
State 4: Baseline 1 (fluid bearings, healthy)
State 5: Baseline 2 (fluid bearings, healthy)
State 6: Gearbox worn tooth fault
Bearing Fault Frequencies (Shaft = 16.7 Hz):
_____
Cage Speed: 50.8 Hz
Outer Race: 50.8 Hz
Inner Race: 82.5 Hz
Ball Spin: 33.2 Hz
```

CBM Data Visualization

```
In [7]: # Visualize CBM data if available
if 'data_cbm' in locals() and 'dataset' in data_cbm:
    cbm_dataset = data_cbm['dataset']
    fs = data_cbm['fs']
    channels = data_cbm['channels']

print(f"CBM Dataset shape: {cbm_dataset.shape}")

# Plot example signals from different channels and conditions
plt.figure(figsize=(14, 10))

# Time vector
t = np.arange(1000) / fs # First 1000 points for visualization

# Plot signals from each channel
```

```
for i, channel in enumerate(channels):
          plt.subplot(2, 2, i+1)
          # Plot baseline condition (assuming condition 0)
          if cbm_dataset.shape[2] > 0:
               baseline_signal = cbm_dataset[:1000, i, 0]
               plt.plot(t, baseline_signal, 'b-', label='Baseline', alpha=0.7)
          # Plot fault condition (assuming later condition exists)
          if cbm_dataset.shape[2] > 64: # If we have fault conditions
               fault_signal = cbm_dataset[:1000, i, 64]
               plt.plot(t, fault_signal, 'r-', label='Fault', alpha=0.7)
          plt.xlabel('Time (s)')
          plt.ylabel('Amplitude')
          plt.title(f'{channel}')
          plt.legend()
          plt.grid(True, alpha=0.3)
      plt.tight_layout()
      plt.show()
 elif 'data_cbm' in locals():
      print("CBM data loaded but 'dataset' variable not found.")
      print("Available variables:", list(data_cbm.keys()))
CBM Dataset shape: (10240, 4, 384)
                    Tachometer
                                                                Gearbox_Accel
                                             1.00
                                             0.75
                                             0.50
                                             0.25
                                             -0.25
                                             -0.50
                                             -0.75
  0
                                             -1.00
    0.0
                                 0.4
            0.1
                                         0.5
                   0.2
                     Time (s)
                  Top_Bearing_Accel
                                                               Side_Bearing_Accel
                                              0.4
 0.4
                                              0.3
                                              0.2
 0.2
                                              0.1
                                              0.0
 0.0
                                             -0.1
-0.2
                                              -0.2
                                              -0.3
                                                                   Fault
                     Time (s)
```

Other Datasets

Brief exploration of the remaining datasets used in specialized examples.

```
In [8]: # Load other datasets if available
        datasets to check = [
            ('Active Sensing', load_active_sensing_data),
            ('Sensor Diagnostic', load_sensor_diagnostic_data),
            ('Modal OSP', load_modal_osp_data)
        loaded_datasets = {}
        for name, loader_func in datasets_to_check:
                data = loader_func()
                loaded_datasets[name] = data
                print(f"\n{name} Dataset:")
                print("=" * (len(name) + 10))
                # Show dataset structure
                total size = 0
                for key, value in data.items():
                    if isinstance(value, np.ndarray):
                        size_mb = value.nbytes / (1024**2)
                        total_size += size_mb
                        print(f" {key}: {value.shape} ({value.dtype}) - {size_mb:.2
                    elif isinstance(value, (list, dict)):
                        print(f" {key}: {type(value).__name__} (length: {len(value)
                    else:
                        print(f" {key}: {value}")
                print(f"Total estimated size: {total_size:.2f} MB")
            except FileNotFoundError:
                print(f"\n{name} dataset not available.")
            except Exception as e:
                print(f"\nError loading {name} dataset: {e}")
```

```
Active Sensing Dataset:
  header : b'MATLAB 5.0 MAT-file, Platform: PCWIN, Created on: Thu Jul 29
10:24:53 2010'
  __version__: 1.0
  __globals__: list (length: 0)
  waveformBase: (10000, 496) (float32) - 18.92 MB
  waveformTest: (10000, 496) (float32) - 18.92 MB
  borderStruct: (1, 1) ([('outside', '0')]) - 0.00 MB
  sampleRate: (1, 1) (float32) - 0.00 MB
  sensorLayout: (3, 32) (float32) - 0.00 MB
  pairList: (2, 496) (float32) - 0.00 MB
  actuationWaveform: (469, 1) (float32) - 0.00 MB
  damageLocation: (2, 1) (float32) - 0.00 MB
Total estimated size: 37.85 MB
Sensor Diagnostic Dataset:
_____
   header : b'MATLAB 5.0 MAT-file, Platform: PCWIN, Created on: Thu Jul 29
10:28:02 2010'
  __version__: 1.0
  __globals__: list (length: 0)
  sd_ex_broken: (801, 13) (float32) - 0.04 MB
  sd_ex: (801, 10) (float32) - 0.03 MB
Total estimated size: 0.07 MB
Modal OSP Dataset:
_____
  __header__: b'MATLAB 5.0 MAT-file, Platform: PCWIN, Created on: Thu Jul 29
10:26:27 2010'
  __version__: 1.0
  __globals__: list (length: 0)
  modeShapes: (1260, 13) (float32) - 0.06 MB
  respDOF: (1260, 2) (float32) - 0.01 MB
  nodeLayout: (4, 420) (float32) - 0.01 MB
  elements: (9, 216) (float32) - 0.01 MB
Total estimated size: 0.09 MB
```

Dataset Usage Examples

Demonstrate common data access patterns for different types of SHMTools examples.

```
In [9]: # Example 1: Loading data for outlier detection examples (PCA, Mahalanobis,
    print("Example 1: Outlier Detection Data Loading")
    print("=" * 50)

try:
    # This convenience function preprocesses the 3-story data for outlier de
    pca_data = load_example_data('pca')

    print(f"Preprocessed signals shape: {pca_data['signals'].shape}")
    print(f"Channels included: {pca_data['channels']}")
    print(f"Time points (t): {pca_data['t']}")
    print(f"Channels (m): {pca_data['m']}")
```

```
# Show how to split into baseline and damaged conditions
             signals = pca data['signals']
             damage_states = pca_data['damage_states']
             # Extract baseline conditions (states 1-9)
             baseline indices = np.where(damage states <= 9)[0]</pre>
             damaged indices = np.where(damage states >= 10)[0]
             baseline_signals = signals[:, :, baseline_indices]
             damaged signals = signals[:, :, damaged indices]
             print(f"Baseline conditions: {baseline signals.shape[2]} tests")
             print(f"Damaged conditions: {damaged signals.shape[2]} tests")
         except FileNotFoundError:
             print("3-story dataset required for outlier detection examples not avail
        Example 1: Outlier Detection Data Loading
           _____
        Preprocessed signals shape: (8192, 4, 170)
       Channels included: ['Ch2', 'Ch3', 'Ch4', 'Ch5']
       Time points (t): 8192
       Channels (m): 4
        Conditions (n): 170
        Baseline conditions: 90 tests
       Damaged conditions: 80 tests
In [10]: # Example 2: Accessing specific damage states
         print("\nExample 2: Accessing Specific Damage States")
         print("=" * 50)
         if 'pca data' in locals():
             damage_states = pca_data['damage_states']
             state_descriptions = pca_data['state_descriptions']
             signals = pca data['signals']
             # Access specific states
             target states = [1, 10, 14] # Baseline, first damage, severe damage
             for state in target states:
                 state_indices = np.where(damage_states == state)[0]
                 state_signals = signals[:, :, state_indices]
                 # Calculate RMS for each test in this state
                 rms_values = np.sqrt(np.mean(state_signals**2, axis=0)) # RMS over
                 mean_rms = np.mean(rms_values, axis=1) # Mean RMS across tests
                 print(f"State {state}: {state descriptions[state]}")
                 print(f" Tests: {len(state_indices)}")
                 print(f" Mean RMS per channel: {mean_rms}")
                 print()
```

print(f"Conditions (n): {pca_data['n']}")

```
In [11]: # Example 3: Training/Testing splits commonly used in examples
             damage states = pca data['damage states']
             # Common split: Use baseline conditions for training
             baseline indices = np.where(damage states <= 9)[0] # States 1-9
             damaged_indices = np.where(damage_states >= 10)[0] # States 10-17
             training_signals = signals[:, :, baseline_indices]
             testing_signals = signals[:, :, np.concatenate([baseline_indices, damage
             # Create binary labels for testing (0=undamaged, 1=damaged)
             test_damage_states = damage_states[np.concatenate([baseline_indices, dam
             binary_labels = (test_damage_states >= 10).astype(int)
             print(f"Training set: {training signals.shape[2]} undamaged conditions")
             print(f"Testing set: {testing signals.shape[2]} total conditions")
             print(f" - Undamaged: {np.sum(binary_labels == 0)} tests")
             print(f" - Damaged: {np.sum(binary labels == 1)} tests")
             # Alternative split: Use subset of each state for training
             print("\nAlternative split (subset training):")
             train indices = []
             test indices = []
             for state in range(1, 18): # States 1-17
                 state_indices = np.where(damage_states == state)[0]
                 # Use first 7 tests for training, last 3 for testing
                 train indices.extend(state indices[:7])
                 test_indices.extend(state_indices[7:])
             train_indices = np.array(train_indices)
             test_indices = np.array(test_indices)
             print(f"Training set: {len(train indices)} conditions from all states")
             print(f"Testing set: {len(test_indices)} conditions from all states")
```

Dataset Integrity Validation

Automated validation of all datasets to ensure they're properly loaded and structured.

```
In [12]: # Run comprehensive dataset validation
         print("Dataset Integrity Validation")
         print("=" * 50)
         validation_results = validate_dataset_integrity()
         # Create summary table manually (without pandas)
         print(f"{'Dataset':<30} {'Size (MB)':<10} {'Available':<10} {'Valid':<8} {'E</pre>
         print("-" * 80)
         for dataset name, result in validation results.items():
             dataset_info = get_available_datasets()[dataset_name]
             dataset_file = dataset_info['file']
             size_mb = dataset_info['size_mb']
             available = '/' if result['available'] else 'x'
             valid = '/' if result['valid'] else 'x'
             errors = len(result['errors'])
             warnings = len(result['warnings'])
             print(f"{dataset_file:<30} {size_mb:<10} {available:<10} {valid:<8} {err</pre>
         # Show detailed errors/warnings if any
         print("\nDetailed Issues:")
         print("-" * 30)
         issues found = False
         for dataset_name, result in validation_results.items():
             if result['errors'] or result['warnings']:
                 issues found = True
                 dataset_info = get_available_datasets()[dataset_name]
                 print(f"\n{dataset info['file']}:")
                 for error in result['errors']:
                     print(f" ERROR: {error}")
                 for warning in result['warnings']:
                     print(f" WARNING: {warning}")
```

nac	0120 (1.0)	7.701 000 00	V G C E G	211015	
ngs					
data3SS.mat	25	✓	/	0	0
data_CBM.mat	54	✓	1	0	0
data_example_ActiveSense.mat	32	1	/	0	0
dataSensorDiagnostic.mat	0.06	1	/	0	0
data_OSPExampleModal.mat	0.05	1	/	0	0

Detailed Issues:

No issues found. All available datasets are valid.

Dataset File Information

Detailed file information and download guidance.

```
In [13]: # Show data directory and file information
         data_dir = get_data_dir()
         print(f"Data Directory: {data_dir}")
         print(f"Directory exists: {data_dir.exists()}")
         print()
         if data_dir.exists():
             print("Files in data directory:")
             print("-" * 40)
             # List all .mat files
             mat_files = list(data_dir.glob('*.mat'))
             if mat files:
                 for mat_file in sorted(mat_files):
                     size_mb = mat_file.stat().st_size / (1024**2)
                     print(f" {mat_file.name:30} ({size_mb:6.2f} MB)")
             else:
                 print(" No .mat files found")
             # List other files
             other_files = [f for f in data_dir.iterdir() if f.is_file() and not f.na
             if other_files:
                 print("\n0ther files:")
                 for other_file in sorted(other_files):
                     print(f" {other_file.name}")
         else:
             print(f"Data directory does not exist: {data_dir}")
             print("Please create the directory and download the dataset files.")
         print("\nDataset Download Information:")
         print("-" * 40)
```

```
print("All datasets are from the original SHMTools library (LA-CC-14-046)")
print("developed by Los Alamos National Laboratory.")
print("")
print("To obtain the datasets:")
print("1. Download from the original MATLAB SHMTools distribution")
print("2. Extract the .mat files from the Examples/ExampleData/ directory")
print(f"3. Place them in: {data_dir}")
print("")
print("See the README.md file in the data directory for detailed instruction
```

Data Directory: /Users/eric/repo/shm/shmtools-python/examples/data Directory exists: True

Files in data directory:

```
data3SS.mat ( 24.74 MB)
dataSensorDiagnostic.mat ( 0.06 MB)
data_CBM.mat ( 54.03 MB)
data_OSPExampleModal.mat ( 0.05 MB)
data_example_ActiveSense.mat ( 31.63 MB)
```

Other files:

.gitignore
README.md

Dataset Download Information:

All datasets are from the original SHMTools library (LA-CC-14-046) developed by Los Alamos National Laboratory.

To obtain the datasets:

- 1. Download from the original MATLAB SHMTools distribution
- 2. Extract the .mat files from the Examples/ExampleData/ directory
- 3. Place them in: /Users/eric/repo/shm/shmtools-python/examples/data

See the README.md file in the data directory for detailed instructions.

Summary

This notebook provides comprehensive dataset management utilities for SHMTools Python. Key takeaways:

Available Datasets

- 1. data3SS.mat: Primary 3-story structure dataset (25 MB)
- 2. data_CBM.mat: Condition-based monitoring rotating machinery (54 MB)
- 3. data_example_ActiveSense.mat: Guided wave measurements (32 MB)
- 4. dataSensorDiagnostic.mat: Sensor health monitoring (63 KB)
- 5. data_OSPExampleModal.mat: Modal analysis and sensor placement (50 KB)

Key Functions

- load_3story_data(): Primary structural dataset with detailed metadata
- load_cbm_data(): Rotating machinery with fault information
- load example data(type): Convenient preprocessing for specific examples
- validate_dataset_integrity(): Automated validation and checking
- check_data_availability(): Quick availability status

Usage Patterns

- Outlier Detection: Use load_example_data('pca') for preprocessed 3-story data
- Training/Testing: Split by damage states or use subset sampling
- Validation: Run integrity checks before starting analysis
- Exploration: Use metadata and descriptions for understanding structure

The enhanced data loading utilities provide comprehensive documentation, metadata, and validation capabilities that simplify working with SHMTools datasets while maintaining compatibility with the original MATLAB examples.