

# Exploratory Notebook Implementation Plan

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## Overview

This document outlines the implementation plan for an initial exploratory notebook that:

1. Fetches source FITS files from available APIs/integrations
2. Computes differences from previous versions (reference images)
3. Prepares data for anomaly detection in subsequent notebooks
4. Explores options for sourcing ZTF anomalies or creating synthetic anomalies

## Notebook Structure

### Phase 1: Data Acquisition

**Goal:** Fetch FITS files from multiple sources and establish a baseline dataset

#### 1.1 Source FITS File Acquisition

- **MAST API Integration** (`src/adapters/external/mast.py`)
  - Use `MASTClient.query_observations_by_position()` to find observations
  - Target sky regions:
    - Known transient regions (e.g., SN2011fe field, M101)
    - Deep field regions (COSMOS, GOODS-N)
    - Avoid galactic plane for cleaner images
  - Fetch observations from:
    - Pan-STARRS1 (PS1) - deep reference images
    - HST/JWST - high-quality deep images
    - TESS - time-series data
  - Download FITS files using `MASTClient.download_files()`
  - Store locally with metadata tracking
- **SkyView Integration** (`src/adapters/external/skyview.py`)
  - Use for on-demand image cutouts
  - Surveys: DSS2, SDSS, GALEX, 2MASS, WISE
  - Standardize image size (e.g., 240x240 pixels) and pixel scale
  - Create reference image library from multiple surveys
- **PS1 Cutout Helper** (`MASTClient.fetch_ps1_cutout()`)
  - Use for quick reference image fetching
  - Already implemented with FITS/JPEG fallback
  - Good for establishing baseline reference images

#### 1.2 Data Organization

- Create directory structure:

```
notebooks/data/exploratory/  
├── source_fits/           # Original FITS files  
│   ├── science/          # Current observations  
│   └── reference/        # Historical reference images  
├── processed/            # Processed images  
│   ├── aligned/          # WCS-aligned images  
│   └── normalized/       # Normalized images  
├── differences/          # Difference images  
│   ├── zogy/             # ZOGY algorithm results  
│   └── classic/          # Simple subtraction  
└── metadata/             # JSON metadata files
```

- Track metadata for each observation:
  - Observation ID, RA/Dec, observation time
  - Survey/instrument, filter band
  - File paths, image dimensions
  - Processing status flags

## Phase 2: Image Differencing

**Goal:** Compute differences between current and reference images

### 2.1 Image Preprocessing

- **FITS I/O** (`src/adapters/imaging/fits_io.py`)
  - Use `FITSProcessor.read_fits()` to load images
  - Extract WCS information for alignment
  - Extract metadata from headers
- **Image Alignment**
  - Use WCS information from `src/domains/preprocessing/processors/wcs_processor.py`
  - Align images to common pixel grid
  - Handle different pixel scales and orientations
  - Use `astropy` coordinate transformations
- **Normalization**
  - Background subtraction
  - Flux normalization
  - Handle different exposure times and zero points
  - Store normalization parameters

### 2.2 Difference Computation

- **ZOGY Algorithm**

(`src/domains/preprocessing/processors/astronomical_image_processing.py`)

- Use `ImageDifferencingProcessor.zogy_differencing()`
- Input: science image, reference image
- Optional: PSF models, noise maps
- Output: difference image + metrics (significance map, SNR)

- **Classic Differencing** (baseline)

- Simple subtraction for comparison
- Use `ImageDifferencingProcessor.classic_differencing()`

- **Difference Image Quality Assessment**

- Calculate noise properties
- Identify artifacts (cosmic rays, bad pixels)
- Compute quality metrics:
  - Noise level (std dev of background)
  - Dynamic range
  - Detection threshold (SNR)

## 2.3 Difference Image Storage

- Save difference images as FITS files
- Include metadata:
  - Original observation IDs
  - Processing parameters (PSF, noise estimates)
  - Quality metrics
  - Timestamps

## Phase 3: Anomaly Preparation

**Goal:** Prepare difference images for anomaly detection

### 3.1 Source Extraction

- Use `SEP` or `photutils` for source detection on difference images
- Extract candidate sources:
  - Position (x, y and RA/Dec)
  - Flux and significance (SNR)
  - Shape parameters (ellipticity, size)
  - Quality flags
- Filter candidates:
  - Minimum SNR threshold (e.g.,  $5\sigma$ )
  - Exclude known artifacts (bad pixels, diffraction spikes)
  - Size constraints (point sources vs extended)

### 3.2 Image Cutouts

- Extract cutouts around each candidate:
  - Science image cutout
  - Reference image cutout
  - Difference image cutout
  - Standard size (e.g., 64x64 or 128x128 pixels)
  - Include padding for context
- Store cutouts:
  - As individual FITS files or
  - In a multi-extension FITS file
  - Include metadata (position, SNR, timestamp)

### 3.3 Data Format for Next Notebook

- Create structured dataset:
  - NumPy arrays or PyTorch tensors
  - Image triplets: (science, reference, difference)
  - Labels: (anomaly flag, source type, confidence)
  - Metadata: (coordinates, timestamps, quality metrics)
- Export formats:
  - HDF5 file for large datasets
  - Pickle/numpy format for smaller sets
  - CSV metadata file
  - JSON configuration file

## Phase 4: Anomaly Data Sources

**Goal:** Explore options for real and synthetic anomalies

### 4.1 ZTF Database Options

#### Option A: ZTF Public Data Release

- **ZTF Public Data Portal** (IRSA)
  - Access: <https://irsa.ipac.caltech.edu/Missions/ztf.html>
  - Public data releases with 6-month delay
  - Difference images and alert packets
  - Light curves for confirmed transients
- **ZTF Alert Stream** (via brokers)
  - Kowalski database (MongoDB) - query interface
  - ZTF alert schema with image cutouts

- Requires: API access, database connection
- Filter by: object type, magnitude, location

- **ZTF Data Products**

- Difference images: science - reference
- Science image stamps
- Reference image stamps
- Metadata: RA/Dec, mag, filter, MJD

### Option B: ZTF via API

- Query ZTF public catalogs:
  - Use `astroquery` or direct API calls
  - Filter confirmed transients/supernovae
  - Download associated image stamps
  - Cross-match with known object catalogs

### Option C: ZTF Simulation/Reconstruction

- Use ZTF-like parameters:
  - Instrument characteristics (PSF, pixel scale)
  - Survey strategy (cadence, depth)
  - Generate synthetic ZTF-like images

## 4.2 Synthetic Anomaly Generation

### Leverage existing code (`standalone_training.py`)

- **SyntheticAnomalousDataset** (lines 359-508)
  - Already generates synthetic astronomical images
  - Anomaly types: transient, variable, supernova, asteroid
  - Configurable anomaly ratio and noise levels
- **Enhancement for Difference Images:**
  - Generate reference image (steady state)
  - Generate science image with injected anomaly
  - Compute difference image
  - Vary anomaly parameters:
    - Brightness (magnitude range)
    - Size (point source vs extended)
    - Position (random vs known locations)
    - Type (SN Ia, SN II, nova, etc.)
- **Realistic Anomaly Injection:**
  - Add real transient light curves
  - Use PSF models from real surveys
  - Include realistic noise characteristics

- Match ZTF-like image properties

### 4.3 Hybrid Approach

- Start with synthetic anomalies for validation
- Gradually incorporate real ZTF data as available
- Use synthetic data for:
  - Training anomaly detection models
  - Testing pipeline robustness
  - Handling edge cases
- Use real ZTF data for:
  - Validation and benchmarking
  - Understanding real-world challenges
  - Model generalization testing

## Phase 5: Data Validation and Quality Control

**Goal:** Ensure data quality before anomaly detection

### 5.1 Image Quality Checks

- Verify FITS file integrity
- Check image dimensions and data types
- Validate WCS solutions
- Check for NaN/inf values

### 5.2 Difference Image Validation

- Verify alignment quality (residuals)
- Check noise characteristics (should be Gaussian)
- Identify and flag artifacts
- Validate source extraction results

### 5.3 Metadata Completeness

- Ensure all required metadata present
- Check coordinate consistency
- Validate timestamps
- Verify cross-references (science ↔ reference)

## Phase 6: Visualization and Exploration

**Goal:** Visualize results and understand data characteristics

### 6.1 Image Visualization

- Display science, reference, and difference images
- Show source extraction results (overlay)
- Create comparison plots (ZOGY vs classic)

- Generate quality assessment plots

## 6.2 Statistical Analysis

- Noise distribution analysis
- Source detection statistics
- Anomaly candidate distribution
- Quality metrics distribution

## 6.3 Data Summary

- Generate summary statistics
- Create data quality report
- Document known issues/limitations
- Prepare data catalog for next notebook

# Implementation Details

## Notebook Sections

### 1. Setup and Configuration

- Import libraries
- Set up paths and directories
- Configure API clients
- Set random seeds for reproducibility

### 2. Data Acquisition

- Define target sky regions
- Fetch observations from MAST/SkyView
- Download FITS files
- Organize and catalog files

### 3. Image Processing

- Load FITS files
- Align images using WCS
- Normalize images
- Quality assessment

### 4. Image Differencing

- Compute ZOGY differences
- Compute classic differences (baseline)
- Calculate quality metrics
- Save difference images

### 5. Source Extraction

- Detect sources in difference images

- Filter candidates
- Extract cutouts
- Create candidate catalog

## 6. Anomaly Data Preparation

- Option 1: Query ZTF database (if available)
- Option 2: Generate synthetic anomalies
- Option 3: Hybrid approach
- Prepare training/validation dataset

## 7. Data Export

- Save processed images
- Export metadata
- Create dataset for next notebook
- Generate summary report

## 8. Visualization

- Display example images
- Show difference images
- Plot source detection results
- Create quality assessment plots

# Technical Requirements

## Dependencies

- Existing AstrlD modules:
  - `src/adapters/external/mast.py` - MAST API client
  - `src/adapters/external/skyview.py` - SkyView client
  - `src/adapters/imaging/fits_io.py` - FITS I/O
  - `src/domains/preprocessing/processors/astronomical_image_processing.py` - ZOGY differencing
  - `src/domains/preprocessing/processors/fits_processing.py` - FITS processing
- External packages:
  - `astropy` - FITS, WCS, coordinates
  - `astroquery` - MAST queries
  - `numpy`, `matplotlib` - Data handling, visualization
  - `sep` or `photutils` - Source extraction
  - `scipy` - Image processing

## Configuration Options

- Target sky regions (RA/Dec, radius)
- Survey selection (MAST missions, SkyView surveys)
- Image size and pixel scale
- Differencing algorithm parameters
- Source detection thresholds



- Anomaly generation parameters (if synthetic)

## Deliverables

### 1. Jupyter Notebook ([notebooks/exploratory\\_data\\_preparation.ipynb](#))

- Complete implementation of all phases
- Well-documented with markdown cells
- Clear section organization

### 2. Data Directory Structure

- Organized FITS files
- Processed images
- Difference images
- Metadata files

### 3. Dataset for Next Notebook

- Processed image triplets (science, reference, difference)
- Candidate source catalog
- Anomaly labels (if available)
- Metadata and configuration files

### 4. Documentation

- README for the notebook
- Data format specification
- Known issues and limitations
- Next steps documentation

## Next Steps (Subsequent Notebook)

The output from this exploratory notebook will feed into:

- **Anomaly Detection Notebook**
  - Load prepared dataset
  - Apply U-Net model for anomaly detection
  - Evaluate detection performance
  - Visualize results
  - Compare with ground truth (if available)

## Notes

- Start small: Begin with 10-20 observations to validate pipeline
- Use existing code: Leverage [standalone\\_training.py](#) for synthetic data
- API rate limits: Be mindful of MAST/SkyView rate limits
- Data storage: Consider disk space for FITS files
- Reproducibility: Save all random seeds and configuration
- Error handling: Robust error handling for API failures

- Progress tracking: Log progress for long-running operations