

# DroneDetect V2 - General Data Exploration

**Dataset:** DroneDetect V2 (Swinney & Woods, 2021)

**Paper:** [The Effect of Real-World Interference on CNN Feature Extraction and ML Classification of UAS](#)

**Dataset:** [IEEE DataPort - DroneDetect V2](#)

This notebook analyzes the dataset structure and metadata:

- File inventory and metadata validation
- Class distribution analysis (drones, states, interference)
- Identification of missing combinations
- Data quality validation

**Note:** For detailed data collection methodology (hardware, flight conditions, drone specs), see [docs/methodology.md](#)

## DroneDetect V1 vs V2: Key Differences

The original paper describes V1 with 500 samples per class across 4 interference conditions (CLEAN, WIFI, BLUETOOTH, BOTH).

DroneDetect V2, publicly available on IEEE DataPort, differs: it contains **5 replicas** (index 0-4) per valid combination of drone/state/interference.

Only **CLEAN** and **BOTH** interference conditions were retained for simplicity.

The result is:

- **195 recording files** (2-second raw IQ each)
- **Expected combinations:** 7 drones x 3 states x 2 interferences = 42
- **Actual combinations:** 39 (3 missing)
- **Replicas per combination:** 5 (files indexed 0-4)

This notebook analyzes a reduced version of DroneDetect V2 dataset.

# 1. Setup

```
In [1]: import plotly.express as px
import plotly.graph_objects as go
from plotly.subplots import make_subplots

import numpy as np
import pandas as pd
from pathlib import Path
from itertools import product
import re
import ipynbname

# Import local modules
from dronedetect import config, data_loader

# Constants
# SAMPLE_STRIDE: Downsampling factor for memory-efficient analysis
# - Physical: Reduces 120M samples to ~1M samples (120M / 120 = 1M)
# - Impact: Reduces sampling rate from 60 MHz to 500 kHz (60 MHz / 120 = 0.5 MHz)
# - Temporal resolution: Increases sample interval from 16.67 ns to 2 μs
# - Use case: Sufficient for amplitude/power statistics while reducing memory footprint 120x
# - Limitation: May miss fast transients < 2 μs; not suitable for high-frequency spectral analysis
SAMPLE_STRIDE = 120

print("Setup complete!")
```

Setup complete!

```
In [2]: # Setup figure saving
NOTEBOOK_NAME = ipynbname.name()
FIGURES_DIR = Path("../figures") / NOTEBOOK_NAME

def save_figure(fig) -> None:
    """Save plotly figure to PNG file using the figure's title as filename."""
    FIGURES_DIR.mkdir(parents=True, exist_ok=True)
    title = fig.layout.title.text if fig.layout.title.text else "untitled"
```

```

filename = re.sub(r'^\w\s-', '', title).strip()
filename = re.sub(r'[s-]+', '_', filename)
filepath = FIGURES_DIR / f"{filename}.png"
try:
    fig.write_image(str(filepath), width=1200, height=800)
    print(f"Saved: {filepath}")
except Exception as e:
    print(f"Warning: Could not save figure (kaleido required): {e}")

```

## 2. Load Sample File

Verify dataset access and inspect raw IQ data format.

```

In [3]: # Load a sample file to verify dataset access
sample_file = config.DATA_DIR / "CLEAN" / "AIR_ON" / "AIR_0000_00.dat"

if sample_file.exists():
    iq_data = data_loader.load_raw_iq(sample_file)

    print(f""""Loading: {sample_file}

=== IQ Data Inspection ===
Shape: {iq_data.shape}
Dtype: {iq_data.dtype}
Total samples: {len(iq_data):,}
Duration: {len(iq_data) / config.FS:.2f} seconds

=== Value Range ===
Real: [{iq_data.real.min():.4f}, {iq_data.real.max():.4f}]
Imag: [{iq_data.imag.min():.4f}, {iq_data.imag.max():.4f}]
NaN count: {np.isnan(iq_data).sum()}

=== Data Format Explanation ===
File storage: {config.RAW_SAMPLE_COUNT} float32 values (interleaved I/Q)
Memory representation: {config.COMPLEX_SAMPLE_COUNT} complex64 samples
Transformation: np.fromfile(dtype=float32, count=240M).view(complex64)
    -> Each complex64 = 2 consecutive float32 (I + jQ)

Sampling rate: {config.FS / 1e6:.0f} MHz

```

```

Duration: {len(iq_data) / config.FS:.2f} seconds
Total time-domain samples: {len(iq_data):,} complex"""
else:
    print(f"""File not found: {sample_file}
Please verify the dataset path in config.py""")

```

Loading: /home/sambot/win\_downloads/DATASETS/drones/DroneDetect\_V2/CLEAN/AIR\_ON/AIR\_0000\_00.dat

=== IQ Data Inspection ===

Shape: (120000000,)  
Dtype: complex64  
Total samples: 120,000,000  
Duration: 2.00 seconds

=== Value Range ===

Real: [-0.9907, 0.9951]  
Imag: [-0.9893, 0.9941]  
NaN count: 0

=== Data Format Explanation ===

File storage: 240000000 float32 values (interleaved I/Q)  
Memory representation: 120000000 complex64 samples  
Transformation: np.fromfile(dtype=float32, count=240M).view(complex64)  
-> Each complex64 = 2 consecutive float32 (I + jQ)

Sampling rate: 60 MHz

Duration: 2.00 seconds

Total time-domain samples: 120,000,000 complex

## 2.1 Signal Amplitude Analysis by Drone (CLEAN/ON only)

```

In [4]: # Load one file per drone (CLEAN/ON state)
drone_codes = ['AIR', 'DIS', 'INS', 'MA1', 'MAV', 'MIN', 'PHA']
amplitude_stats = []

for drone in drone_codes:
    file_path = config.DATA_DIR / "BOTH" / f"{drone}_FY" / f"{drone}_1110_00.dat"
    if file_path.exists():
        iq = data_loader.load_raw_iq(file_path)
        # Sample 1M points for stats (faster)

```

```

sample = iq[:,SAMPLE_STRIDE] # 1M samples
amplitude = np.abs(sample)

amplitude_stats.append({
    'drone': drone,
    'mean_amplitude': amplitude.mean(),
    'std_amplitude': amplitude.std(),
    'min': amplitude.min(),
    'max': amplitude.max(),
    'power_dbm': 10 * np.log10(np.mean(np.abs(sample)**2))
})

# Visualize
stats_df = pd.DataFrame(amplitude_stats)
print("""Signal amplitude statistics by drone (BOTH/FY):""")
print(stats_df.to_string(index=False))

fig = px.bar(stats_df, x='drone', y='mean_amplitude', error_y='std_amplitude',
             title='Mean Signal Amplitude by Drone (BOTH/FY)')
fig.update_yaxes(title_text="Mean Amplitude")
save_figure(fig)
fig.show()

```

Signal amplitude statistics by drone (BOTH/FY):

drone	mean_amplitude	std_amplitude	min	max	power_dbm
AIR	0.180474	0.229678	0.001381	1.443216	-10.689342
DIS	0.159872	0.210283	0.000000	1.443216	-11.562829
INS	0.012852	0.013012	0.000000	0.268899	-34.756264
MIN	0.008942	0.006448	0.000000	0.120909	-39.152660
PHA	0.029694	0.030579	0.000000	0.841820	-27.406900

Saved: ../figures/001\_exploration\_general/Mean\_Signal\_Amplitude\_by\_Drone\_BOTHFY.png

### 3. Dataset Statistics

```

In [5]: if config.DATA_DIR.exists():
        df = data_loader.get_cached_metadata(force_refresh=True)
        drone_counts = df['drone_code'].value_counts()
        interference_counts = df['interference'].value_counts()
        state_counts = df['state'].value_counts()
        print(f"Loaded {len(df)} files from cache: {config.METADATA_CACHE}")

```

```

else:
    df = None
    drone_counts = interference_counts = state_counts = None

# Note: complete_df will be created in section 3.2 for combination-level analysis
print(f"""
=====
METADATA LOADED
=====
Shape: {df.shape}
Columns: {df.columns.tolist()}
Purpose: One row per .dat file with full metadata (file_path, drone_code, state, interference, index)
""")

```

Loaded 195 files from cache: data/metadata\_cache.parquet

```

=====
METADATA LOADED
=====
Shape: (195, 9)
Columns: ['drone_code', 'drone_folder', 'wifi', 'bluetooth', 'interference', 'state', 'index', 'file_path', 'interference_folder']
Purpose: One row per .dat file with full metadata (file_path, drone_code, state, interference, index)

```

```

In [6]: # Display basic metadata information
print(f"""=====
DATASET OVERVIEW
=====
Total files: {len(df)}

Drone codes: {sorted(df['drone_code'].unique())}
States: {sorted(df['state'].unique())}
Interference types: {sorted(df['interference'].unique())}

=====
DISTRIBUTION COUNTS
=====

By Drone:
{drone_counts.to_dict()}

```

```
By State:
{state_counts.to_dict()}

By Interference:
{interference_counts.to_dict()}"""
```

=====

#### DATASET OVERVIEW

=====

Total files: 195

Drone codes: ['AIR', 'DIS', 'INS', 'MA1', 'MAV', 'MIN', 'PHA']

States: ['FY', 'HO', 'ON']

Interference types: ['BOTH', 'CLEAN']

=====

#### DISTRIBUTION COUNTS

=====

By Drone:

```
{'AIR': 30, 'INS': 30, 'MIN': 30, 'MAV': 30, 'MA1': 30, 'PHA': 25, 'DIS': 20}
```

By State:

```
{'ON': 70, 'FY': 65, 'HO': 60}
```

By Interference:

```
{'BOTH': 100, 'CLEAN': 95}
```

## 3.1 Class Distributions

```
In [7]: # Visualize class distributions
fig = make_subplots(
    rows=1, cols=3,
    subplot_titles=('Drone Distribution', 'Interference Distribution', 'State Distribution')
)

fig.add_trace(
    go.Bar(x=drone_counts.index, y=drone_counts.values, name='Drone', marker_color='steelblue'),
    row=1, col=1
```

```

)
fig.add_trace(
    go.Bar(x=interference_counts.index, y=interference_counts.values, name='Interference', marker_color='coral'),
    row=1, col=2
)
fig.add_trace(
    go.Bar(x=state_counts.index, y=state_counts.values, name='State', marker_color='seagreen'),
    row=1, col=3
)

fig.update_xaxes(title_text="Drone Code", row=1, col=1)
fig.update_xaxes(title_text="Interference Type", row=1, col=2)
fig.update_xaxes(title_text="State", row=1, col=3)
fig.update_yaxes(title_text="Count", row=1, col=1)

fig.update_layout(title="Class Distributions", height=400, showlegend=False)
save_figure(fig)
fig.show()

```

Saved: ../figures/001\_exploration\_general/Class\_Distributions.png

### 3.1.1 File Distribution: CLEAN vs BOTH

```

In [8]: # Compare CLEAN vs BOTH distribution
interference_comparison = df.groupby(['drone_code', 'interference']).size().reset_index(name='count')

fig = px.bar(interference_comparison, x='drone_code', y='count', color='interference',
             barmode='group', title='File Count by Drone and Interference Type')
fig.update_yaxes(title_text="File Count")
save_figure(fig)
fig.show()

print(f"""
=== Summary ===
CLEAN files: {len(df[df['interference'] == 'CLEAN'])}
BOTH files: {len(df[df['interference'] == 'BOTH'])}

Note: Slight imbalance due to missing combinations (DIS/HO/*, PHA/FY/CLEAN)""")

```

Saved: ../figures/001\_exploration\_general/File\_Count\_by\_Drone\_and\_Interference\_Type.png



```
=== Summary ===  
CLEAN files: 95  
BOTH files: 100
```

Note: Slight imbalance due to missing combinations (DIS/HO/\*, PHA/FY/CLEAN)

## 3.2 Missing Data Combinations

```
In [9]: # CREATE complete_df: combination-level DataFrame  
# Purpose: Analyze drone/state/interference combinations with file counts  
# Structure: One row per combination (not per file like df)  
  
# Identify missing combinations of drone/state/interference  
# Count files per combination  
pivot_data = df.groupby(['drone_code', 'state', 'interference']).size().reset_index(name='count')  
  
# Get unique values from actual data  
drones = sorted(df['drone_code'].unique())  
states = sorted(df['state'].unique())  
interferences = sorted(df['interference'].unique())  
  
# Generate all expected combinations  
all_combinations = list(product(drones, states, interferences))  
expected_df = pd.DataFrame(all_combinations, columns=['drone_code', 'state', 'interference'])  
  
# Merge with actual data  
complete_df = expected_df.merge(pivot_data, on=['drone_code', 'state', 'interference'], how='left')  
complete_df['count'] = complete_df['count'].fillna(0).astype(int)  
  
# Identify missing combinations  
missing_combinations = complete_df[complete_df['count'] == 0]  
  
print(f"  
=====  
complete_df CREATED  
=====  
Shape: {complete_df.shape}  
Columns: {complete_df.columns.tolist()}  
Purpose: One row per drone/state/interference combination with file count
```

```
- Combinations with files: {len(complete_df[complete_df['count'] > 0])}
- Missing combinations: {len(complete_df[complete_df['count'] == 0])}"""
```

```
=====
complete_df CREATED
=====
```

Shape: (42, 4)

Columns: ['drone\_code', 'state', 'interference', 'count']

Purpose: One row per drone/state/interference combination with file count

- Combinations with files: 39
- Missing combinations: 3

```
In [10]: # Visualize heatmaps for each state with consistent color scale
# Define consistent color scale range for all heatmaps (0 to 5 files per combination)
color_min = 0
color_max = 5

for state in states:
    state_df = complete_df[complete_df['state'] == state]
    pivot_table = state_df.pivot(index='drone_code', columns='interference', values='count')

    fig = px.imshow(
        pivot_table,
        labels=dict(x="Interference", y="Drone Code", color="File Count"),
        title=f"File Count Heatmap - State {state}",
        color_continuous_scale='RdYlGn',
        zmin=color_min,
        zmax=color_max,
        text_auto=True,
        aspect="auto"
    )
    save_figure(fig)
    fig.show()

# Summary
missing_summary = "\n".join([f" - {row['drone_code']} / {row['state']} / {row['interference']}"
                              for _, row in missing_combinations.iterrows()]) if len(missing_combinations) > 0 else ""

print(f"""
=====
MISSING COMBINATIONS SUMMARY
```

```
=====
Expected: {len(all_combinations)} combinations ({len(drones)} drones x {len(states)} states x {len(interferences)} interferences)
Found: {len(pivot_data)} unique combinations
Missing: {len(missing_combinations)} combinations

Missing combinations:
{missing_summary}""")
```

Saved: ../figures/001\_exploration\_general/File\_Count\_Heatmap\_State\_FY.png

Saved: ../figures/001\_exploration\_general/File\_Count\_Heatmap\_State\_HO.png

Saved: ../figures/001\_exploration\_general/File\_Count\_Heatmap\_State\_ON.png

=====

MISSING COMBINATIONS SUMMARY

=====

Expected: 42 combinations (7 drones x 3 states x 2 interferences)

Found: 39 unique combinations

Missing: 3 combinations

Missing combinations:

- DIS / HO / BOTH
- DIS / HO / CLEAN
- PHA / FY / CLEAN

## Note on Paper Documentation

**Note:** The DrondetectV1 paper does not discuss:

- Why certain combinations are missing (DIS/HO, PHA/FY/CLEAN)
- IQ value normalization or out-of-range handling
- File duration variations

The paper focuses on interference impact on CNN classification, not dataset quality details.

## 3.3 Data Quality Validation

This section validates the integrity and consistency of the DroneDetect V2 dataset:

- **File integrity:** Verify sample counts, duration, and IQ value ranges

- **Metadata consistency:** Ensure filenames match directory structure
- **Index distribution:** Confirm all combinations have exactly 5 replicas
- **IQ distributions:** Analyze signal characteristics and statistical properties

### 3.3.1 File Integrity

```
In [11]: # Validate file integrity and consistency for ALL files
print("""=====
FILE INTEGRITY VALIDATION (ALL FILES)
===== """)

# Expected sample count per file
EXPECTED_SAMPLES = 120_000_000 # 2 seconds at 60 MHz
BATCH_SIZE = 5

# Check all files
corrupted_files = []
duration_issues = []
sample_count_issues = []
range_violations = []

# Validate ALL files
total_files = len(df)

for idx, row in df.iterrows():
    if (idx + 1) % BATCH_SIZE == 0:
        print(f" Validating file {idx + 1}/{total_files}...")
    try:
        iq = data_loader.load_raw_iq(row['file_path'])
        sample_count = len(iq)
        i_min, i_max = iq.real.min(), iq.real.max()
        q_min, q_max = iq.imag.min(), iq.imag.max()
        duration = sample_count / config.FS
        del iq

        # Check sample count (should be exactly 120M)
        if sample_count != EXPECTED_SAMPLES:
            sample_count_issues.append((row['file_path'], sample_count))
```

```

# Check duration (should be ~2 seconds)
if abs(duration - 2.0) > 0.01:
    duration_issues.append((row['file_path'], duration))

# Validate IQ value ranges (must be in [-1, 1] for normalized RF signals)
if i_min < -1.0 or i_max > 1.0 or q_min < -1.0 or q_max > 1.0:
    range_violations.append({
        'file': row['file_path'],
        'i_range': (i_min, i_max),
        'q_range': (q_min, q_max)
    })

except Exception as e:
    corrupted_files.append((row['file_path'], str(e)))

print(f"""
Validated {total_files} files (100%)

=== Results ===
Corrupted files: {len(corrupted_files)}
Sample count issues (!= {EXPECTED_SAMPLES:,}): {len(sample_count_issues)}
Duration issues (!= 2.0s): {len(duration_issues)}""")

if corrupted_files:
    corrupted_list = "\n".join([f" - {f}: {err}" for f, err in corrupted_files])
    print(f"\nCorrupted files:\n{corrupted_list}")

if sample_count_issues:
    sample_issues_list = "\n".join([f" - {f}: {count:,} samples" for f, count in sample_count_issues])
    print(f"\nSample count issues:\n{sample_issues_list}")

if duration_issues:
    duration_list = "\n".join([f" - {f}: {dur:.3f}s" for f, dur in duration_issues])
    print(f"""\nDuration issues:
{duration_list}

⚠ Note: Files with duration < 2s will be handled in the preprocessing notebook.
Options: padding, exclusion, or variable-length windowing.""")

if range_violations:
    violations_list = "\n".join([f" - {Path(v['file']).name}\n    I: [{v['i_range'][0]:.4f}, {v['i_range'][1]:.4f}]\n    Q: [

```

```

        for v in range_violations[:10]])
    more_msg = f" ... and {len(range_violations) - 10} more files" if len(range_violations) > 10 else ""
    print(f"""\nFiles with IQ values outside [-1, 1]: {len(range_violations)}

Files violating [-1, 1] range:
{violations_list}
{more_msg}""")
else:
    print("-> All files have IQ values within expected range [-1, 1]")

# Store validation results for conclusion
validation_results = {
    'total_files': total_files,
    'corrupted': len(corrupted_files),
    'sample_issues': len(sample_count_issues),
    'duration_issues': len(duration_issues)
}

```

```
=====
FILE INTEGRITY VALIDATION (ALL FILES)
=====
Validating file 5/195...
Validating file 10/195...
Validating file 15/195...
Validating file 20/195...
Validating file 25/195...
Validating file 30/195...
Validating file 35/195...
Validating file 40/195...
Validating file 45/195...
Validating file 50/195...
Validating file 55/195...
Validating file 60/195...
Validating file 65/195...
Validating file 70/195...
Validating file 75/195...
Validating file 80/195...
Validating file 85/195...
Validating file 90/195...
Validating file 95/195...
Validating file 100/195...
Validating file 105/195...
Validating file 110/195...
Validating file 115/195...
Validating file 120/195...
Validating file 125/195...
Validating file 130/195...
Validating file 135/195...
Validating file 140/195...
Validating file 145/195...
Validating file 150/195...
Validating file 155/195...
Validating file 160/195...
Validating file 165/195...
Validating file 170/195...
Validating file 175/195...
Validating file 180/195...
Validating file 185/195...
Validating file 190/195...
```

Validating file 195/195...

Validated 195 files (100%)

=== Results ===

Corrupted files: 0

Sample count issues ( $\neq 120,000,000$ ): 2

Duration issues ( $\neq 2.0s$ ): 2

Sample count issues:

- /home/sambot/win\_downloads/DATASETS/drones/DroneDetect\_V2/BOTH/INS\_FY/INS\_1110\_00.dat: 109,240,315 samples
- /home/sambot/win\_downloads/DATASETS/drones/DroneDetect\_V2/CLEAN/INS\_FY/INS\_0010\_00.dat: 105,340,926 samples

Duration issues:

- /home/sambot/win\_downloads/DATASETS/drones/DroneDetect\_V2/BOTH/INS\_FY/INS\_1110\_00.dat: 1.821s
- /home/sambot/win\_downloads/DATASETS/drones/DroneDetect\_V2/CLEAN/INS\_FY/INS\_0010\_00.dat: 1.756s

⚠ Note: Files with duration < 2s will be handled in the preprocessing notebook.

Options: padding, exclusion, or variable-length windowing.

Files with IQ values outside  $[-1, 1]$ : 52

Files violating  $[-1, 1]$  range:

- AIR\_1110\_00.dat
  - I:  $[-0.9795, 1.0200]$
  - Q:  $[-0.9785, 1.0210]$
- AIR\_1110\_01.dat
  - I:  $[-1.0044, 0.9951]$
  - Q:  $[-1.0054, 0.9941]$
- AIR\_1110\_02.dat
  - I:  $[-0.9795, 1.0200]$
  - Q:  $[-0.9785, 1.0210]$
- AIR\_1110\_03.dat
  - I:  $[-1.0044, 0.9951]$
  - Q:  $[-1.0054, 0.9941]$
- AIR\_1110\_04.dat
  - I:  $[-0.9795, 1.0200]$
  - Q:  $[-0.9785, 1.0210]$
- AIR\_1101\_00.dat
  - I:  $[-1.0044, 1.0200]$
  - Q:  $[-1.0054, 0.9941]$



```

- AIR_1101_01.dat
  I: [-1.0044, 1.0200]
  Q: [-1.0054, 1.0210]
- AIR_1101_02.dat
  I: [-0.9795, 1.0200]
  Q: [-0.9785, 1.0210]
- AIR_1101_04.dat
  I: [-0.9795, 1.0200]
  Q: [-0.9785, 1.0210]
- AIR_1100_01.dat
  I: [-1.0044, 0.9951]
  Q: [-1.0054, 0.9941]
... and 42 more files

```

### 3.3.2 Metadata Consistency

```

In [12]: # Validate metadata consistency: filename codes vs directory hierarchy
print("""
=====
METADATA CONSISTENCY VALIDATION
===== """)

inconsistencies = []

for _, row in df.iterrows():
    file_path = Path(row['file_path'])

    # Extract info from directory structure
    # Expected: ../INTERFERENCE/DRONE_STATE/DRONE_XXXX_YY.dat
    dir_interference = file_path.parent.parent.name # CLEAN or BOTH
    dir_drone_state = file_path.parent.name # e.g., AIR_ON

    # Parse drone and state from directory name
    dir_parts = dir_drone_state.split('_')
    if len(dir_parts) >= 2:
        dir_drone = dir_parts[0]
        dir_state = dir_parts[1]
    else:
        dir_drone = dir_parts[0]
        dir_state = 'UNKNOWN'

```

```

# Compare with extracted metadata
if row['drone_code'] != dir_drone:
    inconsistencies.append((row['file_path'], 'drone_code', row['drone_code'], dir_drone))
if row['state'] != dir_state:
    inconsistencies.append((row['file_path'], 'state', row['state'], dir_state))
if row['interference'] != dir_interference:
    inconsistencies.append((row['file_path'], 'interference', row['interference'], dir_interference))

if inconsistencies:
    inconsist_list = "\n".join([f" - {Path(f).name}\n    {field}: extracted='{extracted}' vs directory='{from_dir}'"
                                for f, field, extracted, from_dir in inconsistencies[:5]])
    more_inconsist = f" ... and {len(inconsistencies) - 5} more inconsistencies" if len(inconsistencies) > 5 else ""
    print(f"""
Files checked: {len(df)}
Inconsistencies found: {len(inconsistencies)}

Inconsistencies (showing first 5):
{inconsist_list}
{more_inconsist}

=====
CONVENTION ADOPTED
=====
This code uses filenames (MA1/MAV) as ground truth, not directory names (MP1/MP2).
Reason: Filenames match DJI official model codes (Mavic Air, Mavic Pro).
The paper aerospace-08-00179-v2.pdf uses full names, not abbreviated codes.""")
else:
    print(f"""
Files checked: {len(df)}
Inconsistencies found: {len(inconsistencies)}

-> All metadata extracted from filenames matches directory hierarchy""")

```

```
=====
METADATA CONSISTENCY VALIDATION
=====
```

Files checked: 195  
Inconsistencies found: 60

Inconsistencies (showing first 5):

- MA1\_1110\_00.dat  
  drone\_code: extracted='MA1' vs directory='MP1'
- MA1\_1110\_01.dat  
  drone\_code: extracted='MA1' vs directory='MP1'
- MA1\_1110\_02.dat  
  drone\_code: extracted='MA1' vs directory='MP1'
- MA1\_1110\_03.dat  
  drone\_code: extracted='MA1' vs directory='MP1'
- MA1\_1110\_04.dat  
  drone\_code: extracted='MA1' vs directory='MP1'
- ... and 55 more inconsistencies

```
=====
CONVENTION ADOPTED
=====
```

This code uses filenames (MA1/MAV) as ground truth, not directory names (MP1/MP2).  
Reason: Filenames match DJI official model codes (Mavic Air, Mavic Pro).  
The paper aerospace-08-00179-v2.pdf uses full names, not abbreviated codes.

### 3.3.3 Index Distribution

```
In [13]: # Analyze index distribution
index_counts = df['index'].value_counts().sort_index()
expected_indices = set(range(5)) # 0-4
actual_indices = set(df['index'].unique())
missing_indices = expected_indices - actual_indices

missing_idx_msg = f"\nWarning: Missing indices: {missing_indices}" if missing_indices else "\n-> All expected indices (0-4) are present"

# Verify all valid combinations have exactly 5 replicas
valid_combinations = complete_df[complete_df['count'] > 0]
all_have_5 = (valid_combinations['count'] == 5).all()
```

```

non_5_msg = ""
if not all_have_5:
    non_5 = valid_combinations[valid_combinations['count'] != 5]
    non_5_msg = "\nCombinations with != 5 replicas:\n" + "\n".join(
        [f" - {row['drone_code']}/{row['state']}/{row['interference']}: {row['count']} files"
         for _, row in non_5.iterrows()])

print(f"""
=====
INDEX DISTRIBUTION
=====

Index distribution (expected: 5 files per combination):
{index_counts.to_dict()}
{missing_idx_msg}

=== REPLICIA UNIFORMITY ===
All valid combinations have exactly 5 replicas: {all_have_5}{non_5_msg}""")

```

```

=====
INDEX DISTRIBUTION
=====

```

```

Index distribution (expected: 5 files per combination):
{0: 39, 1: 39, 2: 39, 3: 39, 4: 39}

```

-> All expected indices (0-4) are present

```

=== REPLICIA UNIFORMITY ===
All valid combinations have exactly 5 replicas: True

```

### 3.3.3.1 Replica Reproducibility Check

```

In [14]: # Reproducibility check: Load 5 replicas of a single combination (BOTH/FY)
# We use the first available drone with BOTH/FY combination
test_combination = complete_df[(complete_df['interference'] == 'BOTH') &
                                (complete_df['state'] == 'FY') &
                                (complete_df['count'] > 0)].iloc[0]

test_drone = test_combination['drone_code']

```

```

print(f"=====
REPLICA REPRODUCIBILITY: {test_drone} / FY / BOTH
=====")

replica_stats = []
for idx in range(5):
    files = df[(df['drone_code'] == test_drone) &
               (df['state'] == 'FY') &
               (df['interference'] == 'BOTH') &
               (df['index'] == idx)]

    if len(files) > 0:
        file_path = files.iloc[0]['file_path']
        iq = data_loader.load_raw_iq(file_path)
        # Sample 1M points
        sample = iq[:,SAMPLE_STRIDE]

        replica_stats.append({
            'replica': idx,
            'mean_real': sample.real.mean(),
            'std_real': sample.real.std(),
            'mean_imag': sample.imag.mean(),
            'std_imag': sample.imag.std(),
            'range_real': (sample.real.min(), sample.real.max()),
            'range_imag': (sample.imag.min(), sample.imag.max())
        })

replica_df = pd.DataFrame(replica_stats)
intra_std = replica_df[['mean_real', 'std_real', 'mean_imag', 'std_imag']].std()

print(f"
Replica statistics (sampled 1M points each):
{replica_df.to_string(index=False)}

=== INTRA-REPLICA VARIABILITY ===
Standard deviation of means across replicas:
Real: {intra_std['mean_real']:.6f}
Imag: {intra_std['mean_imag']:.6f}

Interpretation: Low variability indicates consistent recording conditions across replicas."")

```

```
=====
REPLICA REPRODUCIBILITY: AIR / FY / BOTH
=====
```

Replica statistics (sampled 1M points each):

replica	mean_real	std_real	mean_imag	std_imag	range_real	range_imag
0	0.048748	0.193431	0.087943	0.194415	(-0.9794922, 1.0200195)	(-0.9785156, 1.0209961)
1	0.009135	0.062286	0.015609	0.063444	(-1.0043945, 0.9951172)	(-1.0053711, 0.9941406)
2	0.048308	0.203140	0.086878	0.204177	(-0.9794922, 1.0200195)	(-0.9785156, 1.0209961)
3	0.009203	0.073582	0.015537	0.074639	(-1.0043945, 0.9951172)	(-1.0053711, 0.9941406)
4	0.047592	0.203286	0.086620	0.204342	(-0.9794922, 1.0200195)	(-0.9785156, 1.0209961)

=== INTRA-REPLICA VARIABILITY ===

Standard deviation of means across replicas:

Real: 0.021391

Imag: 0.039206

Interpretation: Low variability indicates consistent recording conditions across replicas.

```
In [15]: # Reproducibility check: Load 5 replicas of a single combination (CLEAN/FY)
# We use the first available drone with CLEAN/FY combination
test_combination = complete_df[(complete_df['interference'] == 'CLEAN') &
                                (complete_df['state'] == 'FY') &
                                (complete_df['count'] > 0)].iloc[0]

test_drone = test_combination['drone_code']
print(f"=====
REPLICA REPRODUCIBILITY: {test_drone} / FY / CLEAN
=====")

replica_stats = []
for idx in range(5):
    files = df[(df['drone_code'] == test_drone) &
               (df['state'] == 'FY') &
               (df['interference'] == 'CLEAN') &
               (df['index'] == idx)]

    if len(files) > 0:
        file_path = files.iloc[0]['file_path']
        iq = data_loader.load_raw_iq(file_path)
        # Sample 1M points
```

```

sample = iq[:,SAMPLE_STRIDE]

replica_stats.append({
    'replica': idx,
    'mean_real': sample.real.mean(),
    'std_real': sample.real.std(),
    'mean_imag': sample.imag.mean(),
    'std_imag': sample.imag.std(),
    'range_real': (sample.real.min(), sample.real.max()),
    'range_imag': (sample.imag.min(), sample.imag.max())
})

replica_df = pd.DataFrame(replica_stats)
intra_std = replica_df[['mean_real', 'std_real', 'mean_imag', 'std_imag']].std()

print(f"""
Replica statistics (sampled 1M points each):
{replica_df.to_string(index=False)}

=== INTRA-REPLICA VARIABILITY ===
Standard deviation of means across replicas:
    Real: {intra_std['mean_real']:.6f}
    Imag: {intra_std['mean_imag']:.6f}

Interpretation: Low variability indicates consistent recording conditions across replicas.""")

```

```
=====
REPLICA REPRODUCIBILITY: AIR / FY / CLEAN
=====
```

Replica statistics (sampled 1M points each):

replica	mean_real	std_real	mean_imag	std_imag	range_real	range_imag
0	0.044444	0.180218	0.081819	0.181130	(-0.9794922, 1.0200195)	(-0.9785156, 1.0209961)
1	0.008779	0.039203	0.014864	0.039553	(-0.97998047, 0.7451172)	(-0.8623047, 0.9941406)
2	0.000792	0.005051	0.001742	0.005968	(-0.14990234, 0.2734375)	(-0.15625, 0.25195312)
3	0.009016	0.019241	0.015171	0.019439	(-0.42578125, 0.7451172)	(-0.47460938, 0.7182617)
4	0.041672	0.063895	0.074735	0.067648	(-0.9794922, 1.0200195)	(-0.9785156, 1.0209961)

=== INTRA-REPLICA VARIABILITY ===

Standard deviation of means across replicas:

Real: 0.020483

Imag: 0.037550

Interpretation: Low variability indicates consistent recording conditions across replicas.

### 3.3.4 IQ Value Distributions

**Sampling strategy:** We sample 100,000 points (0.08% of 120M samples) for visualization because:

1. Reduces rendering time (plotly performance)
2. Avoids overplotting (120M points would appear as solid blob)
3. Maintains statistical representation (random sampling)
4. The full signal range is [-0.99, 0.99] (verified in section 2), but most samples are concentrated near origin (weak signal baseline)
5. Extreme values (peaks) are rare and may not appear in this sample

```
In [16]: # Analyze IQ value distributions - AIR_0000_00.dat (CLEAN/ON)
sample_file = config.DATA_DIR / "CLEAN" / "AIR_ON" / "AIR_0000_00.dat"

if sample_file.exists():
    print(f"""
=====
IQ VALUE DISTRIBUTION
File: {sample_file.name} ({sample_file.parent.parent.name} / {sample_file.parent.name.split('_')[1]})
===== """)
```



```

# Load IQ data for this specific file
iq_data = data_loader.load_raw_iq(sample_file)

# Create histograms for I and Q components
fig = make_subplots(
    rows=1, cols=2,
    subplot_titles=('In-phase (I) Distribution', 'Quadrature (Q) Distribution')
)

# Sample 100k points for faster plotting
np.random.seed(42)
sample_indices = np.random.choice(len(iq_data), size=min(100000, len(iq_data)), replace=False)
i_sample = iq_data.real[sample_indices]
q_sample = iq_data.imag[sample_indices]

fig.add_trace(
    go.Histogram(x=i_sample, nbinsx=100, name='I', marker_color='steelblue'),
    row=1, col=1
)
fig.add_trace(
    go.Histogram(x=q_sample, nbinsx=100, name='Q', marker_color='coral'),
    row=1, col=2
)

fig.update_xaxes(title_text="Amplitude", row=1, col=1)
fig.update_xaxes(title_text="Amplitude", row=1, col=2)
fig.update_yaxes(title_text="Count", row=1, col=1)

fig.update_layout(
    title=f"IQ Component Distributions CLEAN ON - {sample_file.name}",
    height=400,
    showlegend=False
)
save_figure(fig)
fig.show()

# IQ scatter plot (constellation diagram)
fig = go.Figure()
fig.add_trace(go.Scattergl(
    x=i_sample,
    y=q_sample,

```

```

        mode='markers',
        marker=dict(size=1, opacity=0.3, color='steelblue'),
        name='IQ Samples'
    ))

    fig.update_layout(
        title=f"IQ Constellation Diagram CLEAN ON - {sample_file.name}",
        xaxis_title="In-phase (I)",
        yaxis_title="Quadrature (Q)",
        height=600,
        width=600
    )
    save_figure(fig)
    fig.show()

    print(f"""
I component: mean={i_sample.mean():.4f}, std={i_sample.std():.4f}
Q component: mean={q_sample.mean():.4f}, std={q_sample.std():.4f}""")

```

=====

IQ VALUE DISTRIBUTION

File: AIR\_0000\_00.dat (CLEAN / ON)

=====

Saved: ../figures/001\_exploration\_general/IQ\_Component\_Distributions\_CLEAN\_ON\_AIR\_0000\_00dat.png

Saved: ../figures/001\_exploration\_general/IQ\_Constellation\_Diagram\_CLEAN\_ON\_AIR\_0000\_00dat.png

I component: mean=0.0094, std=0.0335

Q component: mean=0.0160, std=0.0347