

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: Downampling 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: 24000000 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_fold  
r']  
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 = []
rangeViolations = []

# 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))
    except Exception as e:
        corrupted_files.append((row['file_path'], str(e)))
        print(f" Error validating file {row['file_path']} due to {e}")

```

```

# 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 rangeViolations[:10]])
moreMsg = f" ... and {len(rangeViolations) - 10} more files" if len(rangeViolations) > 10 else ""
print(f"""\\nFiles with IQ values outside [-1, 1]: {len(rangeViolations)}\n\n{moreMsg}""")

filesViolating[-1, 1] range:
{violationsList}
{moreMsg}"""
else:
    print("-> All files have IQ values within expected range [-1, 1]\")\n\n

# Store validation results for conclusion
validationResults = {
    'total_files': total_files,
    'corrupted': len(corrupted_files),
    'sample_issues': len(sampleCountIssues),
    'duration_issues': len(durationIssues)
}
```

```
=====
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 ($\neq 2.0\text{s}$): 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 $< 2\text{s}$ 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}

==== REPLICA 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

```

==== REPLICA 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	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)

```

=====
```

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

```