

Power Log, Peak Detection, and Histogram Accumulator IP

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1. Problem Statement:- Design an integrated HLS IP core

1. Computes power or log-power per FFT bin
2. Detects peaks that cross a dynamic threshold
3. Accumulates histograms to analyze frequency occupancy over time

2. Workflow:-

1. **Design:** The IP core is implemented in **Vitis HLS C++** (`power.h` and `compute.cpp`). It streams complex FFT samples, computes the power in dBm using a fixed-point log approximation, and compares the result to configurable thresholds:
 - **Upper threshold(max_thresh)** detects unusually strong signals.
 - **Lower threshold(min_thresh)** detects signals that drop below a noise floor or expected minimum.

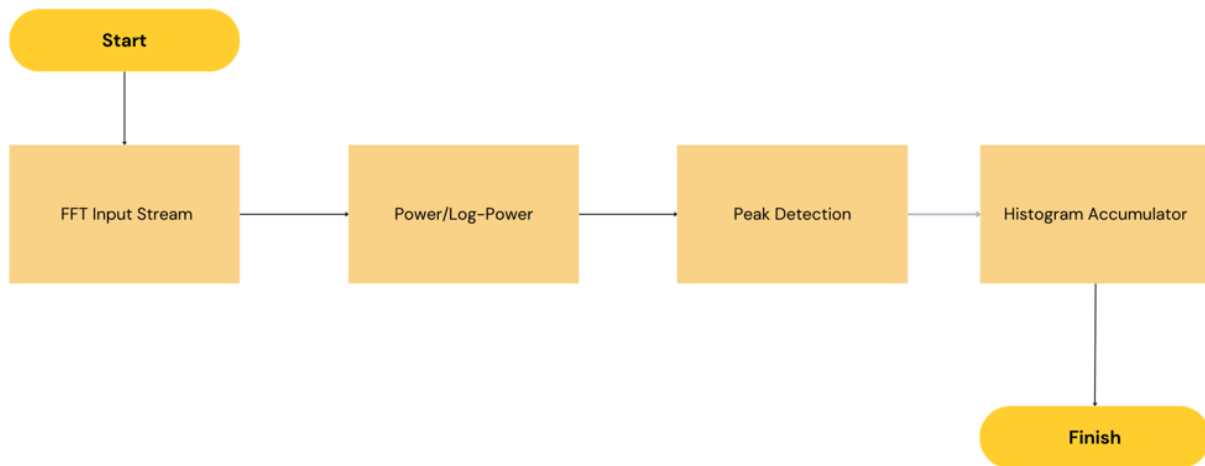
If the computed power crosses either threshold, the sample is flagged as a peak, and the corresponding frequency bin's histogram counter is incremented.

2. **Testbench:** The testbench (`compute_tb.cpp`) verifies the IP core by:
 - Reading FFT samples (`real`, `imag`, `bin_index`) from `3input.csv`.
 - Resetting the histogram to clear previous counts.
 - Streaming all samples through the IP to compute power, detect peaks, and update the histogram.
 - Printing debug information for the first few samples in each bin to check the peak detection logic.
 - Saving the final histogram to `histogram_output.csv` for visualization.

The thresholds `min_thresh` and `max_thresh` are configurable and control whether a sample is flagged as a peak. The output verifies that the frequency bins correctly accumulate the number of detected peaks over time.

3. **Visualization:** The histogram output is plotted in Google Colab using Python. The CSV is loaded, frequency bins are labeled in MHz, and a bar graph shows peak counts per bin. The plot verifies how the spectrum is occupied over time.

3. Flow Chart:-



4. Design Specification:-

<u>Parameter</u>	<u>Description</u>
Input Data Format	16-bit signed integer (real & imag)
Throughput	1 sample/cycle (II=1)
Platform	Vitis HLS Software
Pragma	Pipelining & Unrolling

5. Codes:-

1. Google colab for the input generation in a csv file [\[1\]](#)

```

import numpy as np
import pandas as pd
from scipy.fft import fft
from math import pi

# === Parameters ===
Fs = 100e6 # Sampling frequency
N = 65536 # Number of samples
num_bins = N // 2 # Positive frequency bins = 32768

# === Generate 65536 random samples using (r, theta) ===
r_min = 0.0219
r_max = 0.7096

```

```

# Random magnitudes and phases
r = np.random.uniform(r_min, r_max, N)
theta = np.random.uniform(0, 2 * pi, N)

# Convert to complex
real_time = r * np.cos(theta)
imag_time = r * np.sin(theta)
samples = real_time + 1j * imag_time

# === Perform FFT ===
fft_result = fft(samples)
fft_positive = fft_result[:num_bins] # Take only first 32768 positive bins

# === Scale and quantize ===
scale_factor = 2**10 # Scale float to int16 range for HLS
real_fixed = np.round(np.real(fft_positive) * scale_factor).astype(np.int16)
imag_fixed = np.round(np.imag(fft_positive) * scale_factor).astype(np.int16)
fft_bin_number = np.arange(num_bins, dtype=np.uint16) # 0 to 32767

# === Create CSV dataframe ===
df = pd.DataFrame({
    "real": real_fixed,
    "imag": imag_fixed,
    "fft_bin_number": fft_bin_number
})

# === Save to CSV ===
df.to_csv("fft_input_for_hls.csv", index=False)
print("✅ CSV saved: fft_input_for_hls.csv")

# === Optional: download in Google Colab ===
try:
    from google.colab import files
    files.download("fft_input_for_hls.csv")
except:
    pass

```

2. power.h

```

#ifndef POWER_H
#define POWER_H

```

```

#include <ap_int.h>
#include <ap_fixed.h>
#include <hls_stream.h>

#define N_SAMPLES 32768
#define N_BINS 16

#define BIN_FREQ_RES 1525 //  $\approx 100\text{MHz} / 65536$  (FFT bin
spacing)
#define MAX_FREQ 50000000 // Nyquist frequency: 50 MHz
#define FREQ_BIN_WIDTH (MAX_FREQ / N_BINS) // Each bin  $\approx 3.125$  MHz

// Type definitions
typedef ap_int<16> data_t;
typedef ap_uint<15> bin_index_t;
typedef ap_uint<4> bin_t;
typedef ap_uint<10> hist_t;
typedef ap_fixed<16, 8> dbm_t; // Q8.8 fixed-point type for dBm

// Input sample format
struct sample_t {
    data_t real;
    data_t imag;
    bin_index_t bin_index;
};

// Function to approximate  $10 \cdot \log_{10}(\text{power})$  using fixed-point math
dbm_t log10_approx(ap_uint<32> power);

// Top-level IP function for power computation and histogram update
void geeth_power(
    hls::stream<sample_t>& in_stream,
    hist_t histogram[N_BINS],
    bool reset_histogram,
    dbm_t min_thresh,
    dbm_t max_thresh
);

#endif

```

3. compute.cpp

```
#include "power.h"

// Accurate fixed-point 10*log10(x) approximation using log2(x)
// Q8.8 format → multiply by 256
dbm_t log10_approx(ap_uint<32> power) {
#pragma HLS inline

    // Avoid log(0)
    if (power == 0) power = 1;

    // Count leading zeros to estimate log2
    ap_uint<6> lz = __builtin_clz(power);
    ap_uint<6> log2_int = 31 - lz;

    // Normalize: shift power left so MSB is at bit 31
    ap_uint<32> norm = power << lz;

    // Extract fractional part from high bits
    ap_uint<8> frac = (norm >> 23) & 0xFF; // Top 8 bits after MSB

    // log2_fixed = int_part << 8 + frac (Q8.8)
    ap_uint<16> log2_fixed = (log2_int << 8) | frac;

    // log10(x) ≈ log2(x) * log10(2) → log10(2) ≈ 0.30103
    // So 10*log10(x) ≈ log2(x) * 10 * 0.30103 ≈ log2(x) * 77 in Q8.8
    dbm_t result = (log2_fixed * 77) >> 8;
    return result;
}

void geeth_power(
    hls::stream<sample_t>& in_stream,
    hist_t histogram[N_BINS],
    bool reset_histogram,
    dbm_t min_thresh,
    dbm_t max_thresh
) {
#pragma HLS INTERFACE axis port=in_stream
```

```

#pragma HLS INTERFACE s_axilite port=histogram bundle=CTRL
#pragma HLS INTERFACE s_axilite port=reset_histogram bundle=CTRL
#pragma HLS INTERFACE s_axilite port=min_thresh bundle=CTRL
#pragma HLS INTERFACE s_axilite port=max_thresh bundle=CTRL
#pragma HLS INTERFACE s_axilite port=return bundle=CTRL
#pragma HLS PIPELINE II=1

    static hist_t hist_local[N_BINS] = {0};
#pragma HLS ARRAY_PARTITION variable=hist_local complete dim=1

    if (reset_histogram) {
        for (int i = 0; i < N_BINS; i++) {
#pragma HLS UNROLL
            hist_local[i] = 0;
        }
    }

    // Read one sample per call
    sample_t s = in_stream.read();

    // Shift input to remove LSB noise
    data_t a = s.real >> 6;
    data_t b = s.imag >> 6;

    // Power = a^2 + b^2
    ap_uint<32> power = a * a + b * b;

    // Compute 10*log10(power) in Q8.8
    dbm_t dbm = log10_approx(power);

    // Determine if it's a peak
    bool peak = (dbm > max_thresh || dbm < min_thresh);

    // Get frequency bin from FFT index
    int freq_hz = s.bin_index * BIN_FREQ_RES;
    bin_t bin = freq_hz / FREQ_BIN_WIDTH;
    if (bin >= N_BINS) bin = N_BINS - 1;

    // Accumulate peak count
    if (peak && !reset_histogram) {

```



```

        hist_local[bin]++;
    }

    // Update output histogram
    for (int i = 0; i < N_BINS; i++) {
#pragma HLS UNROLL
        histogram[i] = hist_local[i];
    }
}

```

4. compute_tb.cpp

```

#include "power.h"
#include <fstream>
#include <iostream>
#include <sstream>
#include <string>

int main() {
    hls::stream<sample_t> in_stream;
    hist_t histogram[N_BINS] = {0};
    int debug_count[N_BINS] = {0};

    std::ifstream infile("3input.csv");
    if (!infile.is_open()) {
        std::cerr << "✗ ERROR: Could not open 3input.csv\n";
        return 1;
    }

    std::string line;
    getline(infile, line); // skip header

    const dbm_t min_thresh = -30;
    const dbm_t max_thresh = 20;

    // Step 1: Reset histogram
    geeth_power(in_stream, histogram, true, min_thresh, max_thresh);

    // Step 2: Stream all input samples

```

```

while (getline(infile, line)) {
    sample_t s;
    sscanf(line.c_str(), "%hd,%hd,%hu", &s.real, &s.imag,
&s.bin_index);
    in_stream.write(s);
    geeth_power(in_stream, histogram, false, min_thresh, max_thresh);
}
infile.close();

// Step 3: Print debug info (first 5 samples per bin)
std::ifstream infile_debug("3input.csv");
getline(infile_debug, line); // skip header again

std::cout << "\n Debug Info (First 5 Samples Per Bin):\n";

while (getline(infile_debug, line)) {
    sample_t s;
    sscanf(line.c_str(), "%hd,%hd,%hu", &s.real, &s.imag,
&s.bin_index);

    data_t a = s.real >> 6;
    data_t b = s.imag >> 6;
    ap_uint<32> power = a * a + b * b;

    dbm_t dbm = log10_approx(power);
    bool peak = (dbm > max_thresh || dbm < min_thresh);

    int freq_hz = s.bin_index * BIN_FREQ_RES;
    bin_t bin = freq_hz / FREQ_BIN_WIDTH;
    if (bin >= N_BINS) bin = N_BINS - 1;

    if (debug_count[bin] < 5) {
        std::cout << "  Bin " << (int)bin
            << " | real = " << s.real
            << ", imag = " << s.imag
            << ", power = " << (float)dbm
            << " dBm, peak = " << (peak ? "YES" : "NO") << "\n";
        debug_count[bin]++;
    }
}
}

```

```

infile_debug.close();

// Step 4: Save histogram to CSV
std::ofstream out_csv("histogram_output.csv");
if (!out_csv.is_open()) {
    std::cerr << "✗ ERROR: Could not write to histogram_output.csv\n";
    return 1;
}

out_csv << "Bin,Freq_Start_Hz,Freq_End_Hz,Peak_Count\n";
std::cout << "\n📊 Final Peak Histogram (Frequency Bins):\n";

for (int i = 0; i < N_BINS; i++) {
    int f_start = i * FREQ_BIN_WIDTH;
    int f_end = (i + 1) * FREQ_BIN_WIDTH - 1;

    std::cout << "  Bin " << i << " [" << f_start << " Hz - " << f_end
<< " Hz] → Peaks: "
        << histogram[i] << "\n";

    out_csv << i << "," << f_start << "," << f_end << "," <<
histogram[i] << "\n";
}

out_csv.close();
std::cout << "\n✅ histogram_output.csv has been saved
successfully!\n";

return 0;
}

```

5. Google colab for histogram plotting

```

import pandas as pd
import matplotlib.pyplot as plt

# Read histogram CSV

```

```

df = pd.read_csv("histogram_output.csv")

# Convert frequency to MHz with decimal precision
df["Freq_Range"] = df.apply(
    lambda row: f"{row['Freq_Start_Hz'] / 1e6:.3f}-{row['Freq_End_Hz'] / 1e6:.3f} MHz", axis=1
)

# Plot
plt.figure(figsize=(14, 6))
plt.bar(df["Freq_Range"], df["Peak_Count"], color='steelblue',
        edgecolor='black')
plt.xticks(rotation=45, ha='right')
plt.xlabel("Frequency Range (MHz)")
plt.ylabel("Peak Count")
plt.title("Histogram of Peaks Detected per Frequency Bin")
plt.grid(axis='y', linestyle='--', alpha=0.6)
plt.tight_layout()
plt.savefig("histogram_plot_mhz.png")
plt.show()

```

6. Outputs

Bin	Freq_Start_Hz	Freq_End_Hz	Peak_Count
0	0,3124999	285	
1	3125000	6249999	332
2	6250000	9374999	315
3	9375000	12499999	342
4	12500000	15624999	321
5	15625000	18749999	322
6	18750000	21874999	351
7	21875000	24999999	312
8	25000000	28124999	341
9	28125000	31249999	352
10	31250000	34374999	322
11	34375000	37499999	333
12	37500000	40624999	298
13	40625000	43749999	376
14	43750000	46874999	319
15	46875000	49999999	314

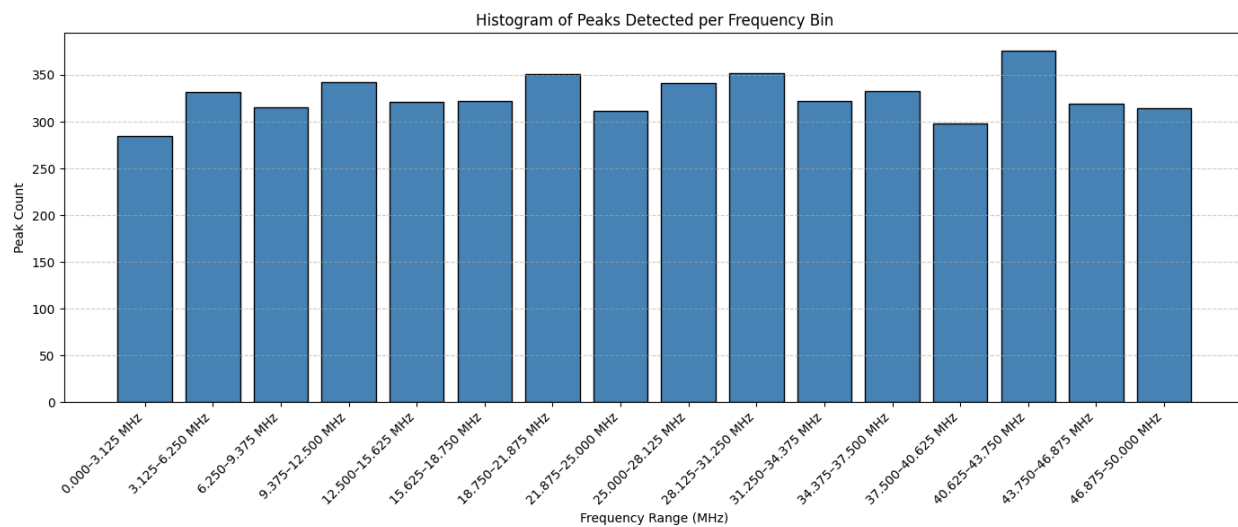


Figure 1: Frequency Range vs Histogram

Performance & Resource Estimates

☒ Modules
 ☒ Loops
 ☒ Hide empty columns

MODULES & LOOPS	LATENCY(CYCLES)	LATENCY(NS)	INTERVAL	PIPELINED	BRAM	DSP	FF
geeth_power	20	200.000	16	yes	0	4	946

Figure 2 : Performance of the Code

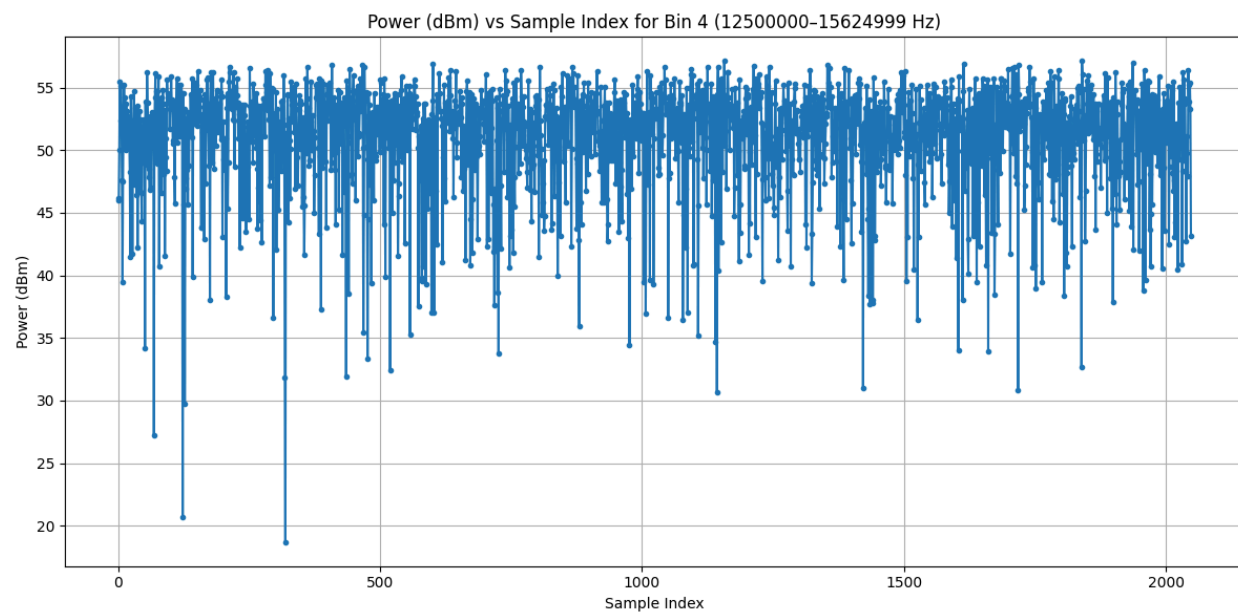


Figure 3 : Sample index vs power (dbm)

6. Challenges Faced:-

- Debugging stream interfaces and ensuring $II=1$ pipelining.
- Designing an accurate fixed-point \log_{10} approximation[2].

7. Conclusion:-

In this project, an integrated Power Log, Peak Detection, and Histogram Accumulator IP was successfully designed, simulated, and verified. FFT samples were processed in real-time to compute power or log-power, detect peaks above or below dynamic thresholds, and accumulate frequency bin histograms.

8. Weekly Report:-

Week 1&2: Learned basics of RTL and HDL Bits

Week 3: Given Problem Statement[3]

Week 4: Did the functionality of the problem statement in vitis[4]

Week 5: Input Generation in Google Colab[5]

Week 6: Integrating the Google Colab and Vitis to get the histogram output

9. References:-

1. <https://docs.google.com/spreadsheets/d/1tYSwBJcjGEta9Bw-4-LUpVkdir5pXfV5g0q6ZWoetrOA/edit?usp=sharing>
2. <https://coretechgroup.com/dbm-calculator/>
3. https://www.xilinx.com/support/documents/sw_manuals/xilinx2022_2/ug1399-vitis-hls.pdf
4. <https://www.youtube.com/@nitinchandrachoodan6783/playlists>
5. <https://www.youtube.com/watch?v=GlbJxX130iE>