# Optimization of a Gravitational Wave Detection Pipeline

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### Structure

#### N-detector

Gravitational Waves

The power of powers of 2

The hunt for hardcoded detectors

#### Complexity analysis

Background

So why do we care about this complexity analysis?

**CUDA** 

Analysis

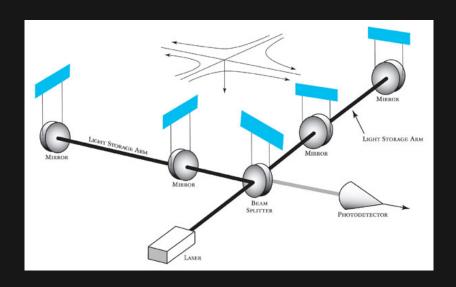
#### Future Work

# N-detector

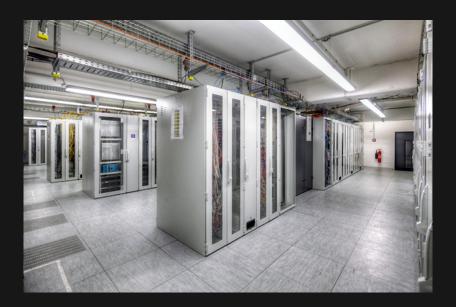
### **Gravitational Waves**



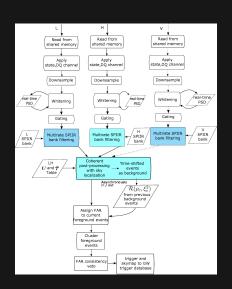
### Detectors



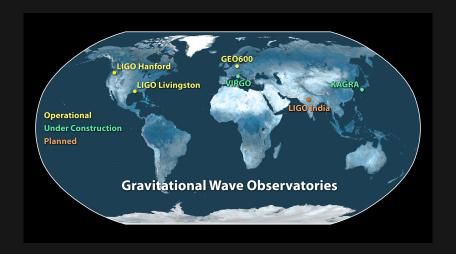
# Processing Pipelines



### The SPIIR Pipeline



### The SPIIR Pipeline



#### Internal datastructures

```
#define MAX_IFO_COMBOS 7 // 2^3-1
static const IFOType IFOComboMap[MAX_IFO_COMBOS] = {
    {"H1", 0},
    {"L1", 1},
    {"V1", 2}.
    {"H1L1", 3},
    {"H1V1", 4},
    {"L1V1", 5},
    {"H1L1V1", 6},
```

#### Internal datastructures

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#define MAX_IFO_COMBOS 7 // 2^3-1
static const IFOType IFOComboMap(MAX_IFO_COMBOS) = {
    {"H1", 0},
    {"L1", 1},
    {"V1", 2},
    {"K1", 3},
    {"H1L1", 4},
    {"H1V1", 5},
    {"H1K1", 6},
    {"L1V1", 7},
    {"L1H1", 8},
    {"V1H1", 9},
    {"H1L1V1", 10},
    {"H1L1K1", 11},
    {"H1V1K1", 12},
    {"L1V1K1", 13},
    {"H1L1V1K1", 14},
```

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#define MAX_IFO_COMBOS 7 // 2^3-1
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```

00000

00001

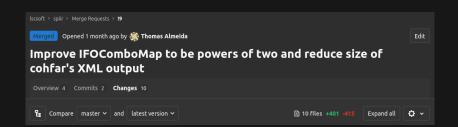
0 0 1 0 1

10101

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```

```
N_{detectors} = popcount(icombo)
Detector being used: icombo & 2^{ifo}
```



### The hunt for hardcoded detectors

```
typedef struct tagPostcohInspiralTable {
    struct tagPostcohInspiralTable *next;
    long process_id;
    long event id:
    LIGOTimeGPS end_time;
    LIGOTimeGPS end_time_H;
    LIGOTimeGPS end time L;
    LIGOTimeGPS end_time_V;
```

### The hunt for hardcoded detectors

```
typedef struct tagPostcohInspiralTable {
    struct tagPostcohInspiralTable *next;
    long process_id;
    long event_id;
    LIGOTimeGPS end_time;
    LIGOTimeGPS end_time_sngl[MAX_NIFO];
```

#### The hunt for hardcoded detectors



# Complexity analysis

### Complexity Analysis

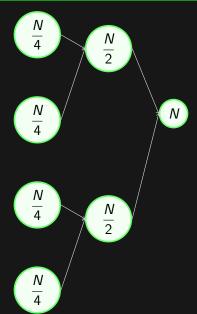
#### Asymptotic growth rates

- · compare algorithms
- compare with inherent performance barriers
- provide simple closed form approximations
- big O upper bounds on growth
- $\bullet$  big  $\Omega$  lower bounds on growth

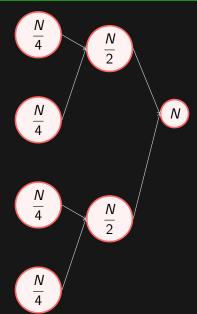
#### Analysis of recursive programs

- express as recurrence relation
- · look for pattern to find closed form
- can then do asymptotic analysis

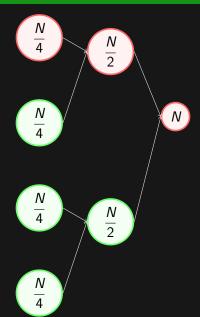
# Complexity Analysis



# Work



# Span

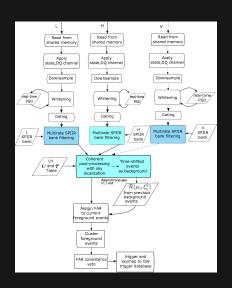


### Parallel definitions

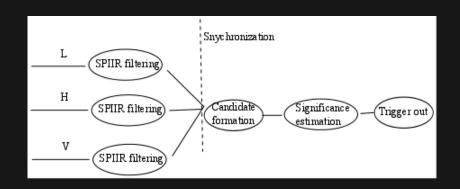
- ightharpoonup Work =  $T_1$
- ▶ Span =  $T_{\infty}$
- Brent's Theorem  $O(\frac{T_1}{P} + T_N)$

So why do we care about this complexity analysis?

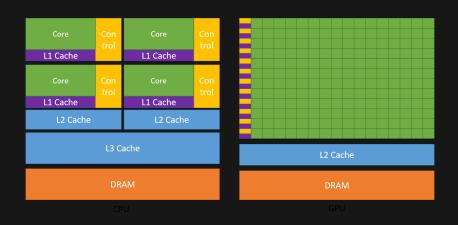
# Why do we care?



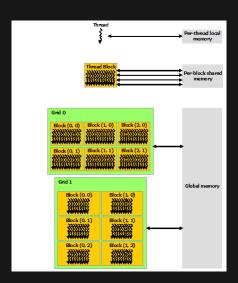
# Why do we care?



### **CUDA**



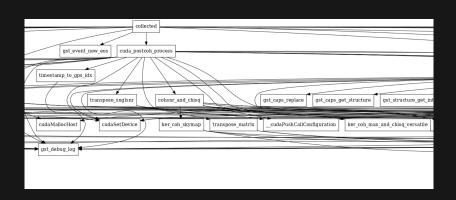
### Compute model



### Callgraph generation

```
tommoa:~/Documents/research# [master] python3 -m utils --help
usage: utils [-h] {combine,co,iir,dot,pipeline} ...
Utilities to help with Tom Almeida's GENG5551 research.
optional arguments:
  -h, --help
                        show this help message and exit
subcommands:
  {combine,co,iir,dot,pipeline}
                        subcommand help
tommoa:~/Documents/research# [master]
```

### Callgraph



### **Analysis**

$$T_{\infty} = O(NT + N^2 + D^3 + D^2 \log N + D \log S + D \log P),$$
  
 $T_1 = O(NT + N^2 + SPD^3 + SPBD^3 + ND^2),$ 

where D is the number of detectors, S is the number of sky directions, T is the number of templates, N is the number of samples, N is the number of time shifts made to background noise; and  $P = \max\{S, T\}$ .

### **Analysis**

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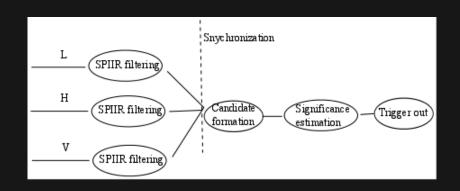
N is the number of samples,

B is the number of time shifts made to background noise; and  $P = \max\{S, T\}$ .

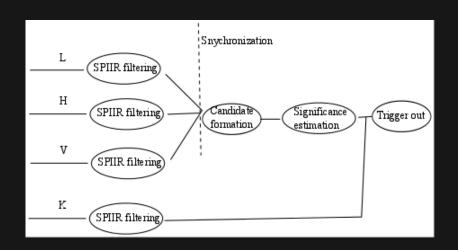
$$\frac{4^3}{3^3} = \frac{64}{27} \approx 2.37$$

# Future Work

### Pipeline Structure



### Pipeline Structure



### Algorithms

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### Algorithms

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$$\frac{4^2}{3^2} = \frac{16}{9} \approx 1.78$$

# Questions?