

### Part I

$$\bar{C}_{\max}(1 - \alpha, \mu)$$

Computing  $\bar{C}_{\max}(C, \mu)$  for optimum interval calculation, where  $\mu$  is the number of expected events and  $1 - \alpha$  is how frequently you reject the null hypothesis when it is true. The single-event energy spectrum, that is, the probability density function which tells us which energy depositions are likely to occur, is independent of the chosen WIMP model – we always expect a simple exponential recoil spectrum.

The number of dark matter events detected does depend on the WIMP mass and cross-section. We know, however, that it must follow a Poisson distribution, which leaves the Poisson mean (which equals the expected number of events) as the only parameter left to estimate. From an upper limit on this mean, an upper limit curve in the dark matter mass – cross-section plane can be computed.

- A `list_of_energies` list of reconstructed energy depositions of single events (from here on simply ‘energies’), either measured during some run of an actual detector, or generated using Monte Carlo.)
- An interval is an interval in energy space.
- The size of an interval is the fraction of energies expected in that interval. Clearly, this depends on which energy spectrum we assume, but is independent of the Poisson mean we are trying to constrain. By definition this is a number between 0 and 1.
- The K-largest interval of a run is the largest interval containing K events in that run. Recall our definition of size: a ‘large’ interval is one which is unusually empty in that run. Clearly k-largest intervals will terminate at (or technically, just before) an observed energy, or at one of the boundaries of our energy space. Again, which interval in a run is the k-largest, depends on our energy spectrum, but not on our Poisson mean.
- The extremeness of a K-largest interval is the probability of finding the K-largest interval in a run to be smaller. This clearly does depend on the Poisson mean: if we expect very few events, large gap sizes are more likely. Clearly extremeness is a number between 0 and 1; values close to 1 indicate unusually large intervals, that is, usually large (almost-)empty regions in the measured energies. For example, if the extremeness of a k-largest interval in a run is 0.8, that means that 80% of runs have k-largest intervals which are smaller than the k-largest interval in this run.
- The optimum interval statistic of a run is extremity of the most extreme k-largest interval in a run.
- The extremeness of the optimum interval statistic is the probability of finding a lower optimum interval statistic, that is, of finding the optimum interval in a run to be less extreme.

The max gap method rejects a theory (places a mean outside the upper limit) based on a run if the 0-largest interval (the largest gap) is too extreme.

The optimum interval method rejects a theory based on a run if the optimum interval statistic is too large.

- The energy cumulant  $\epsilon(E)$  is the fraction of energies expected below the energy  $E$ . Whatever the (1-normalized)

energy distribution  $dN/dE$ ,  $dN/d\epsilon$  is uniform[0,1], where 0 and 1 correspond to the boundaries of our experimental range.

```
In [16]: import functools
from scipy.optimize import brentn
import matplotlib.pyplot as plt
import numpy as np
import pickle
import os
```

```
In [17]: def kLargestIntervals(list_of_energies, spectrumCDF = lambda x: x):
    """
    Returns a list of the sizes of the K-largest intervals in that run according to th
    That is, kLargestIntervals(...)[i] is the size of the largest interval containing

    * Transform energies to energy cumulants
    * Add events at 0 and 1
    * Foreach k, compute interval sizes, take max
    """
    answer = {}

    list_of_energies.sort()

    energy_cumulants = spectrumCDF(list_of_energies)

    for interval_size in range(len(energy_cumulants)):
        if (1 + interval_size) >= len(energy_cumulants):
            continue

        temp_data = energy_cumulants.copy()
        gap_sizes = temp_data[(1+interval_size):] - temp_data[0:-1*(1 + interval_size)]

        answer[interval_size] = np.max(gap_sizes)

    return answer

assert kLargestIntervals(np.array([0.0, 0.1, 0.2, 0.84, 0.85]))[0] == (0.84 - 0.2) #
assert kLargestIntervals(np.array([0.0, 0.1, 0.2, 0.84, 0.85]))[2] == (0.84 - 0.0) #
assert kLargestIntervals(np.array([0.85, 0.0, 0.1, 0.84, 0.2]))[2] == (0.84 - 0.0) #
```

```
In [18]: def extremenessOfInterval(x, k, mu):
    """
    Returns the extremeness of a k-largest interval of size, if the poisson mean is mu

    (Number of itvSizes[mu][k] smaller than size) / mcTrials[mu]

    x - also size in above comment
    k - gap (rename k)
    """
    # [0] is because where returns list, where [0] is answer
    if k not in itvSizes[mu]:
        return 0

    return np.where(itvSizes[mu][k] < x)[0].size / mcTrials[mu]
```

```
In [19]: def optimumItvStatistic(list_of_energies, mu, spectrumCDF = lambda x: x):
    """
    Returns the optimum interval statistic of the run.

    Max of extremenessOfInterval's
    """
    return np.max([extremenessOfInterval(x, k, mu) for k, x in kLargestIntervals(list_of_energies, spectrumCDF)])
```

```

In [20]: def extremenessOfOptItvStat(stat, mu):
        """
        Returns the extremeness of the optimum interval statistic stat, given mu

        (Number of optItvs[mu] smaller than stat) / mcTrials[mu]
        """
        return np.where(optItvs[mu] < stat)[0].size / mcTrials[mu]

In [26]: def optItvUpperLimit(list_of_energies, c, spectrumCDF = lambda x: x,
                             n = 1000):
        """
        Returns the c- confidence upper limit on mu using optimum interval

        For which mu is extremenessOfOptItvStat( optimumItvStatistic(run), mu ) = c

        c - e.g., 0.9
        """
        def f(mu, list_of_energies, c, spectrumCDF, n):
            generate_table(mu, n)
            x = optimumItvStatistic(list_of_energies, mu, spectrumCDF)
            prob = extremenessOfOptItvStat(x, mu)
            return prob - c

        mu = 0

        for mu in np.arange(10, 2 * list_of_energies.size):
            if f(mu, list_of_energies, c, spectrumCDF, n) > 0:
                print('Found seed mu=%f' % mu)
                break

        try:
            xsec = brentn(f, mu - 5, mu + 5,
                          args=(list_of_energies, c, spectrumCDF, n),
                          xtol=1e-2)
            print('Improved xsec:', xsec)
        except:
            print("ERROR: could not minimize", mu)
            return mu
        return xsec

In [22]: def generate_trial_experiment(mu, n):
        trials = []

        for index in range(n):
            this_mu = np.random.poisson(mu)

            rand_numbers = np.random.random(size=this_mu)
            rand_numbers = np.append(rand_numbers, [0.0, 1.0])
            rand_numbers.sort()
            trials.append(rand_numbers)

        return trials

```

## 1 Monte Carlo for populating $\text{itvSizes}[\mu][k]$ and $\text{optItvs}[\mu]$

```

In [36]: def get_filename():
        return 'saved_intervals.p'

def load_table_from_disk():
    global itvSizes
    global optItvs
    global mcTrials

    if os.path.exists(get_filename()):
        f = open(get_filename(), 'rb')
        itvSizes = pickle.load(f)
        optItvs = pickle.load(f)
        mcTrials = pickle.load(f)
        f.close()

def write_table_to_disk():
    f = open(get_filename(), 'wb')
    pickle.dump(itvSizes, f)
    pickle.dump(optItvs, f)
    pickle.dump(mcTrials, f)
    f.close()

itvSizes = {}
optItvs = {}
mcTrials = {}
load_table_from_disk()

def generate_table(mu, n):
    """ #Generate trial runs"""
    if mu in mcTrials and mcTrials[mu] >= n:
        return

    print("Generating", mu)

    mcTrials[mu] = n
    trials = generate_trial_experiment(mu, mcTrials[mu])

    itvSizes[mu] = {}
    optItvs[mu] = []

    for trial in trials:
        intermediate_result = kLargestIntervals(trial)

        for k, v in intermediate_result.items():
            if k not in itvSizes[mu]:
                itvSizes[mu][k] = []

            itvSizes[mu][k].append(v)

    # Numpy-ize it
    for k, array in itvSizes[mu].items():
        itvSizes[mu][k] = np.array(array)

    for trial in trials:
        optItvs[mu].append(optimumItvStatistic(trial, mu))

    # Numpy-ize it
    optItvs[mu] = np.array(optItvs[mu])

def cache_values(my_max=200, n=100):
    for i in range(3, my_max):
        generate_table(i, n)
        write_table_to_disk()

```

```
In [52]: def plot_something():
x, y = [], []

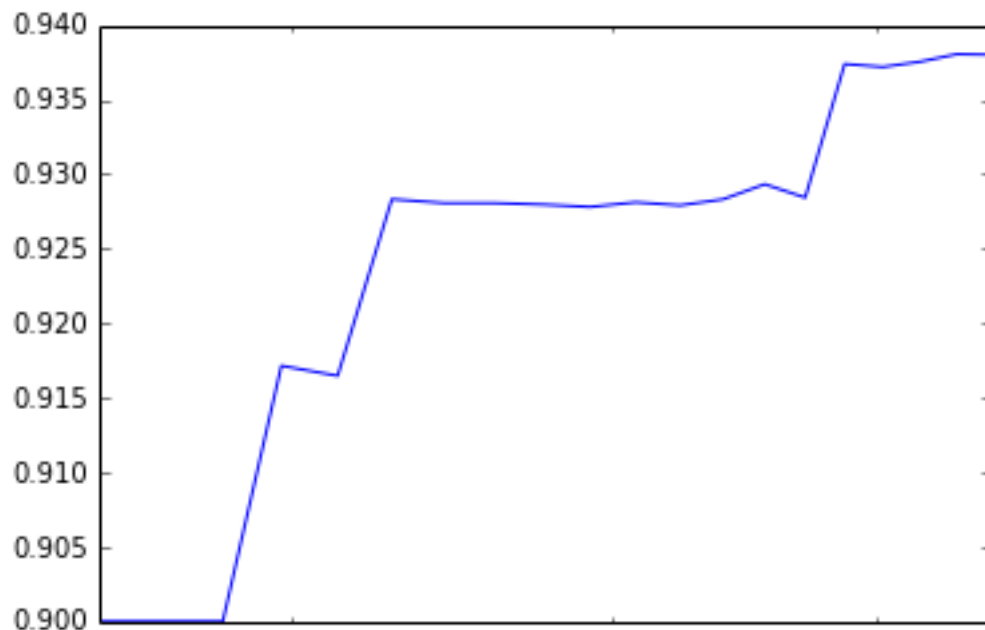
for mu in np.linspace(3.5, 6.5, 20):
    generate_table(mu, 10000)
    x.append(mu)

    a = brentn(lambda x: extremenessOfOptItvStat(x, mu) - 0.9,
               0,
               1,
               xtol=1e-2)

    y.append(a)

plt.plot(x,y)
plt.xscale('log')
plt.xlim(3.5, 6.5)

#plot_something()
```



```
In [71]: def simple_test_uniform():
test_list_of_energies = generate_trial_experiment(mu=100, n=1)[0]
print(len(test_list_of_energies))
answer = optItvUpperLimit(test_list_of_energies, 0.9)
```

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107
testing mu=3.000000
... 0.000000, so continuing...
testing mu=4.000000
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testing mu=5.000000
... 0.000000, so continuing...
testing mu=6.000000
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