Search for Short Transient Neutrino Emission with ${\bf Deep Core\text{-}Ice Cube}$

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ABSTRACT

We present the results of a search for sources of brief transient neutrino emission using IceCube and DeepCore data acquired between May 15th 2012 and April 30th 2013. While the search methods employed in this analysis are similar to those used in previous IceCube point source searches, the data set being examined consists of a sample of predominantly sub-TeV muon neutrinos obtained through a novel event selection method. Thus, this search represents a first attempt at identifying astrophysical neutrino sources in this relatively unexplored energy range. The reconstructed direction and time of arrival of neutrino events is used to search for any significant self-correlation in the dataset; there is no comparison of the data set to a list of possible sources. This search encompasses the Northern sky ranging in declination from -5° to 90°. Examination of the data revealed no significant source of transient neutrino emission. This result has been used to construct limits on generic soft-spectra transients as well as a specific model of neutrino emission from soft jets in core-collapse supernovae.

Subject headings: neutrino astronomy, neutrinos, GRB, supernova, astroparticle physics

1. Introduction

The nascent field of neutrino astronomy exhibits great potential in its ability to answer several open questions in astrophysics.

The detection of astrophysical neutrino sources is a chief goal of the IceCube Neutrino Observatory (IceCube Collaboration et al. 2006). Located at the geographic South Pole, IceCube utilizes the clear glacial ice of the Antarctic ice cap as a detection medium for the secondary products of neutrino interactions. The detector consists of 5,160 Digital Optical Modules (DOMs) distributed among 86 cables to form a km³ instrumented volume. These DOMs house photomultiplier tubes for the capture of Cherenkov photons as well as digitizing electronics for initial processing of the PMT data. A centrally located region of denser instrumentation featuring DOMs with more sensitive PMTs comprises the sub-array DeepCore (Abbasi et al. 2012). This extension to IceCube array augments the detector's response to lower energy neutrino events.

IceCube analyses attempting to resolve astrophysical neutrino sources typically make use of a high-energy muon neutrino sample ($E_{\nu} \gtrsim 1$ TeV) of high-purity to look for both steady (Aartsen et al. 2014) and transient sources (Aartsen et al. 2015). As of yet, these searches have not found any significant self-correlations within the data sample nor correlations between the neutrino data and known astrophysical objects of interest. These analyses have largely eschewed low energy neutrino events collected by DeepCore due to poor resolution of these events as well as an increasingly strong irreducible background at lower energies given by the soft spectrum of atmospheric neutrinos. However, application of these predefined search techniques to a sample of low energy (30 GeV $\leq E_{\nu} < 1$ TeV) muon neutrino events from DeepCore can enhance IceCube's sensitivity to short transient neutrino sources with softer spectra.

Because of the strong atmospheric neutrino background in this energy range,

searches using a data set composed of these low energy events will only be sensitive to time-dependent emission. Some potential sources include flaring from active galactic nuclei (AGN) due to brief periods of enhanced accretion (Becker & Biermann 2009), 100-GeV scale sub-photospheric neutrino emission from gamma-ray bursts (Murase et al. 2013), and neutrino emission from mildly relativistic jets in core-collapse supernovae. If the emission spectra for these sources are sufficiently soft or feature an energy cutoff below the optimum energy for IceCube, they may not be visible to the traditional IceCube point source searches.

A promising potential source for this study is a special class of core-collapse supernova referred to as a choked GRB. There is an observed correlation between long duration gamma-ray bursts (GRBs) and core-collapse supernovae (SNe) ((Woosley & Bloom 2006), (Modjaz 2011)). The standard GRB model assumes that relativistic jets are generated during accretion of material onto the compact object formed during core-collapse (Rees & Meszaros 1992). Fermi-acceleration of charged particles occurs within internal shocks of these jets leading to gamma ray emission once the jets breach the surrounding stellar envelope. The observed fraction of SNe resulting in the occurrence of a GRB is quite low, however, it may be that a larger fraction of core-collapse SNe still manage to produce mildly relativistic jets. Due to insufficient energy, these jets fail to break through the stellar envelope and any gamma ray emission is effectively 'choked' off. If protons are accelerated in these jets, then neutrino production will occur in the shocks of the jet irrespective of whether or not the jet successfully escapes. A model of this neutrino emission proposed by Razzaque et al. (2004) and extended upon by Ando & Beacom (2005) suggests that these neutrinos may be detectable by IceCube-DeepCore for nearby supernovae (Taboada 2010).

We present the results of a search for transient neutrino emission on a set of low-energy neutrino event data collected from May 15th, 2012 to April 30th, 2013. The data selection methods used to acquire this unique event sample will be detailed in Sec. 2. Analysis

methods and search techniques are discussed in Sec. 3. Lastly, the results of the search are given in Sec. 4 in addition to how these results may be interpreted within the context of generic neutrino flares as well as choked GRBs.

2. Event Selection

The IceCube detector is primarily designed for the detection of high-energy ($E_{\nu} \geq 1$ TeV) muon neutrinos originating from the Northern sky. However, the addition of the DeepCore sub-array in combination with veto techniques allow for significant lowering of the detection threshold energy. The veto method, which is described in detail in Abbasi et al. (2012), features prominently in the isolation of the event sample used in this search. Several cuts related to event topology and directional reconstruction are also used to further separate background events from potential signal. Finally, a multi-variate cut developed through machine learning techniques is used to generate the final event sample used in the search.

2.1. Veto and Topology Cuts

The selection process begins with the output of the DeepCore filter, a data filter designed to eliminate down-going cosmic ray muons that produce light in the detector outside a defined volume encompassing DeepCore and the central IceCube detector.

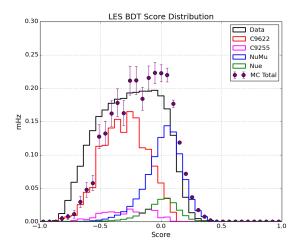


Fig. 1.— [add caption]

2.2. Boosted Decision Tree

3. Analysis Method

The search methods employed in the analysis of this data are nearly identical to those used in previous time-dependent IceCube analyses (see Braun et al. (2008) and Aartsen et al. (2015)). The arrival times and directions of events within the dataset are fed to a likelihood function which is then used to perform a likelihood ratio test to compare a signal plus background hypothesis for the data to the background only hypothesis.

Construction of this likelihood function begins with the assignment of individual event probabilities that reflect the likelihood of seeing an event i with arrival time t_i , reconstructed direction \vec{x}_i , and angular uncertainty σ_i given a hypothetical source located at \vec{x}_s with strength n_s having a Gaussian time profile with mean time t_0 and width σ_w .

$$\mathcal{P}_i(\vec{x}_i, t_i, \sigma_i | \vec{x}_s, n_s, t_0, \sigma_w) = \frac{n_s}{n_{\text{tot}}} \mathcal{S}_i + \left(1 - \frac{n_s}{n_{\text{tot}}}\right) \mathcal{B}_i$$
 (1)

The S_i and B_i terms listed in Eq. 1 are the signal and background probability density

functions (p.d.f.) respectively. The p.d.f.s used in this search differ slightly from those in previously reported searches in that they use no reconstructed energy information. The signal p.d.f. is given by

$$S_i(|\mathbf{x}_i - \mathbf{x}_s|, t_i, t_o, \sigma_w, \sigma_i) = S_i(|\mathbf{x}_i - \mathbf{x}_s|, \sigma_i) \cdot T_i(t_i, t_o, \sigma_w)$$
(2)

where

$$S_i(|\mathbf{x}_i - \mathbf{x}_s|, \sigma_i) = \frac{\kappa}{4\pi \sinh \kappa} \exp(\kappa \cos|\mathbf{x}_i - \mathbf{x}_s|)$$
(3)

and

$$T_i(t_i, t_o, \sigma_w) = \frac{1}{\sqrt{2\pi}\sigma_w} \exp\left(-\frac{(t_i - t_o)^2}{2\sigma_w^2}\right)$$
(4)

The spatial component of the signal p.d.f., S_i , features another difference with respect to previous searches (see Aartsen et al. (2014)). Due to the larger estimated error in reconstructed direction for low energy events, the Kent-Fisher distribution is used to model the probability of event-source association, and it is analogous to the Gaussian distribution normalized to the 2-sphere (Kent 1982).

The background p.d.f., \mathcal{B}_i , is derived from the final level data set which is dominated by background. It has the following form

$$\mathcal{B}_i(\mathbf{x}_i, t_i) = P_{BkgDec}(\delta_i) \frac{P_{BkgAz}(\alpha_i)}{T}$$
(5)

where T is the total livetime of the search, $P_{BkgDec}(\delta_i)$ is a p.d.f. describing the event declination distribution, and $P_{BkgAz}(\alpha_i)$ is a p.d.f. describing the event distribution in detector azimuth. These p.d.f.s are generated directly from data and do not depend on any background simulation.

The likelihood function itself is simply the product sum of all individual event probabilities:

$$\mathcal{L}(\mathbf{x}_s, n_s, t_0, \sigma_w) = \prod \mathcal{P}_i(|\mathbf{x}_i - \mathbf{x}_s|, n_s, t_i, t_0, \sigma_w, \sigma_i)$$
(6)

The ratio between the likelihood function values under the background only hypothesis $(n_s = 0)$ and the signal plus background hypothesis is maximized through variation of the source parameters n_s , σ_w , and t_0 . The test statistic $\hat{\lambda}$ is then defined as the maximum value of the likelihood ratio:

$$\hat{\lambda} = -2\log\left[\frac{\sqrt{2\pi}\hat{\sigma}_w}{T}\frac{\mathcal{L}(n_s = 0)}{\mathcal{L}(\vec{x}_s, \hat{n}_s, \hat{t}_o, \hat{\sigma}_w)}\right]$$
(7)

with $\mathcal{L}(n_s=0)$ corresponding to the likelihood of the null hypothesis and $\mathcal{L}(\vec{x}_s, n_s, \hat{t}_o, \hat{\sigma}_w)$ the likelihood of the signal plus background hypothesis with the best-fit values of the source parameters. Because this is a search for sources of finite duration over a timescale of limited duration, the number of potential short duration flares within the data set exceeds that of flares of longer duration leading to an effective trials factor. This results in a bias towards flares of shorter duration. We counteract this effect through the introduction of a marginalization term $T/\sqrt{2\pi}\hat{\sigma_w}$ in test statistic formulation which serves to penalize flares of shorter duration. More details about this term and its justification can be found in Braun et al. (2010).

The test statistic is defined such that it will asymptotically follow a χ^2 distribution with degrees of freedom corresponding to the number of fitted parameters for data consisting solely of background events. This enables the value of the maximized test statistic $\hat{\lambda}$ to be used to estimate the pre-trials p-value of the best-fit flare.

This procedure involves scrambling the event arrival times in the final dataset which also serves to scramble event right ascension.

In order to preserve generality, the presented search makes no use of information outside of the data set to designate source regions or time periods of interest. Instead, each point in the sky over a declination band ranging from -5° to 90° is examined. This is accomplished by discretizing the sky into separate bins and letting the location of these bins serve as the location of a hypothetical flaring source. Maximization of the likelihood is

then performed at each bin to obtain a test statistic $\hat{\lambda}$ for each bin. The first iteration of this scan uses a relatively coarse 2° by 2° binning. Following completion of this first scan, a followup scan with finer 0.5° by 0.5° binning is performed over coarse bins featuring a pre-trials p-value more significant than a predefined threshold $(-\log_{10}(\text{p-value}) > 1.75)$. The result is a map of pre-trials p-values which shows the estimated significance of the best-fit flare hypothesis at each bin. The best-fit flare from the bin featuring the most significant maximized test statistic after both scans is returned as the hottest spot in the search.

4. Results and Interpretations

Applying the described analysis method on the unscrambled dataset yields a skymap of the pre-trials p-values derived from the maximized test statistic for each bin. This map is shown in Figure 2. The black circle in Figure 2 shows the hottest spot after the completion of the sky scan. The best fit to flare parameters for this hot spot are listed in Table 1.

R.A.	Dec	\hat{n}_s	$\hat{t}_0[MJD]$	$\hat{\sigma}_w[days]$	p-value	p-value post-trial
268.75°	54.25°	13.528	56107.8	5.89	4.1751	56%

Table 1: Best-fit values for flare location, duration, strength and time.

We obtain an event upper limit for a generic neutrino flare through the same method detailed in).

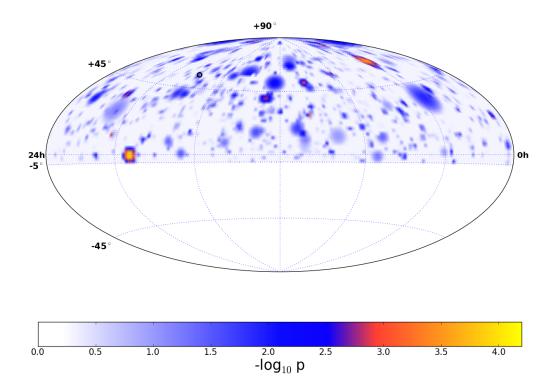


Fig. 2.— Sky map of pre-trials p-values for best fit flares per bin. The black circle identifies the location of the most significant flare found at $RA = 268.75^{\circ}$ and Declination = 54.25° .

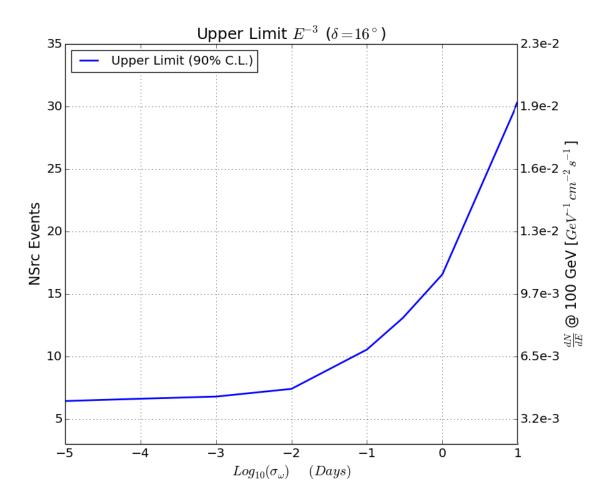


Fig. 3.— [add caption]

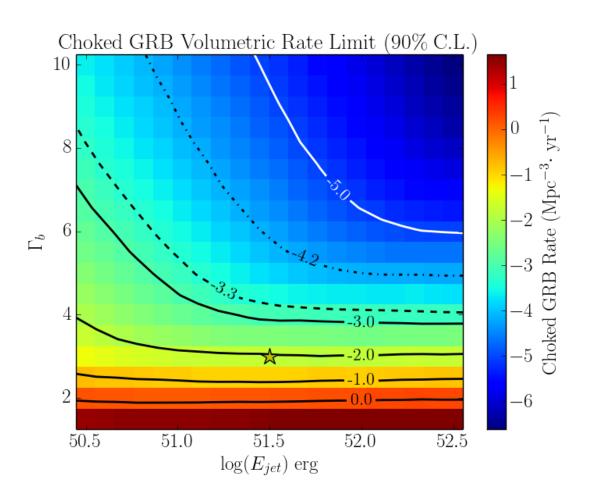


Fig. 4.— [add caption]

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