



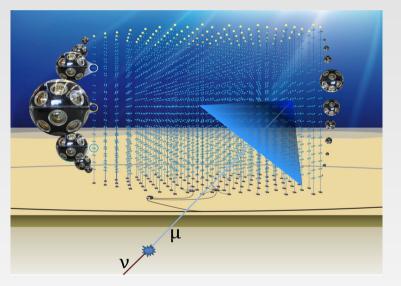




European Union European Social Fund

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Muon and Neutrino Energy Reconstruction for KM3NeT



Drakopoulou Evangelia N.C.S.R. Demokritos

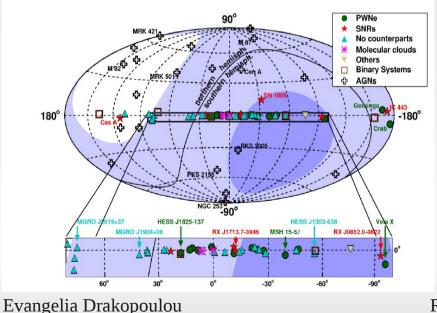


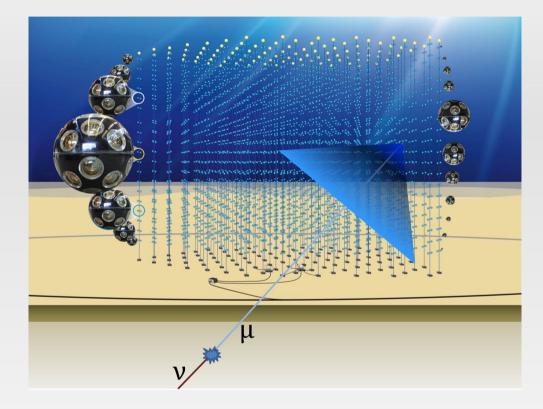




European Union MANAGING AUTHORITY European Social Fund Co-financed by Greece and the European Union

- KM3NeT Neutrino Telescope with volume of several km³ which will be placed in the Mediterranean Sea.
- The telescope will search for neutrinos from <u>galactic</u> and <u>extragalactic</u> astrophysical sources (like Gamma Ray Bursts, Supernovae, Colliding Stars).





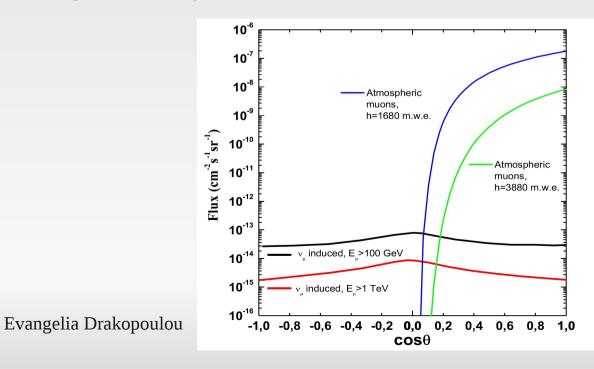
- Sky coverage in galactic coordinates for a detector located in the Mediterranean Sea.
- Dark (light) areas are visible at least 75% (25%) of the time.

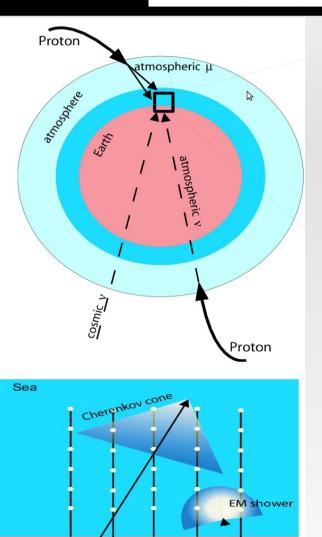


KM3NeT - Backgrounds

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- **Atmospheric Muons :** contained in the extensive air showers produced by cosmic rays in the atmosphere.
- Atmospheric Neutrinos : produced by charged kaons or pions in cosmic rays interactions in the atmosphere.
- K^{40} : radioactive potassium isotope
- **Bioluminescence :** life forms that inhabit in the deep sea emit light.





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Earth

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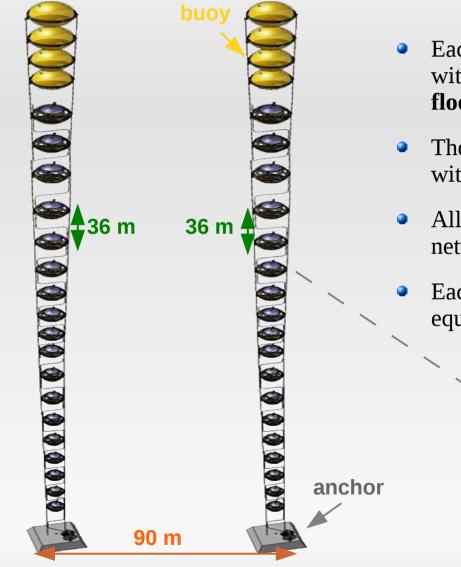
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Detector Configuration





- Each of the 6 blocks has an almost hexagonal geometry with 115 strings at 90m distance. Each string has 18 floors and each floor has 1 Optical Module (OM).
- The optical modules are arranged in **vertical strings** with a height of almost **600m**.
- All data are transmitted to shore via an optical fibre network.
- Each optical module consists of a **17**" glass sphere, equipped with **31 3 inches photomultipliers**.

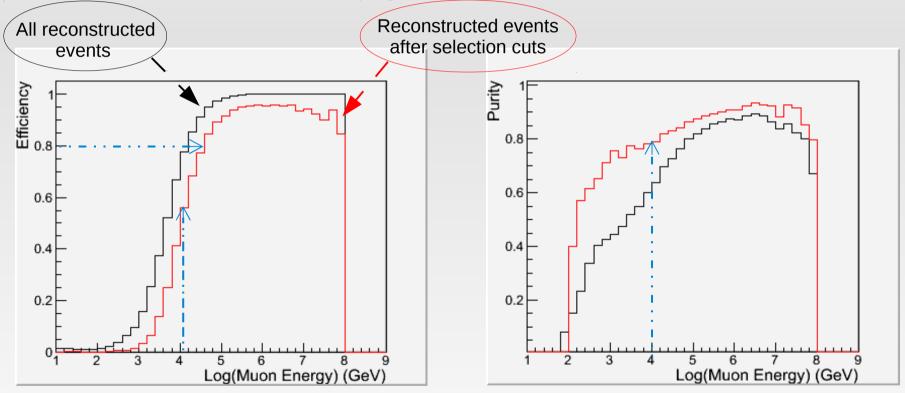




Track Reconstruction



 <u>Muons produced by neutrinos via Charged Current interactions</u> were reconstructed and used for the energy estimation study. Hits from signal and K⁴⁰ background are considered. The muon tracks under consideration are <u>crossing the detector volume</u>.



- A good purity (\geq **80%**) of the reconstructed events for $E_{\mu} \geq 10 TeV$ can be attained, with an efficiency of **55%** for $E_{\mu} \geq 10 TeV$.
- The efficiency is \geq **80%** for $E_{\mu} \geq$ 40 *TeV* rising with energy to ~ **95%** for $E_{\mu} \geq$ 100 *TeV*.



MultiVariate Analysis



- Muon and Neutrino Energy Estimation: a Multi-Layer Percepton (MLP) Neural Network has been trained using information referring to:
- Number of photomultipliers (PMTs) with signal

(weighted considering the PMT distance from the reconstructed muon track)

- <u>Total Time over Threshold (ToT) in PMTs</u> (as a measure of charge in PMTs)
- <u>Number of OMs with signal</u>

(weighted taken into account that muons with lower energies travel shorter distances inside the detector than muons with higher energies)

<u>Number of PMTs without signal</u>

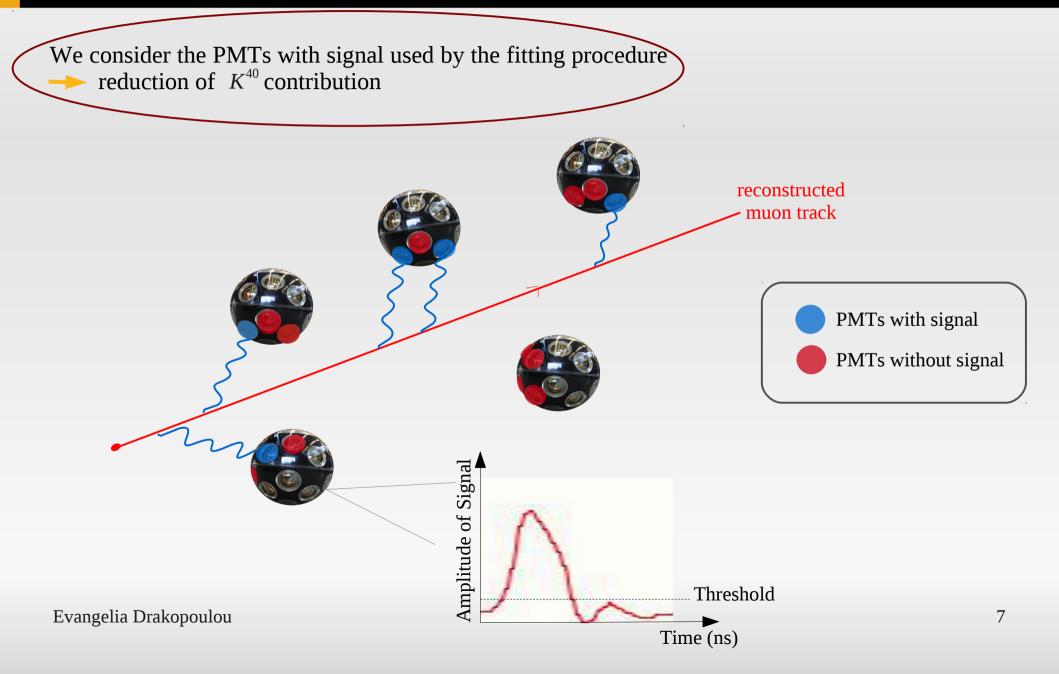
(weighted considering that the number of PMTs that have no signal is larger for muons with lower energies)

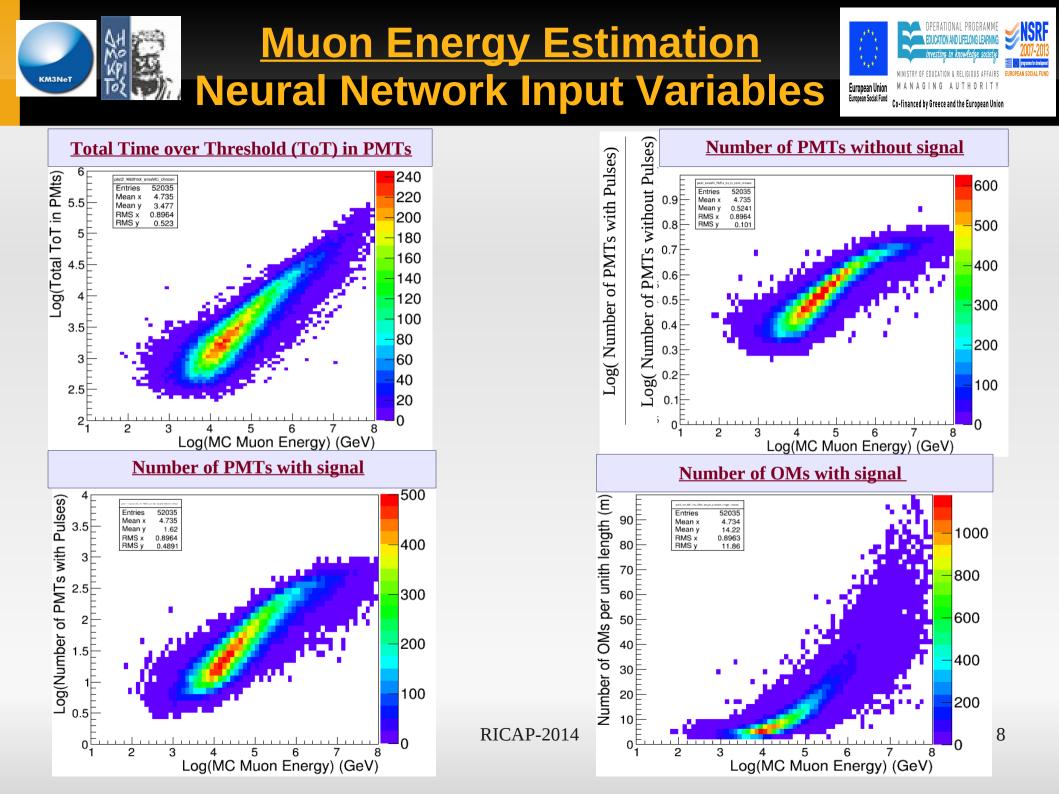
• A minimum muon track length inside the detector volume is required in order to estimate muon and neutrino energy.



MultiVariate Analysis



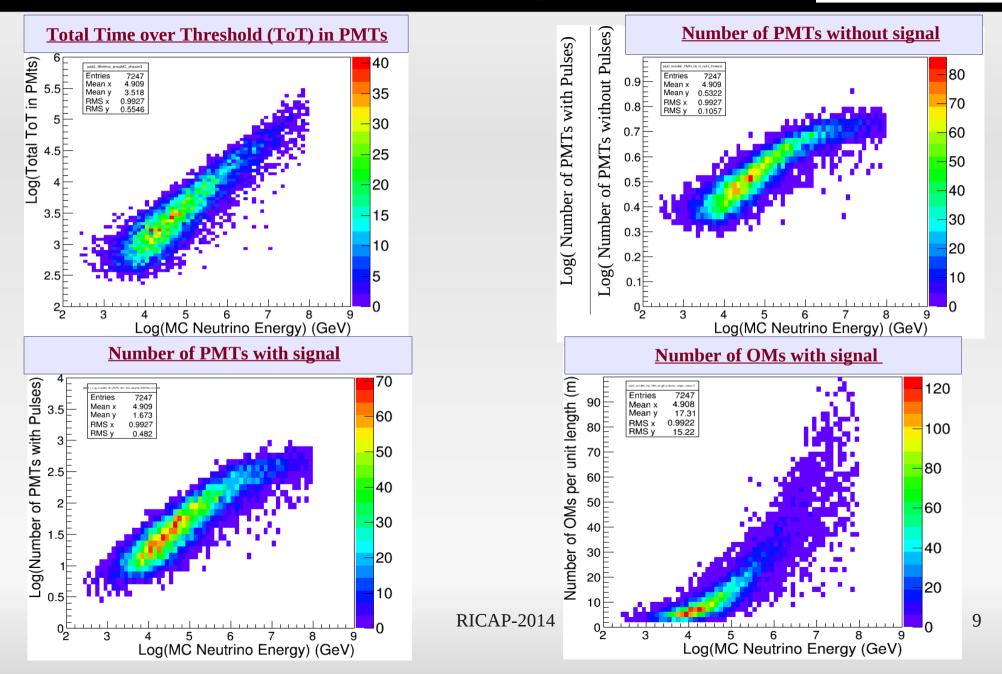


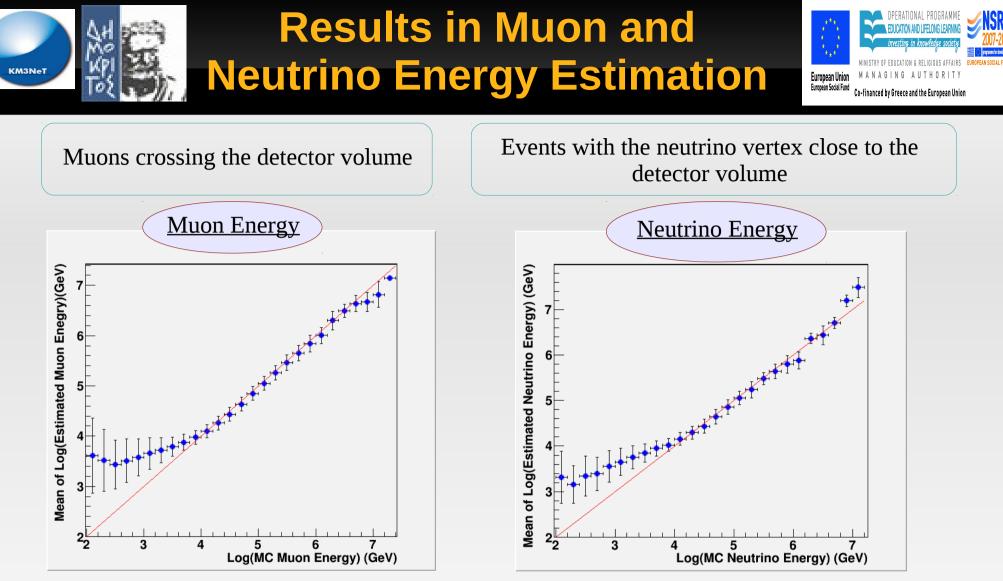


Neutrino Energy Estimation Neural Network Input Variables

KM3NeT







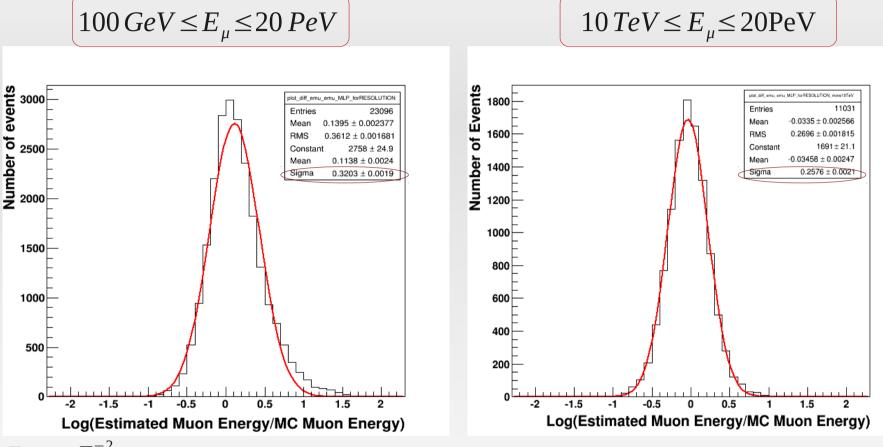
• For muon energies $E_{\mu} < 1 TeV$ we overestimate the energy, while there is a very good linear correlation for $E_{\mu} \ge 10 TeV$.

• The neutrino energy is better determined (with respect to the muon energy) for the lower energy range. There is also a very good agreement for $E_v \ge 10 \, TeV$.



Muon Energy Resolution

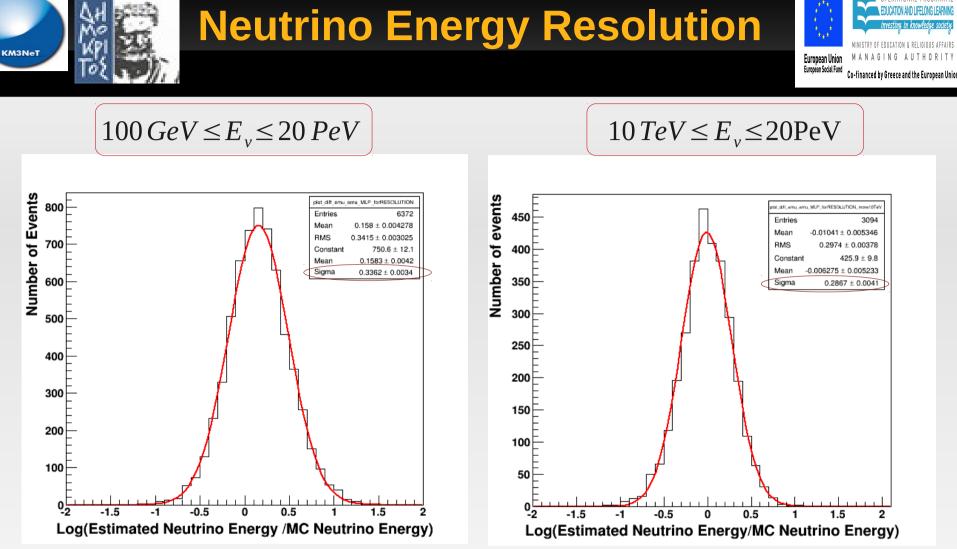




For an E^{-2} spectrum:

- An energy resolution of ~ 0.32 has been achieved for all events for which the energy was estimated.
- The energy resolution is ~ 0.26 for $E_{\mu} \ge 10 \, TeV$.

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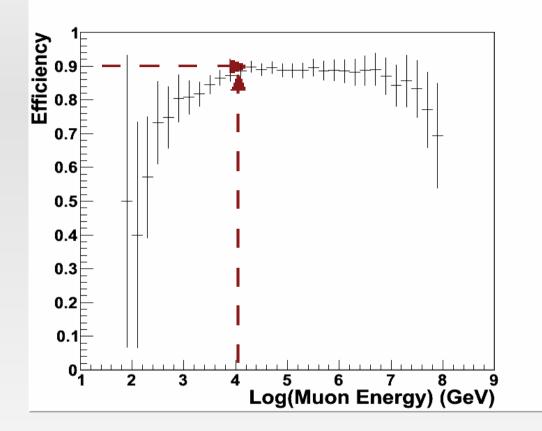
For an E^{-2} spectrum:

- An energy resolution of ~ 0.34 has been achieved for all events for which the energy was estimated.
- The energy resolution is ~ 0.29 for $E_v \ge 10 \, TeV$.



Efficiency of the Energy Reconstruction





- A very high efficiency for the energy reconstruction is achieved of ~ **88%** for all events crossing the detector volume.
- The efficiency is ~ **89%** for $E_{\mu} \ge 10 \, TeV$ peaking at ~ **90%** for $E_{\mu} \simeq 20 \, TeV$.





- A new method for the muon and neutrino energy estimation using a Multi-Layer Percepton Neural Network with appropriate input variables was presented.
- The performance of the energy estimator is very good, particularly in the high energy region ($E_{\mu} \ge 10 TeV$) which is the energy regime we are mostly interested in.
- The energy resolution is about 0.32 in $\log_{10} E_{\mu}$ for muons over all energies and becomes **0.26** for $E_{\mu} \ge 10 \, TeV$.
- The energy resolution is about 0.34 in $\log_{10} E_v$ for neutrinos over all energies and becomes **0.29** for $E_v \ge 10 TeV$.





Backup Slides

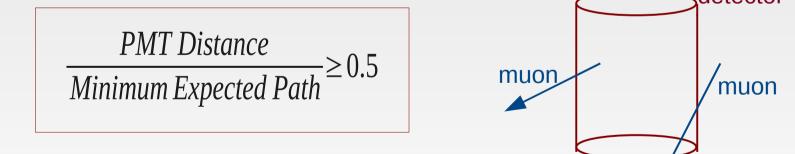


Energy Reconstruction Selecting Cuts



Minimum Expected Path = $0.5^{\circ}h + (R - 0.5^{\circ}h) * \sin(\theta rec)$

- For <u>Horizontal Muons</u> Minimum Expected Path is the <u>Detector Radius</u>
- For <u>Vertical Muons</u> Minimum Expected Path is <u>0.5*String height</u>
- <u>PMT Distance</u> : distance between first and last PMT positions
- The PMT Distance should be more than the half of minimum expected path or at least:



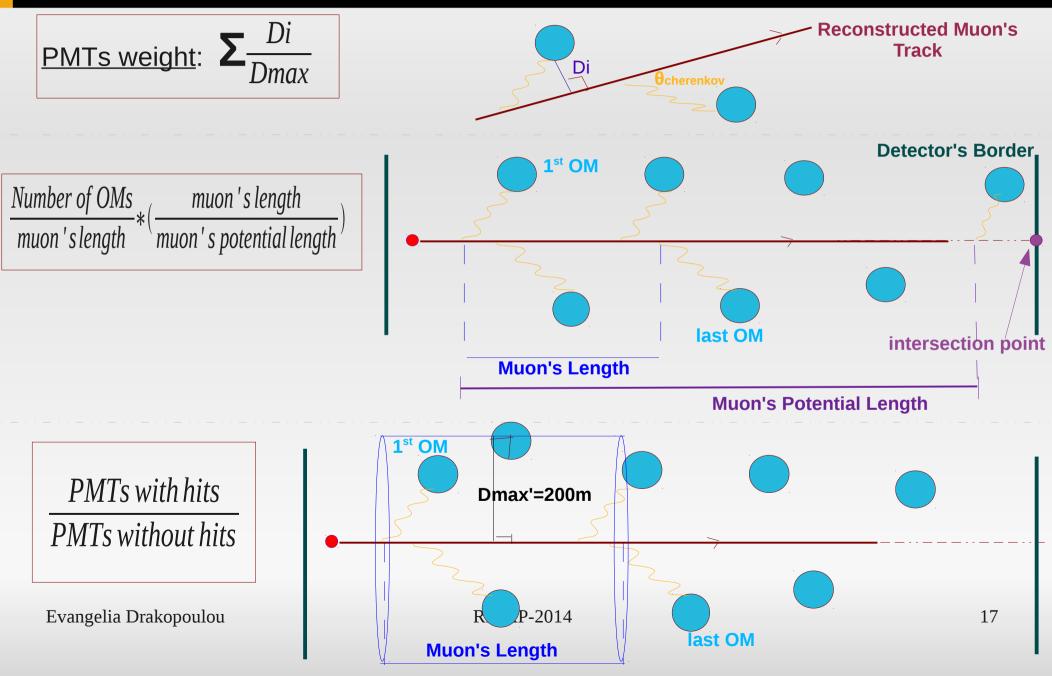
- The distances between OMs are different in horizontal and vertical direction
- minimum expected path length should change with respect to the muon zenith

<u>where</u>: **h**: string's height, **R**: detector radius, θ rec: reconstructed muon angle

Explaining Selection Cuts



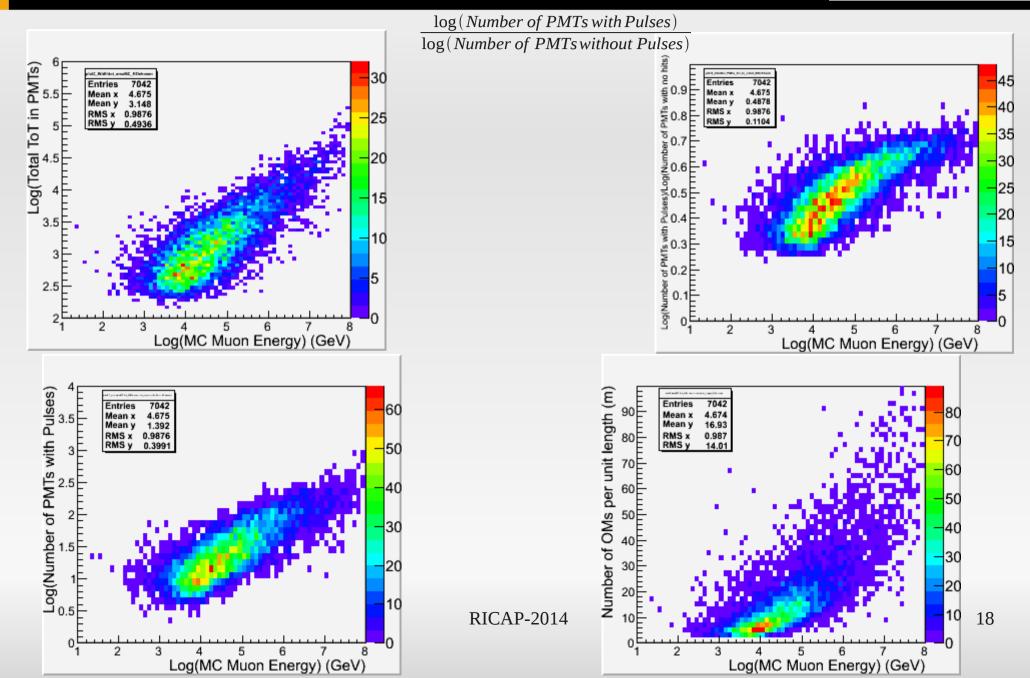






Events that fail to satisfy the selection criteria

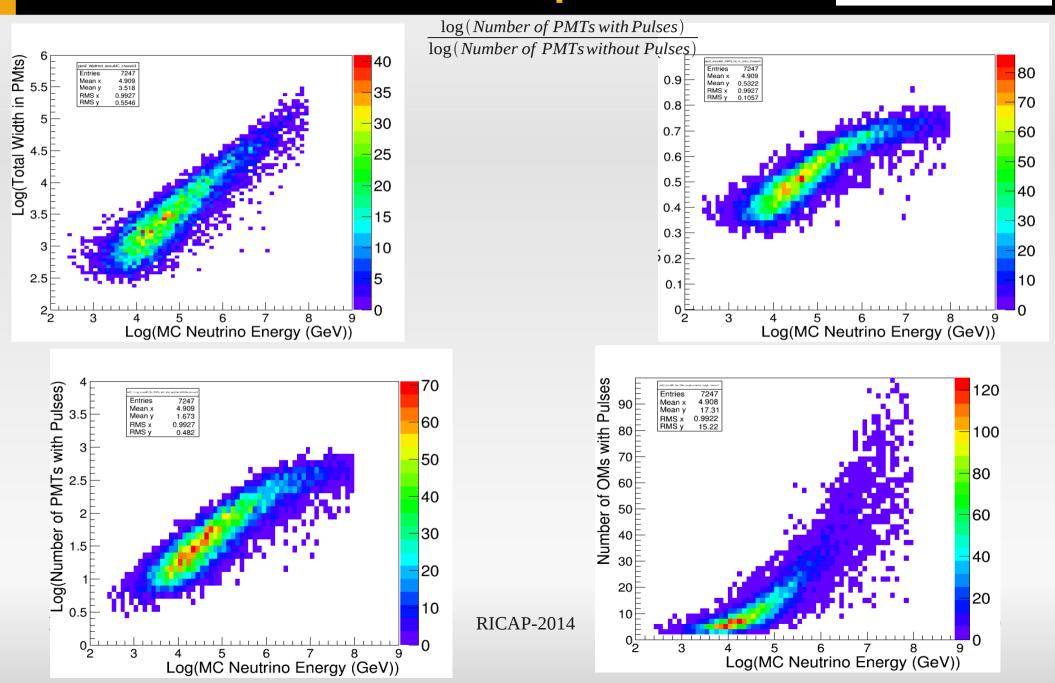




Neutrino Energy Estimation Neural Network Input Variables

KM3NeT

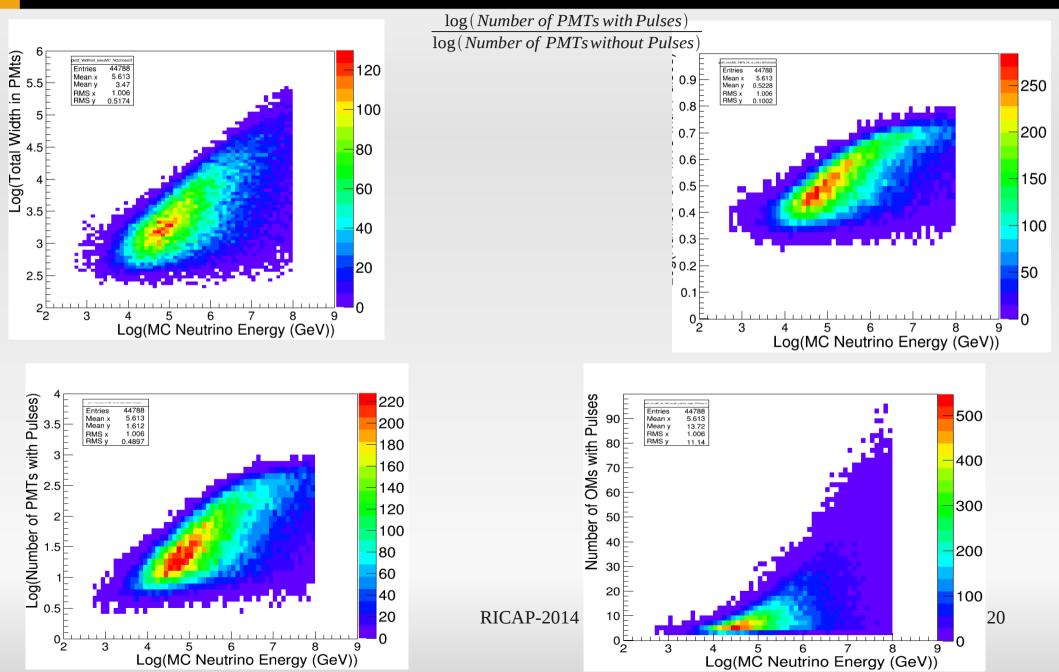
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Events that fail to satisfy the selection criteria

KM3NeT

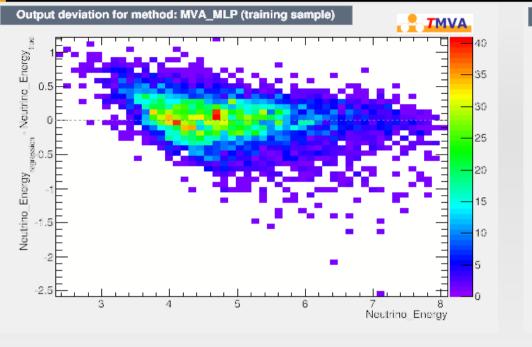




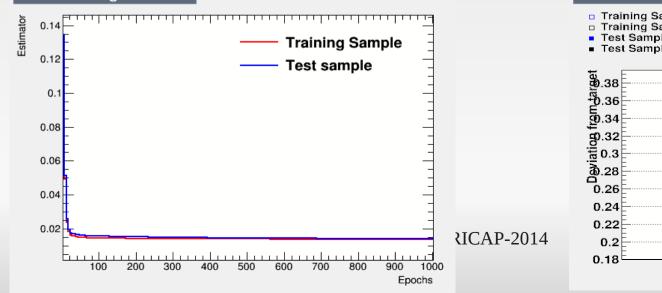


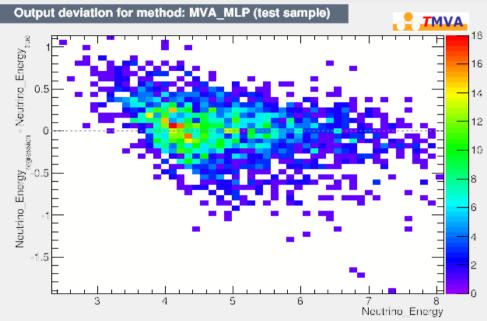
Neutrino Energy Estimation MLP Performance

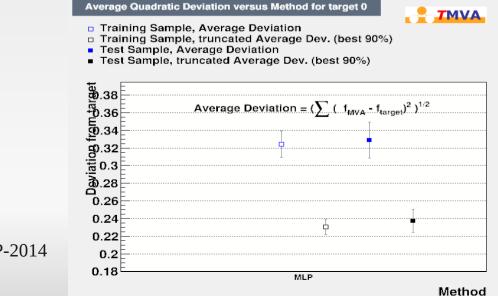




MLP Convergence Test



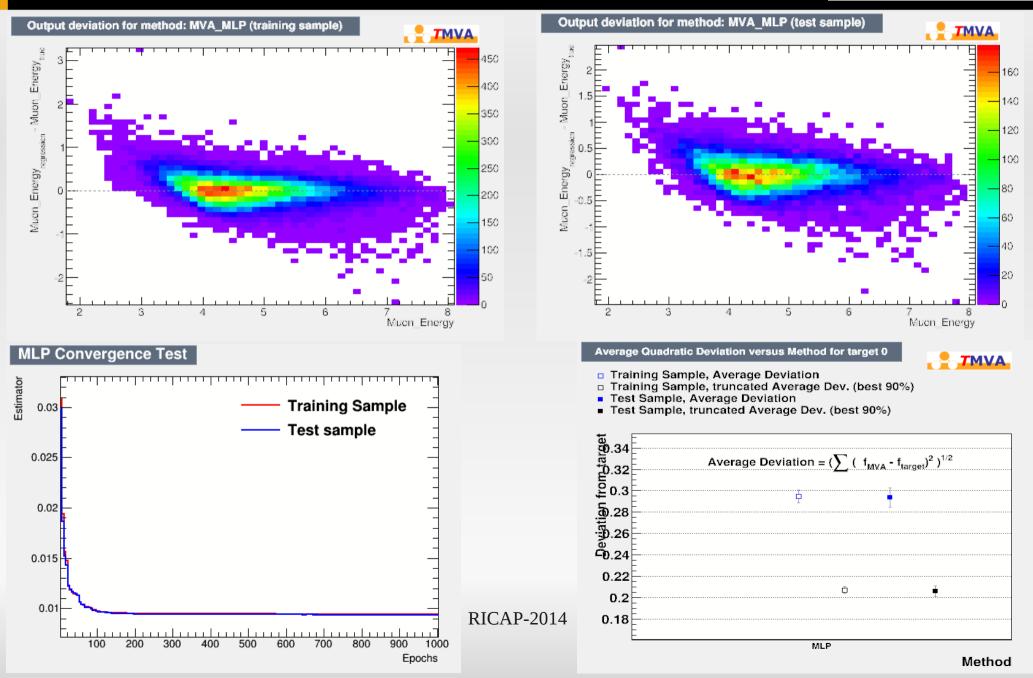






Muon Energy Estimation MLP Performance



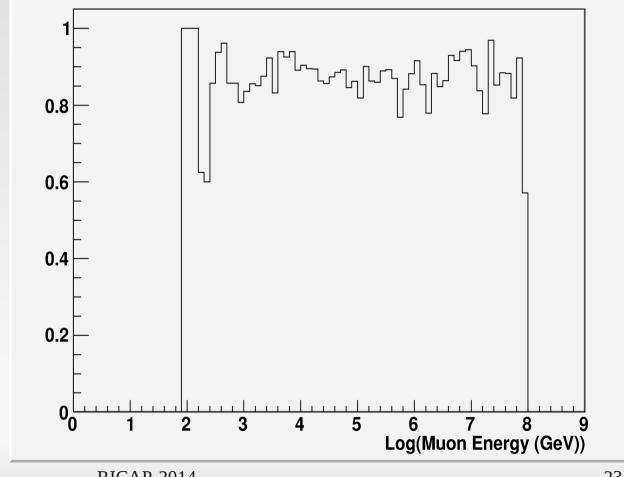




Well Reconstructed Events



Selected Well Reconstructed Events for Energy Estimation Well Reconstructed Events



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- The energy is one of the most important parameters to <u>discriminate between atmospheric</u> <u>and astrophysical events</u> recorded by neutrino telescopes.
- Muons lose energy via :
- Ionization
- Stochastic processes

bremsstrahlung pair production photonuclear interactions

$$\frac{-dE_{\mu}}{dx} = A(E) + B(E)E_{\mu}$$

where: **A** \approx 0.274 *GeV* m^{-1} accounts for the energy loss due to ionization, and **B** \approx 0.000349 m^{-1} is due to the stochastic energy loss.

<u>Muon's Energy Estimation</u> :

- Low Energy Muons
- + muon length
- High Energy Muons
- light observation in PMTs

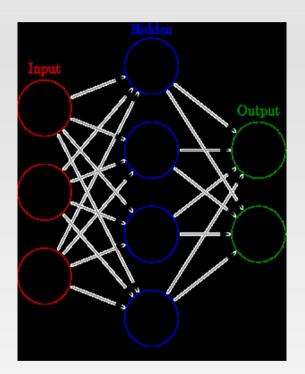




if (Use["MLP"]) factory->BookMethod(TMVA::Types::kMLP, "MLP", "!H:!V:VarTransform=Norm :NeuronType=tanh:NCycles=1000:HiddenLayers=N+6:TestRate=5:TrainingMethod=BFGS : !UseRegulator");

Backup Slides – MVA Methods





BFGS Training Method at MLP

The Broyden-Fletcher-Goldfarb-Shannon (BFGS) method differs from back propagation by the use of second derivatives of the error function to adapt the synapse weight by an algorithm which is composed of four main steps.

→ During the learning process the network is supplied with N training events $x_a = (x1, ..., xnvar)_{a,a} = 1, ..., N$. For each training event a the neural network output $y_{ANN,a}$ is computed and compared to the desired output y_a . The error function E, is measuring the agreement of the network response with the desired one. The set of weights that minimises the error function can be found using the method of steepest or gradient descent, provided that the neuron response function is differentiable with respect to the input weights.

The advantage of the BFGS method compared to BG is the smaller number of iterations.

Backup Slides – MVA Methods



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(Configuration options for MVA method :							ignor energineignearmaining	110	TODA:		performance evaluation)	ſ	2
	Configuration options reference for MVA method: MLP											Train with Back-Propagation (BP), BFGS		
	Option	Array	Default value	Predefined values	Description		L	TrainingMethod	No	BP	BP, GA, BFGS	Algorithm (BFGS), or Genetic Algorithm (GA - slower and worse)		
1	lCycles	No	500	-	Number of training cycles		L	LearningRate	No	0.02	-	ANN learning rate parameter		
	HiddenLayers	No	N,N-1	-	Specification of hidden layer architecture			DecayRate	No	0.01	-	Decay rate for learning parameter		
	NeuronType	No	sigmold	-	Neuron activation function type		1	TestRate	No	10	_	Test for overtraining performed at each #th		
	RandomSeed	No	1	-	Random seed for initial synapse weights (0 means unique seed for each run; default value '1')				No		-	epochs Provide epoch-wise monitoring plots according		
- 1	EstimatorType	No	MSE	MSE, CE, linear, sigmoid, tanh, radial	MSE (Mean Square Estimator) for Gaussian Likelihood or CE(Cross-Entropy) for Bernoulii			EpochMonitoring	No	False	-	to TestRate (caution: causes big ROOT output file!) Only 'Sampling' (randomly selected) events an		
					Likelihood			Sampling	No	1	-	trained each epoch		
. !	NeuronInputType	No	sum	sum, sqsum, abssum	Neuron input function type			Franklin Frank		1	-	Sampling is used for the first 'SamplingEpoch'		
	/	No	False	-	Verbose output (short form of VerbosityLevel below - overrides the latter one)			SamplingEpoch	No			epochs, afterwards, all events are taken for training		
	/erbosityLevel	No	Default	Default, Debug, Verbose, Info, Warning, Error, Fatal				SamplingImportance	No	1	-	The sampling weights of events in epochs which successful (worse estimator than before) are multiplied with Samplingimportance, else they are divided.		
	VarTransform	No	None	-	List of variable transformations performed before training, e.g., D_Background,P_Signal,G,N_AllClasses for: Decorrelation, PCA-transformation, Gaussianisation, Normalisation, each for the		L	SamplingTraining	No	True	-	The training sample is sampled		
								SamplingTesting	No	False	-	The testing sample is sampled		
							L	ResetStep	No	50	-	How often BFGS should reset history		
					given class of events ('AliClasses' denotes all events of all classes, if no class indication is			Tau	No	3	-	LineSearch size step		
	4	No	False	-	given, 'All' is assumed) Print method-specific help message	=		BPMode	No	sequential	sequential, batch	Back-propagation learning mode: sequential or batch		
	CreateMVAPdfs	No	False	-	Create PDFs for classifier outputs (signal and background)	_		Batc	No	-1	-	Batch size: number of events/batch, only set if i Batch Mode, -1 for BatchSize=number_of_event		
	gnoreNegWeightsinTraining	No	False	-	Events with negative weights are ignored in the training (but are included for testing and performance evaluation)			ConvergenceImprove	No	1e-30	-	Minimum improvement which counts as improvement (+0 means automatic convergence check is turned off)		
	rainingMethod	No	BP	BP, GA, BFGS	Train with Back-Propagation (BP), BFGS Algorithm (BFGS), or Genetic Algorithm (GA - slower and worse)			ConvergenceTests	No	4	-	Number of steps (without improvement) required for convergence (+0 means automatic convergence check is turned off)		
	earningRate	No	0.02	-	ANN learning rate parameter			UseRegulator	No	False	-	Use regulator to avoid over-training		
	DecayRate	No	0.01	-	Decay rate for learning parameter			UpdateLimit	No	10000	-	Maximum times of regulator update		
	lestRate	No	10	-	Test for overtraining performed at each #th epochs			CalculateErrors	No	False	-	Calculates inverse Hessian matrix at the end of the training to be able to calculate the uncertainties of an MVA value		
	EpochMonitoring	No	False	-	Provide epoch-wise monitoring plots according to TestRate (caution: causes big ROOT output file!)			WeightRange	No	1	-	Take the events for the estimator calculations from small deviations from the desired value to large deviations only over the weight range		•
Sampling		No	1	-	Only 'Sampling' (randomly selected) events are trained each epoch	~		Configuration options for MVA method :					. (~

TMVA Options Referen...

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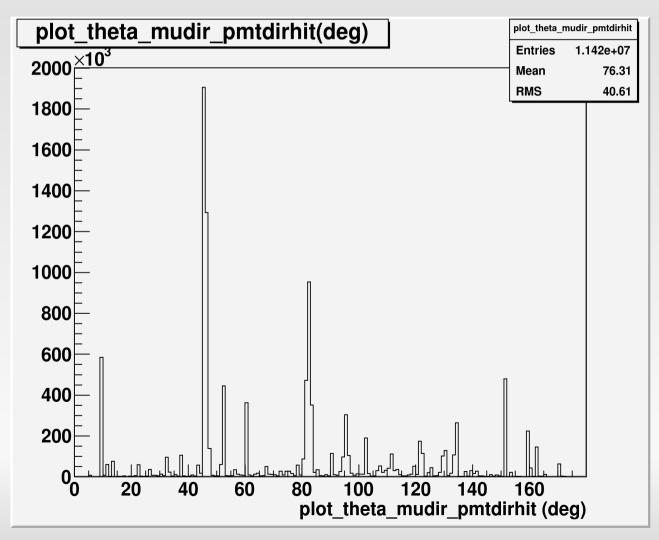
KM3NeT

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Counting PMTs that could have hits but they don't \rightarrow checking only PMTs whose angle between their direction with the reconstructed track is more than 6 degrees and less than 86 degrees



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