APPEARANCE PROBABILITY MODEL OF ELECTRON ANTI-NEUTRINOS ACCOUNTING FOR DIFFERENT REACTOR DISTANCES

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- 56 nuclear reactors and one detector
- Detector is located on the island of Honshu, Japan
- Each nuclear reactor contains Uranium 235 and 238 & Plutonium 239 and 241
- □ Fission occurs:
 - 57.1% from U 235
 - 7.8% from U 238
 - 29.5% from Pu 239
 - 5.6% from Pu 241



Source: http://kamland.lbl.gov/Pictures/kamland-ill.html

The reactors:



The detector:



Source: http://kamland.lbl.gov/Pictures/kamland-ill.html

- The Liquid Scintillator inside the detector contains C9 H12 (pseudocumene) and C12 H26 (dodecane)
- Some of the anti-neutrinos coming from the reactors collide with protons found in these molecules
- Inverse beta decay



The detector:

CH₃ A visible photon is **Fluorescence** emitted from the Η CH_3 molecule The UV photon hits one of the molecules and is absorbed Optical Η photon Η CH_3 DETECTED The visible photon is detected by all the photomultiplier tubes



as only one

Simultaneously in the detector...



The detector:





Source: kamland.lbl.gov/Pictures/picgallery.html

Number of Counts

Our main equation:

$$N(E_1 < E_p < E_2) = \sum_{1}^{M} \int_{E_1}^{E_2} \int_{1.8}^{10} \frac{S_i(E_\nu)}{4\pi L_i^2} \sigma(E_\nu) T(E_p, E_\nu) P_\nu(\frac{L_i}{E_\nu}) dE_\nu dE_p$$



A simple derivation



A simple derivation

$$N(E_1 < E_p < E_2) = \sum_{1}^{M} \int_{E_1}^{E_2} \int_{1.8}^{10} \frac{S_i(E_\nu)}{4\pi L_i^2} \sigma(E_\nu) T(E_p, E_\nu) P_\nu(\frac{L_i}{E_\nu}) dE_\nu dE_p$$

$$(Reactions)T(E_p, E_\nu) = Flux * \sigma_p N_p T(E_p, E_\nu)$$

where,
$$Flux = \frac{S_i(E_\nu)P_\nu(\frac{L_i}{E_\nu})}{4\pi L_i^2}$$

Terms

$$N(E_1 < E_p < E_2) =$$

The number of counts at each energy prompt

$$\frac{S_i(E_\nu)}{4\pi L_i^2} P_\nu(\frac{L_i}{E_\nu}) =$$

The flux of anti-neutrinos expected at the detector

$$P_{\nu}(\frac{L_i}{E_{\nu}}) =$$

Probability that an electron anti-neutrino will stay an electron anti-neutrino by the time it reaches the detector

$$\sigma(E_{\nu}) =$$

The cross section of one proton that could interact with the antineutrinos coming into the detector

 $\frac{S_i(E_\nu)}{4\pi L_i^2}\sigma(E_\nu)P_\nu(\frac{L_i}{E_\nu}) =$

The number of reactions

$$T(E_p, E_\nu) =$$

Probability of detecting a reaction from the reactions that have occurred (due to experimental error)



KamLAND events graph

Events graph with what was observed in the detector

http://www.awa.tohoku.ac.jp/KamLAND/4th_result_data_release/4th_result_data_release.html



KamLAND appearance probability graph

Appearance probability from an average reactor length

http://arxiv.org/pdf/1009.4771v2.pdf

Our theoretical research

$$N(E_1 < E_p < E_2) = \sum_{i=1}^{M} \int_{E_1}^{E_2} \int_{1.8}^{10} \frac{S_i(E_\nu)}{4\pi L_i^2} \sigma(E_\nu) T(E_p, E_\nu) P_\nu(\frac{L_i}{E_\nu}) dE_\nu dE_p$$

Make a change of variables:

$$l = \frac{L_i}{E_\nu}$$

$$N(E_{1} < E_{p} < E_{2}) = \sum_{i=1}^{M} \int_{E_{1}}^{E_{2}} \int_{l=0}^{l=\infty} dl dE_{p} \frac{1}{4\pi L_{i} l^{2}} S_{i}(\frac{L_{i}}{l}) \sigma(\frac{L_{i}}{l}) T(E_{p}, \frac{L_{i}}{l}) P_{\nu}(l)$$

$$P$$

$$Q$$

Our theoretical research

$$N(E_p) = Q(E_p, l)P(l)$$

For simplicity, we shall call this: N = Q P what we want to find empirically We need to minimize $\chi^2_{
m for N(Ep):}$ $\chi^2 = \sum_{i=1}^n \frac{(N_i - \bar{N})^2}{\sigma_i^2}$ where, $\sigma = \sqrt{\bar{N}}$



Our theoretical research

$$\chi^{2} = (N - \bar{N})^{T} V^{-1} (N - \bar{N})$$
where, $V^{-1} = \begin{pmatrix} \frac{1}{\sigma_{1}^{2}} & 0 & 0\\ 0 & \frac{1}{\sigma_{2}^{2}} & 0\\ 0 & 0 & \frac{1}{\sigma_{3}^{2}} \end{pmatrix}$
By taking the derivative of χ^{2} and χ^{2}

transformation of the Q matrix

d setting it equal to zero, we get

Our observed values $V^{-1}Q)^{-1}Q^TV^{-1}$ P

Small proof

 $(Q^T V^{-1} Q)^{-1} Q^T V^{-1} N = P$



Why we want to do this

- Prove neutrino oscillations and KamLAND's conclusions empirically
- Gain knowledge about how neutrinos behave, which could lead to a better understanding of dark matter
- Gain knowledge about neutrinos to be able to control nuclear reactors efficiently by monitoring neutrinos that leave

Forming the Q matrix

For ex:

$$\begin{pmatrix} N(E_{p1}) \\ N(E_{p2}) \\ N(E_{p3}) \\ N(E_{p4}) \end{pmatrix} = \begin{pmatrix} Q(E_{p1}, l_1) & Q(E_{p1}, l_2) & Q(E_{p1}, l_3) \\ Q(E_{p2}, l_1) & Q(E_{p2}, l_2) & Q(E_{p2}, l_3) \\ Q(E_{p3}, l_1) & Q(E_{p3}, l_2) & Q(E_{p3}, l_3) \\ Q(E_{p4}, l_1) & Q(E_{p4}, l_2) & Q(E_{p4}, l_3) \end{pmatrix} \begin{pmatrix} P(l_1) \\ P(l_2) \\ P(l_3) \end{pmatrix}$$

- Test for as many I's as possible, binning them
- If the E_{ν} lies between 1.8 MeV-10 MeV, then plug the values into the Q equation
- If the E_{ν} lies outside of that range, it does not contribute to the detector, so we input zero for that matrix element
- Obtain a different Q matrix for each reactor
- Superpose all the Q matrices

Forming the Q matrix



Our 'no oscillations' graph



Setting up the test

$$N_0 = Q_0 P_0$$

where, $P_0 = -$ appearance probability if there were no oscillations

$$(\begin{bmatrix} Q_0'^T V^{-1} Q_0' \\ 0 \end{bmatrix})^{-1} Q_0'^T V^{-1} N = P$$

$$C \qquad Y$$
where,
$$V^{-1} = \begin{pmatrix} \frac{1}{N_{0_1}} & 0 & 0 & \cdots & 0 \\ 0 & \frac{1}{N_{0_2}} & 0 & \cdots & 0 \\ 0 & 0 & \frac{1}{N_{0_3}} & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \frac{1}{N_{0_{17}}} \end{pmatrix}$$

Setting up the test

Since,

Where R is a matrix containing the orthonormal eigenvectors for each eigenvalue, and D is a diagonal matrix containing all the eigenvalues of C

$$\begin{split} C &= C^{T} \\ C &= RDR^{T} \\ \text{the} & C^{-1} = R^{T^{-1}}D^{-1}R^{-1} \\ \text{ad} \\ \text{alues} & \tilde{C}^{-1} = R^{T^{-1}}\tilde{D}^{-1}R^{-1} \\ N_{1true} &= Q_{0}P_{1} \end{split} \text{Has the smallest eigenvalue} \\ N_{1true} &= N_{0}P_{1} \qquad \text{Has the smallest eigenvalue} \\ N_{1observed} &= N_{1true}' + \eta_{noise} \end{split}$$

$$\tilde{C}^{-1}YN'_{1observed} \approx P_1$$

The Binning

Why bin the Ep's?

- The greater the counts per bin, the smaller the relative error
- As a result, approaches a Gaussian

Why bin the I's?

- More functions than unknowns
- A higher sum in each I column will provide for a smaller error

Why find the eigenvalues of C?

- If product of eigenvalues is big, error is small when inverting C
- If difference is big, magnifies error

Why test it this way

Accounting for bias by using No prime to calculate V inverse instead of N1 prime:

- This method gives each element in N1 prime their corresponding importance according to how many number of counts they each contribute and, therefore, how much data they contain



$$V^{-1}N'_{0} = \begin{pmatrix} 1\\1\\1\\1\\1 \end{pmatrix} \quad \text{while} \quad V^{-1}N'_{1} = \begin{pmatrix} 0.97\\1.00\\0.10\\1.32\\0.61 \end{pmatrix}$$

Why omit the smallest eigenvalue?

$$P_{1} \approx \tilde{C}^{-1} Y N_{1}^{\prime} \qquad \underset{\text{background noise}}{\overset{\text{Contains}}{\underset{\text{background noise}}{\overset{\text{contains}}{\underset{\text{background noise}}{\overset{\text{contains}}{\underset{\text{background noise}}{\overset{\text{contains}}{\underset{\text{background noise}}{\overset{\text{contains}}{\underset{\text{background noise}}{\overset{\text{contains}}{\underset{\text{background noise}}}}}}$$

Contains inverse eigenvalues

$$P_1 \approx \tilde{C}^{-1} Y N'_{1true} + \tilde{C}^{-1} Y \eta_{noise}$$

The smaller the eigenvalue, the more noise error it contributes

Error

Biggest error contributors:

- Background noise in the data N1 from the experiment
- Approximation of I values due to the I binning in Q0

Producing reasonable error bars for our test of specific P(I)s :

- □ Create N'1 true with a specific P(I)
- □ Add randomized background noise to N'1true
- Create 1000 different P(I)s, each using a different randomized N'_{1observed}
- Find the average P(I) and its standard deviation to obtain different error bars for each P(I) entry



Omitting the smallest eigenvalue

16 Ep bins, 11 | bins

Smallest error bars so far, but not as small as were expected



Omitting vs not omitting smallest eigenvalue

16 Ep bins, 11 | bins

Smaller error bars

Testing KamLAND's N



Comparing Chi Squares

Chi square of the N between our estimate and the N observed:	$\begin{pmatrix} 50\\ 88\\ 106 \end{pmatrix}$	(0.1) (3.2) (6.8)	$\begin{pmatrix} 53.1\\ 85.5\\ 102.5 \end{pmatrix}$
6.68	73	9.7 0.3	81.4 83.3
Chi square of the P between our	92	2.5	84.1 65.4
estimate and the closest straight line	65	5.8 N	69.8
of 0.44 without taking into account covariance:	$N_1 = \begin{bmatrix} 83\\97 \end{bmatrix}$	$X_{1_{estimate}} = X_{1_{estimate}}$	= 87.8 89.0
9.65	137	7.3 9.4	$129.1 \\ 136.0$
The above chi square with covariance:	126	6.2	126.5
67.95	169	9.0 2.2	157.0 115.5
	$\begin{pmatrix} 104\\ 102 \end{pmatrix}$	4.9 2.6	$\begin{pmatrix} 109.9\\ 99.2 \end{pmatrix}$

The Covariance Matrix



Conclusions

- Obtained appearance probabilities for 11 values of L/Ev without assuming an average L
- Appearance probability cannot be constant
- Predicted N matched KamLAND's observed N

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