

**Metropolis-Hastings Approach For Continuous Wave and Inspiral
Searches**

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In Collaboration with Réjean Dupuis and Graham Woan

Bayesian Inference, Posterior Computation and Parameter Estimation

Antenna output data $\mathbf{z}(\mathbf{t})$

Joint probability distribution function (PDF) $p(\mathbf{z}|\boldsymbol{\theta})$ conditional on unobserved parameters $\boldsymbol{\theta} = (\theta_1, \dots, \theta_d)$.

The PDF $p(\mathbf{z}|\boldsymbol{\theta})$ is usually referred to as the *likelihood* and regarded as a function of the parameters $\boldsymbol{\theta}$.

Bayes' Theorem: Condition on the post-experimental knowledge about $\boldsymbol{\theta}$ expressed through the *posterior* PDF

$$p(\boldsymbol{\theta}|\mathbf{z}) = \frac{p(\boldsymbol{\theta})p(\mathbf{z}|\boldsymbol{\theta})}{m(\mathbf{z})} \propto p(\boldsymbol{\theta})p(\mathbf{z}|\boldsymbol{\theta}) \quad (1)$$

$m(\mathbf{z}) = \int p(\mathbf{z}|\boldsymbol{\theta})p(\boldsymbol{\theta})d\boldsymbol{\theta}$ the marginal PDF of \mathbf{z} and regarded as a normalizing constant.

Posterior PDF is proportional to product of prior and likelihood.

The point estimate of a single parameter, say θ_i , is the posterior mean

$$[\theta_i] = \int \theta_i p(\theta_i|\mathbf{z})d\theta_i \quad (2)$$

where

$$p(\theta_i|\mathbf{z}) = \int \dots \int p(\boldsymbol{\theta}|\mathbf{z})d\theta_1 \dots d\theta_{i-1}d\theta_{i+1} \dots d\theta_d. \quad (3)$$

is the marginal posterior PDF.

The Metropolis-Hastings Algorithm: Candidate is generated from an auxiliary PDF and accepted or rejected with some probability.

Candidate generating PDF, $q(\boldsymbol{\theta}|\boldsymbol{\theta}_n)$ depends on the current state $\boldsymbol{\theta}_n$ of the Markov chain. A new candidate $\boldsymbol{\theta}'$ is accepted with a certain *acceptance probability* $\alpha(\boldsymbol{\theta}'|\boldsymbol{\theta}_n)$ also depending on the current state $\boldsymbol{\theta}_n$ given by:

$$\alpha(\boldsymbol{\theta}'|\boldsymbol{\theta}_n) = \min \left\{ \frac{p(\boldsymbol{\theta}')p(\mathbf{z}|\boldsymbol{\theta}')q(\boldsymbol{\theta}_n|\boldsymbol{\theta}')}{p(\boldsymbol{\theta}_n)p(\mathbf{z}|\boldsymbol{\theta}_n)q(\boldsymbol{\theta}'|\boldsymbol{\theta}_n)}, 1 \right\}$$

if $(p(\boldsymbol{\theta}_n)p(\mathbf{z}|\boldsymbol{\theta}_n)q(\boldsymbol{\theta}'|\boldsymbol{\theta}_n)) > 0$ and $\alpha(\boldsymbol{\theta}'|\boldsymbol{\theta}_n) = 1$ otherwise. For good efficiency a multivariate normal distribution is used for $q(\boldsymbol{\theta}|\boldsymbol{\theta}_n)$. The steps of the MH algorithm are therefore:

Step 0: Start with an arbitrary value $\boldsymbol{\theta}_0$

Step $n + 1$: Generate $\boldsymbol{\theta}'$ from $q(\boldsymbol{\theta}|\boldsymbol{\theta}_n)$ and u from $U(0, 1)$

If $u \leq \alpha(\boldsymbol{\theta}'|\boldsymbol{\theta}_n)$ set $\boldsymbol{\theta}_{n+1} = \boldsymbol{\theta}'$ (acceptance)

If $u > \alpha(\boldsymbol{\theta}'|\boldsymbol{\theta}_n)$ set $\boldsymbol{\theta}_{n+1} = \boldsymbol{\theta}_n$ (rejection)

The signals were generated by Réjean Dupuis. Data consisted of 14000 points for \mathbf{y}_k , the heterodyned signal. Signals had the following parameters:

$$\psi = 0.392699081 \quad \phi = 0.785398163 \quad \iota = 0.785398163$$

$$\text{RA} = 0.0 \quad \text{DEC} = 0.785398163$$

$$\sigma(\text{y.re}) = \sigma(\text{y.im}) = 1.0$$

The prior ranges used were:

$$\cos(\iota) \rightarrow [0, 1], \quad \psi \rightarrow [0, \pi/4], \quad \phi \rightarrow [0, 2\pi], \quad h \rightarrow [0, \text{inf})$$

For results see

<http://physics.carleton.edu/Faculty/Nelson/MCMCpulsar/MCMCpulsar.htm>

Signals had the following parameters:

$$\psi = 0.2 \quad \phi = 0.1 \quad \iota = 0.7$$

$$\text{RA} = 0.0 \quad \text{DEC} = 0.785398163$$

$$\sigma(\text{y.re}) = \sigma(\text{y.im}) = 1.0$$

The prior ranges used were:

$$\cos(\iota) \rightarrow [0, 1], \quad \psi \rightarrow [0, \pi/4], \quad \phi \rightarrow [-\pi, \pi], \quad h \rightarrow [0, \text{inf})$$

What Next?

Primary aim is to conduct a *DirectedSearch*

Uncertainty (small at first) in the frequency f_0

After that - other parameters such as \dot{f}_0

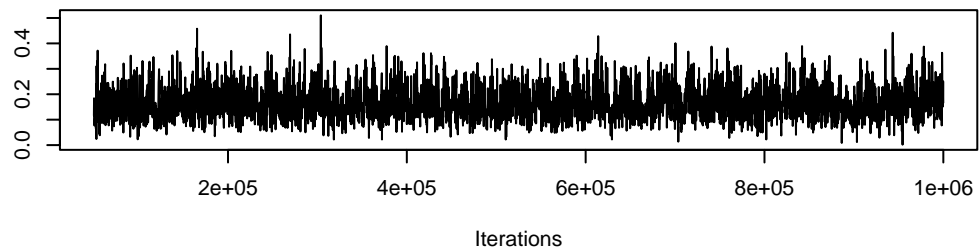
Sky Location???

Also Using Metropolis-Hastings
To Find Inspiral Events

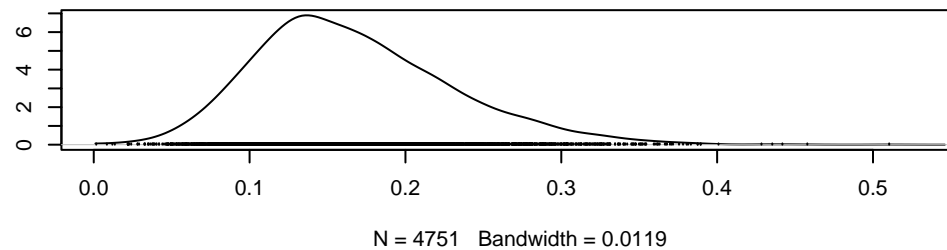
Example with $m_1 = 3.0$, $m_2 = 4.5$

$\eta = 0.24$, $m_t = 7.5$, $SNR = 14$

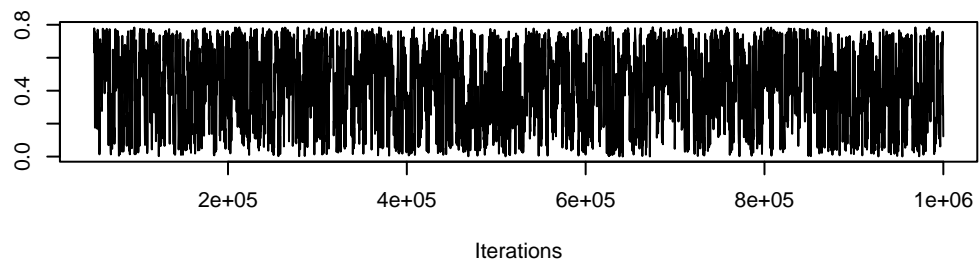
Trace of h0



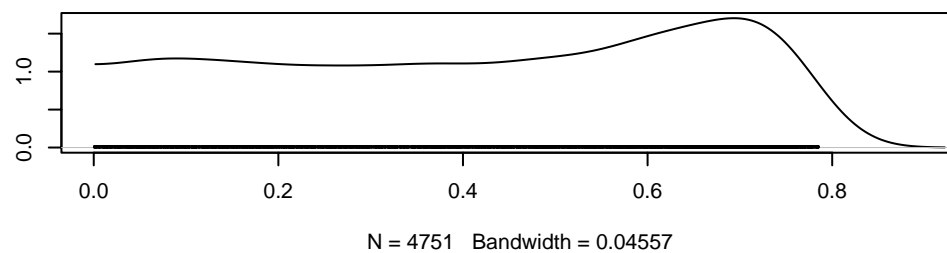
Density of h0



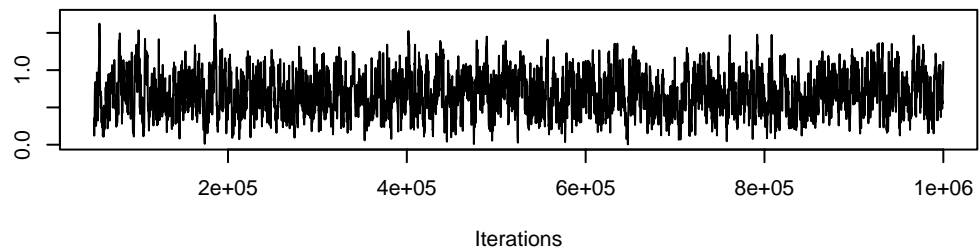
Trace of psi



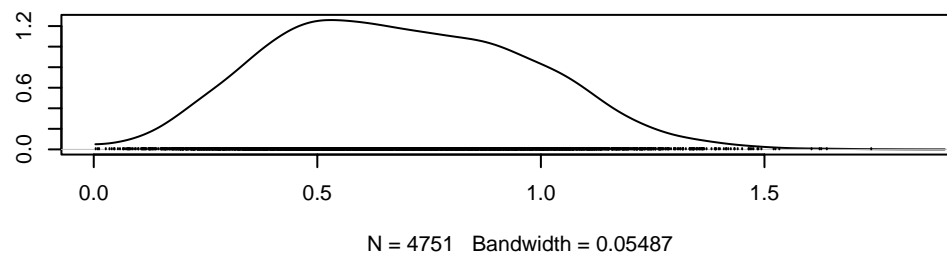
Density of psi



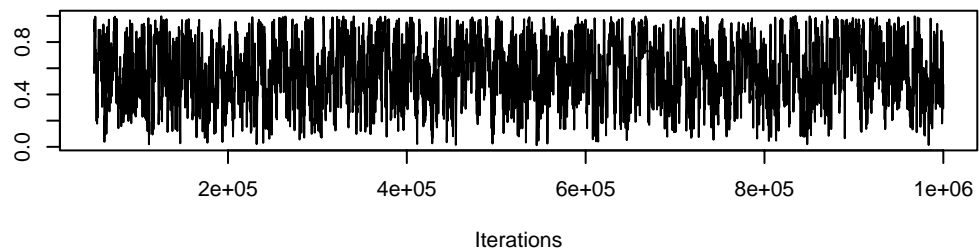
Trace of phase



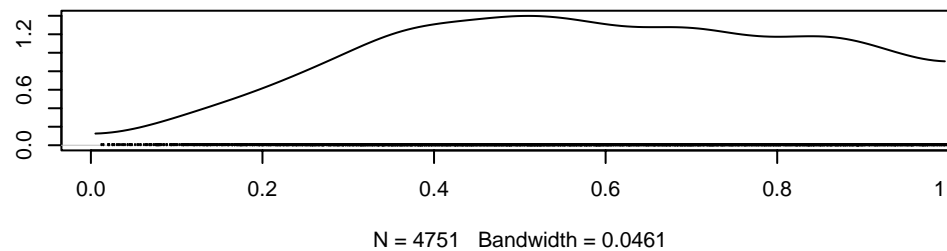
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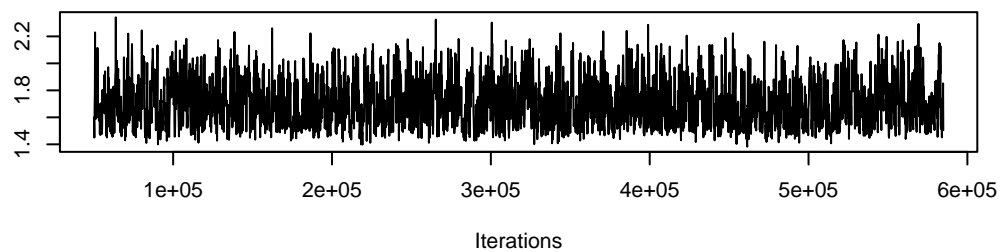
Trace of cosiota



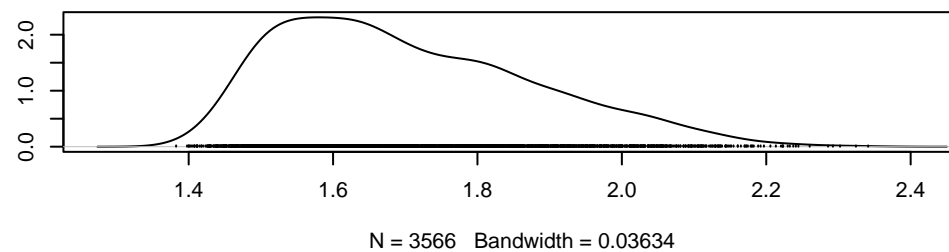
Density of cosiota



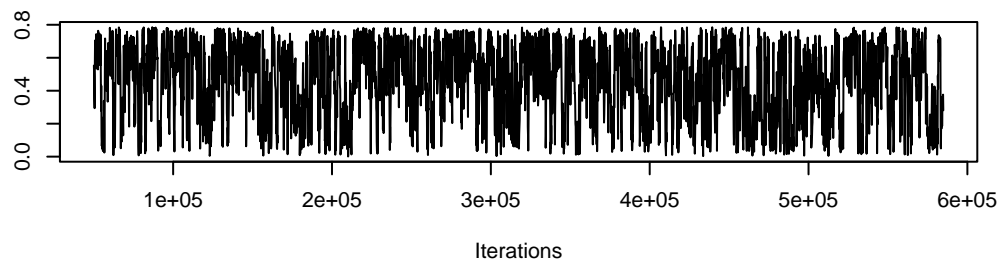
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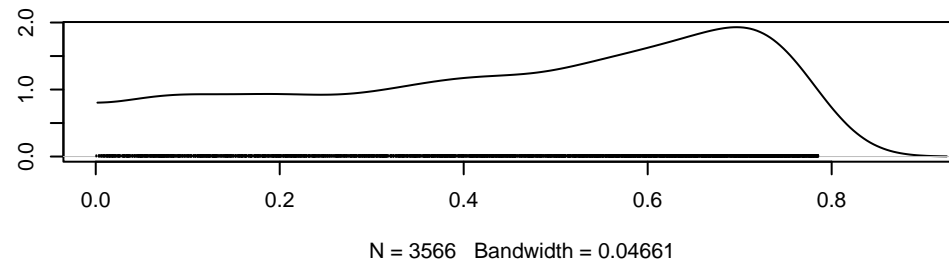
Density of h0



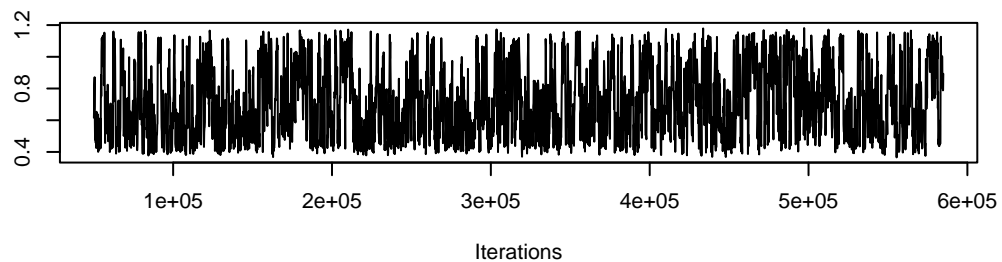
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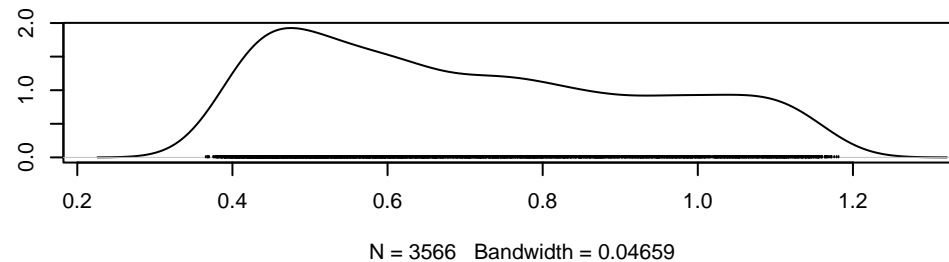
Density of psi



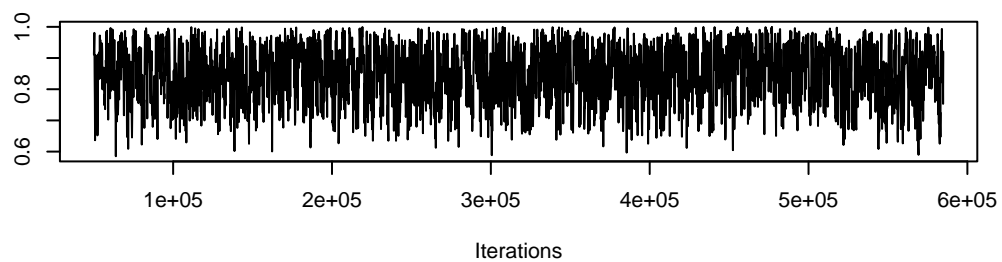
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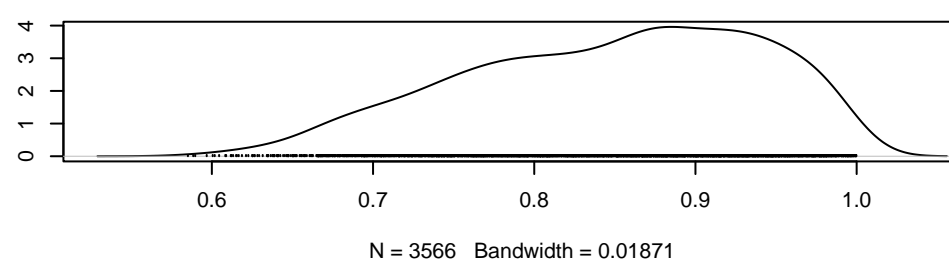
Density of phase



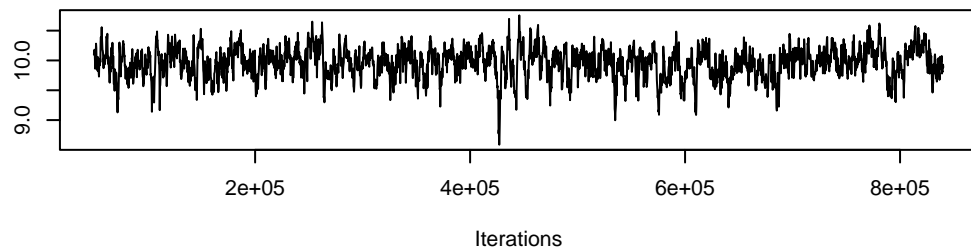
Trace of cosiota



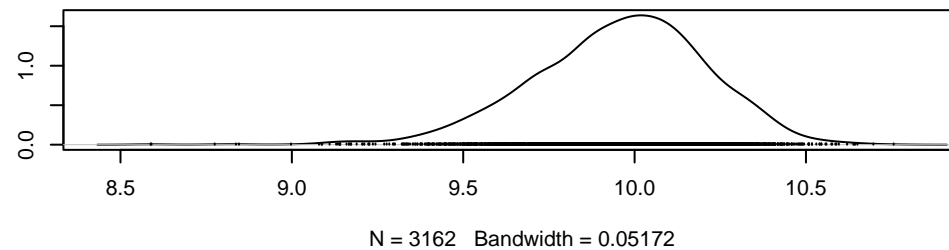
Density of cosiota



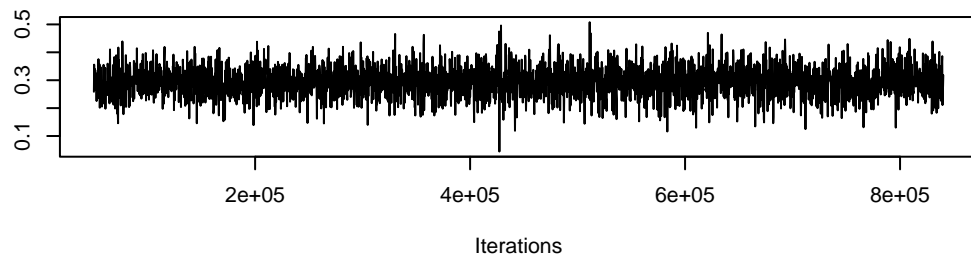
Trace of h0



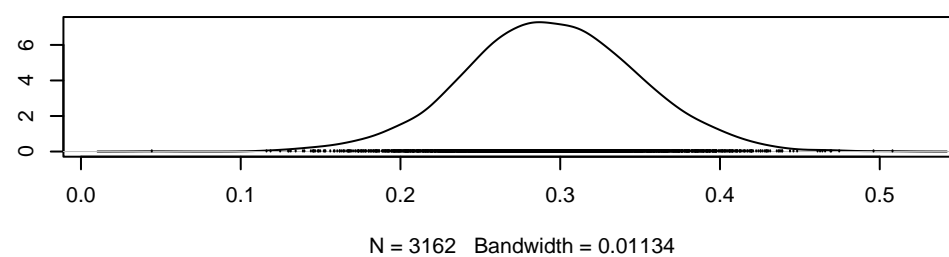
Density of h0



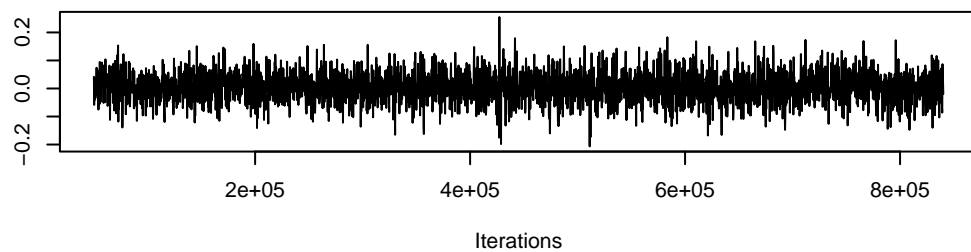
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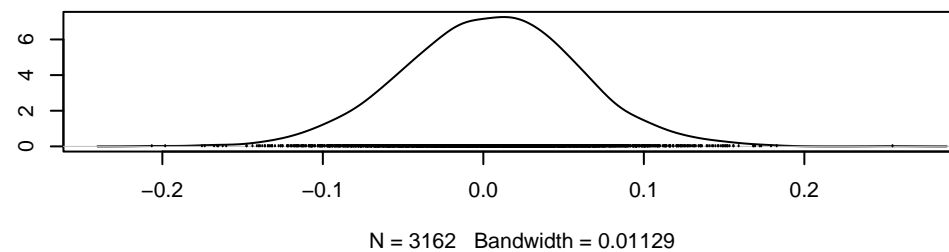
Density of psi



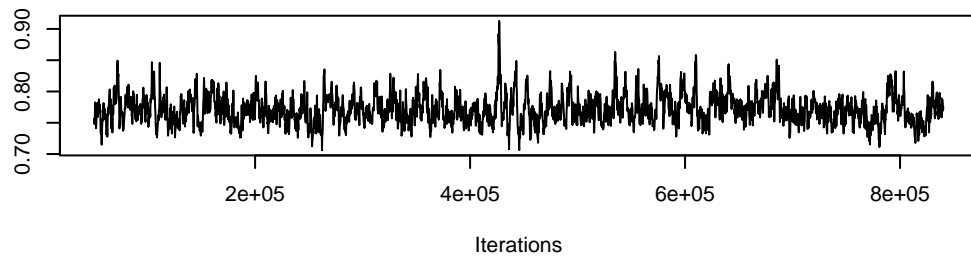
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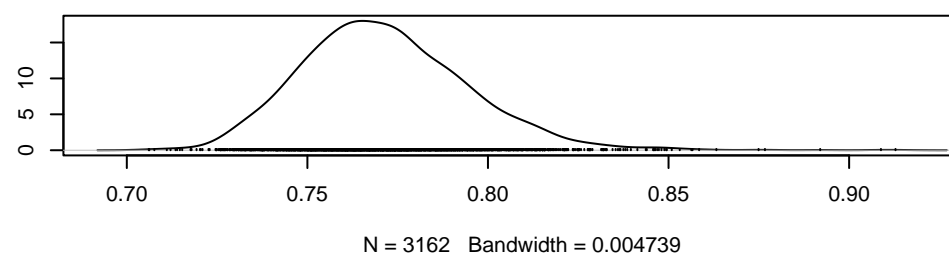
Density of phase



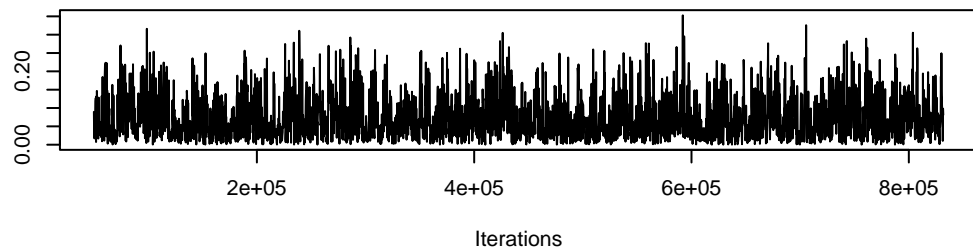
Trace of cosiota



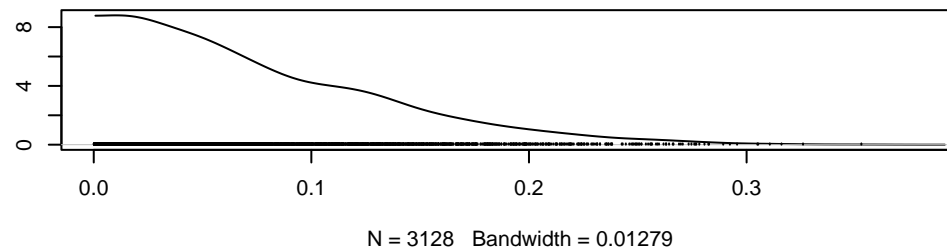
Density of cosiota



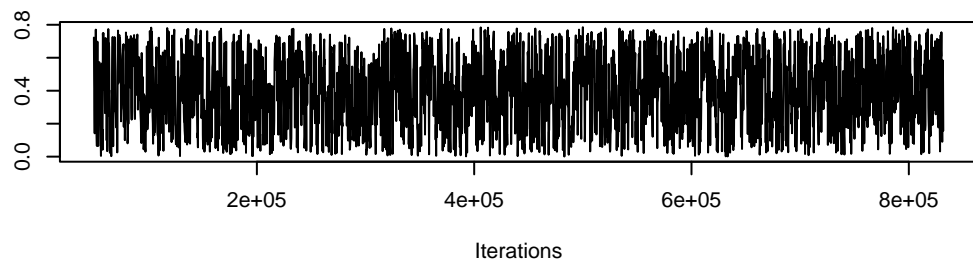
Trace of h_0



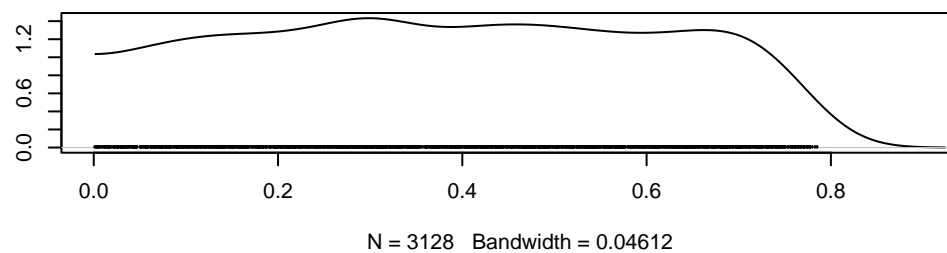
Density of h_0



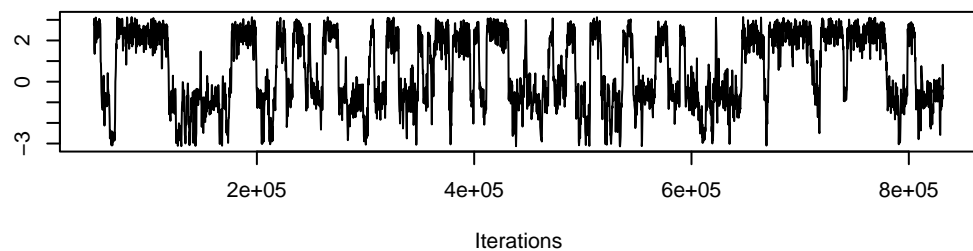
Trace of ψ



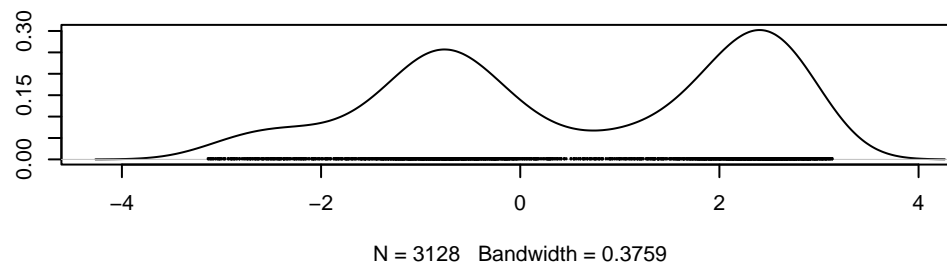
Density of ψ



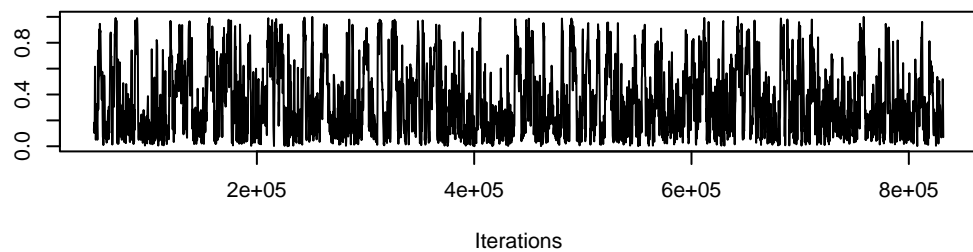
Trace of phase



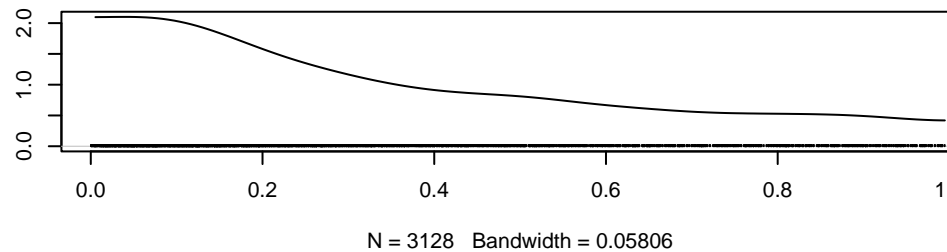
Density of phase



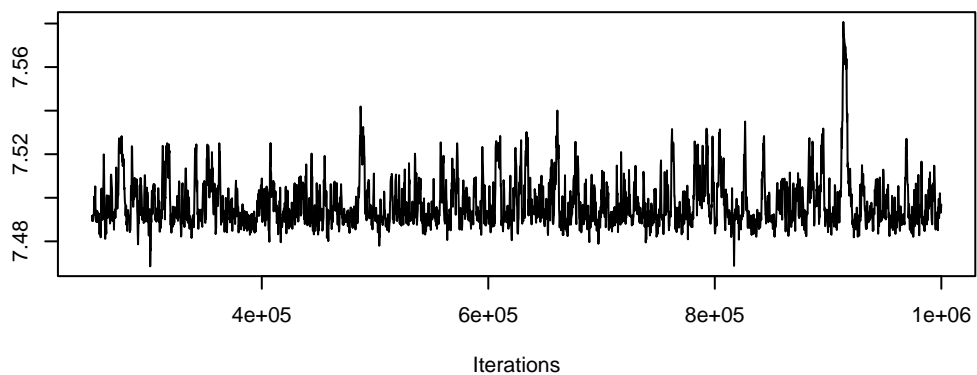
Trace of cosiota



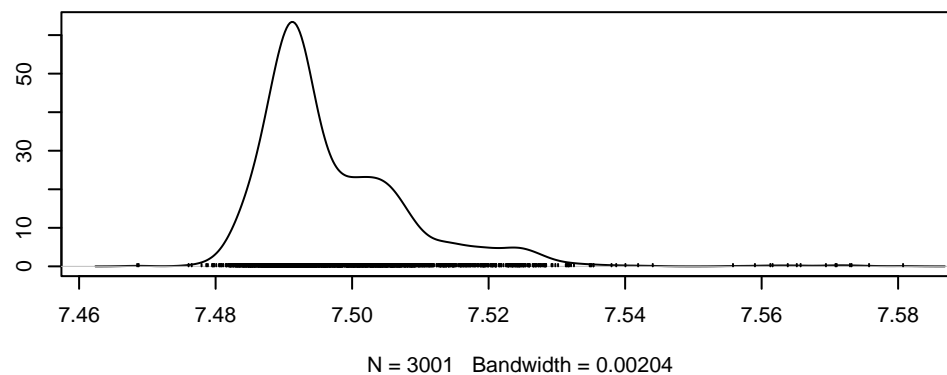
Density of cosiota



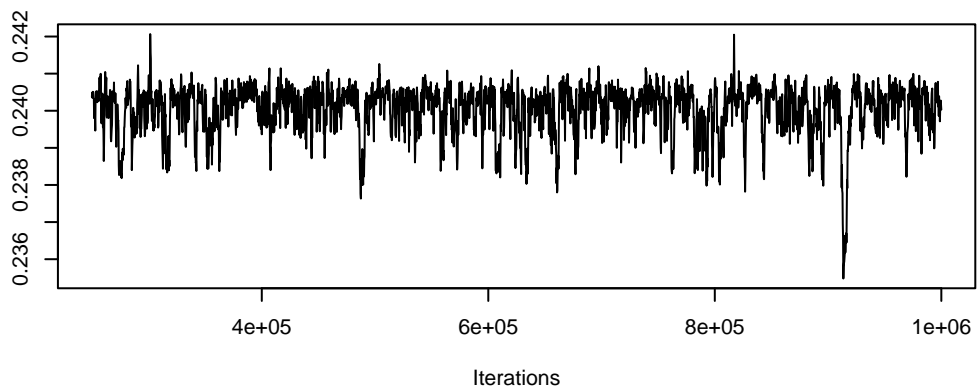
Trace of mtotal



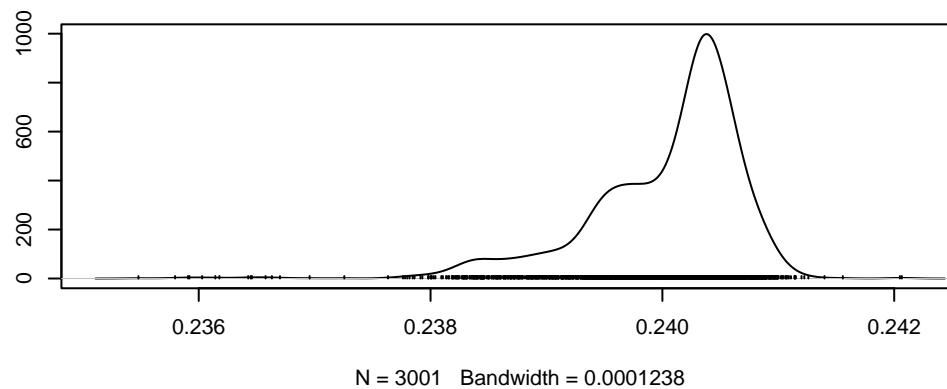
Density of mtotal



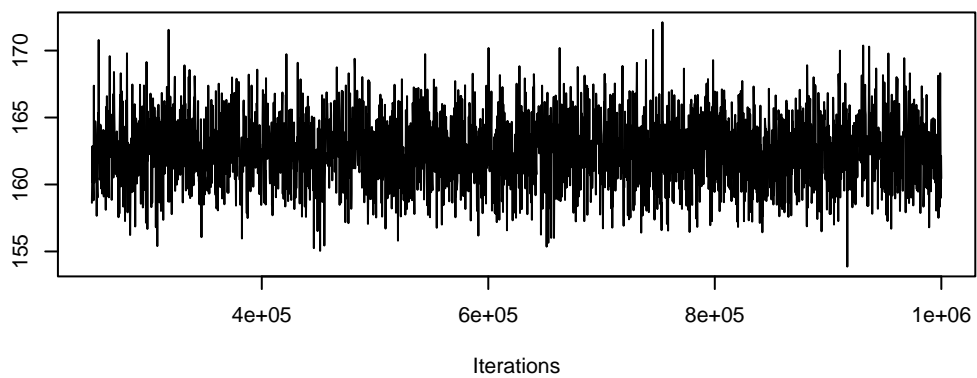
Trace of eta



Density of eta



Trace of dist



Density of dist

