

Manual: PEPICOBayes

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1 Introduction

PEPICOBayes is a program for reconstructing the probe spectrum from pump-probe and pump-only photoelectron-photoion coincidence (PEPICO) measurements applying Bayesian probability theory. The theory behind the program is described in Refs. [1, 2]. The PEPICOBayes program is written in C and executed in Matlab. We offer a simple example Matlab-file to demonstrate the usage of the PEPICOBayes program.

Ref. [2] extends the algorithm developed in Ref. [1] by including fluctuating laser intensities. We offer the input data for the mock data analysis and experimental data presented in Ref. [2] in the download folder. These data can be analysed using the PEPICOBayes program.

2 Abbreviations

The abbreviations in this manual are similar to the abbreviations used in reference [1].

Abbreviation	Description
j	Symbol that refers to $j \in \{1, 2\}$ 1... events due to pump-only laser pulse 2... events due to probe-only laser pulse
ρ	Symbol that refers to $\rho \in \{\alpha, \beta\}$ α ... events in the pump-only experiment β ... events in the pump-probe experiment
$n^{(\rho)}$	measured count rates $n^{(\rho)} = \{n_{\mu\nu}^{(\rho)}\}$, dimension: $[\mathcal{N}_\mu \times \mathcal{N}_\nu]$
$q^{(j)}$	Monte Carlo time series of the spectrum $q^{(j)} = \{q_{\mu\nu}^{(j)}\}$, dimension: $[\mathcal{N}_\mu \times \mathcal{N}_\nu \times N_{\text{sweep}}]$
N_{run}	Number of Monte Carlo steps between Monte Carlo measurements
N_{sweep}	Number of Monte Carlo measurements
\mathcal{N}_μ	Number of elements in dimension μ
\mathcal{N}_ν	Number of elements in dimension ν
\mathcal{N}_p	Number of measurements
N_{N_e, N_i}	Number of measurements where N_e electrons <i>and</i> N_i ions were detected.
π	Time series of experimental parameters: $\pi = \{\underline{\lambda}_1, \underline{\lambda}_2, \sigma_1, \sigma_2, \xi_i, \xi_e\}$, dimension: $[6 \times N_{\text{sweep}}]$
$\underline{\lambda}_j$	Mean number of events in channel j
σ_j	Fluctuations of mean number of events in channel j
ξ_i	Detection probability for ions
ξ_e	Detection probability for electrons

3 PEPICOBayes-Program

3.1 Function definition

The function returns four arguments and can be executed with six or eight arguments as input. Arguments seven and eight ($q_0^{(1)}$ and $q_0^{(2)}$) are start values for the spectra and optional. If no start spectra are passed the algorithm starts with flat spectra.

Input variables:

	Abbreviation	Variable name	Description
1	$n^{(\beta)}$	n_beta	measured pump-probe coincidence spectrum (dimension $[\mathcal{N}_\mu \times \mathcal{N}_\nu]$)
2	$n^{(\alpha)}$	n_alpha	measured pump-only coincidence spectrum (dimension $[\mathcal{N}_\mu \times \mathcal{N}_\nu]$)
3	N_{run}	Nrun	Number of Monte Carlo steps between measurements
4	N_{sweep}	Nsweep	Number of Monte Carlo measurements
5	parameter	parameter	Parameter vector: parameter := $\{\lambda_1, \lambda_2, \sigma_1, \sigma_2, \xi_i, \xi_e,$ $\sigma_{q^{(1)}}, \sigma_{q^{(2)}}, \sigma_{\lambda_1}, \sigma_{\lambda_2}, \sigma_{\sigma_1}, \sigma_{\sigma_2}, \sigma_{\xi_i}, \sigma_{\xi_e},$ $\mathcal{N}_p^{(\beta)}, N_{00}^{(\beta)}, N_{01}^{(\beta)}, N_{02}^{(\beta)}, N_{03}^{(\beta)}, N_{10}^{(\beta)}, N_{11}^{(\beta)}, N_{12}^{(\beta)}, N_{20}^{(\beta)}, N_{21}^{(\beta)}, N_{30}^{(\beta)},$ $\mathcal{N}_p^{(\alpha)}, N_{00}^{(\alpha)}, N_{01}^{(\alpha)}, N_{02}^{(\alpha)}, N_{03}^{(\alpha)}, N_{10}^{(\alpha)}, N_{11}^{(\alpha)}, N_{12}^{(\alpha)}, N_{20}^{(\alpha)}, N_{21}^{(\alpha)}, N_{30}^{(\alpha)}\}$
6	$q_0^{(1)}$		(<i>optional</i>) start values of $q^{(1)}$ (dimension $[\mathcal{N}_\mu \times \mathcal{N}_\nu]$)
7	$q_0^{(2)}$		(<i>optional</i>) start values of $q^{(2)}$ (dimension $[\mathcal{N}_\mu \times \mathcal{N}_\nu]$)

The size of a Monte Carlo step of variable $T \in \{q^{(1)}, q^{(2)}, \lambda_1, \lambda_2, \sigma_1, \sigma_2, \xi_i, \xi_e\}$ is Gaussian distributed with mean value zero and standard deviation σ_T . σ_T should be adjusted that p_{acc} is at the order of 50% in order to efficiently sample the probability distribution. Note that the standard deviation of for the step distribution for σ_j is $\min\{\sigma_{\sigma_j}, \sigma_j\}$. This is important if the fluctuations are very small and allows to visit phase space with a very small σ_j . For numerical stability the smallest value of σ_j is 10^{-8} .

Output variables:

	Abbreviation	Variable name	Description
1	$q^{(1)}$	q1_timeseries	Time series of $q^{(1)}$ (dimension $[\mathcal{N}_\mu \times \mathcal{N}_\nu \times N_{\text{sweep}}]$)
2	$q^{(2)}$	q2_timeseries	Time series of $q^{(2)}$ (dimension $[\mathcal{N}_\mu \times \mathcal{N}_\nu \times N_{\text{sweep}}]$)
3	π	Pi	Time series of π (dimension $[6 \times N_{\text{sweep}}]$)
4	p_{acc}	p_acc	Vector of acceptance rates with 6 entries: $p_{\text{acc}}(1)$: Acceptance rate of $\underline{\lambda}_1$ $p_{\text{acc}}(2)$: Acceptance rate of $\underline{\lambda}_2$ $p_{\text{acc}}(3)$: Acceptance rate of σ_1 $p_{\text{acc}}(4)$: Acceptance rate of σ_2 $p_{\text{acc}}(5)$: Acceptance rate of ξ_i $p_{\text{acc}}(6)$: Acceptance rate of ξ_e $p_{\text{acc}}(7)$: Acceptance rate of $q^{(1)}$ $p_{\text{acc}}(8)$: Acceptance rate of $q^{(2)}$

Note that the first entry in the time series is the start value of the corresponding variable.

3.2 Pseudocode of the algorithm

A Metropolis Hastings Monte Carlo algorithm is used to sample the probability distribution for $q^{(1)}$, $q^{(2)}$ and π .

```

PEPICOBayes( $n^{(\beta)}$ ,  $n^{(\alpha)}$ ,  $N_{\text{run}}$ ,  $N_{\text{sweep}}$ , parameter,  $q_0^{(1)}$ ,  $q_0^{(2)}$  )
  read all input variables
  for  $k < N_{\text{sweep}}$ 
    for  $i < N_{\text{run}}$ 
      suggest Monte Carlo step in  $\underline{\lambda}_1$ ,  $\underline{\lambda}_2$ ,  $\sigma_1$ ,  $\sigma_2$ ,  $\xi_e$ ,  $\xi_i$ ,  $q^{(1)}$  or  $q^{(2)}$ 
      compute Monte Carlo step probability
      if update is accepted
        overwrite corresponding variable with the new value
      else
        do nothing
      end
    end
    copy  $\underline{\lambda}_1$ ,  $\underline{\lambda}_2$ ,  $\sigma_1$ ,  $\sigma_2$ ,  $\xi_e$ ,  $\xi_i$ ,  $q^{(1)}$  and  $q^{(2)}$  into output variables at Monte Carlo time  $k$ 
  end
end

```

3.3 Start value for π

The start values of π can for example be approximated by the formula given by Stert et. al. [3]. In summary this means solving the following equation for λ

$$\frac{N_e N_i}{\lambda \mathcal{N}_p} \left(1 + \lambda \left(1 - \frac{N_e}{\lambda \mathcal{N}_p} \right) \left(1 - \frac{N_i}{\lambda \mathcal{N}_p} \right) \right) e^{-\lambda \left(1 - \left(1 - \frac{N_e}{\lambda \mathcal{N}_p} \right) \left(1 - \frac{N_i}{\lambda \mathcal{N}_p} \right) \right)} \approx N_{11} \quad (1)$$

and insert the resulting λ into

$$\xi_e \approx \frac{N_e}{\lambda \mathcal{N}_p} \quad (2)$$

$$\xi_i \approx \frac{N_i}{\lambda \mathcal{N}_p} \quad (3)$$

N_{11} is the total number of measured coincidences and N_e and N_i are the total number of detected electrons and ions, respectively.

4 Results of example.m

In the example two ion types ($\mathcal{N}_\mu = 2$) and seven different electron flight times ($\mathcal{N}_\nu = 2$) were detected.

The result of the converged Monte Carlo computation for the example data should look like figure 3. Converged time series are shown in figure 1 and 2 for demonstration purposes. Check for correlations by evaluating the autocorrelation function or with techniques like jackknife or binning.

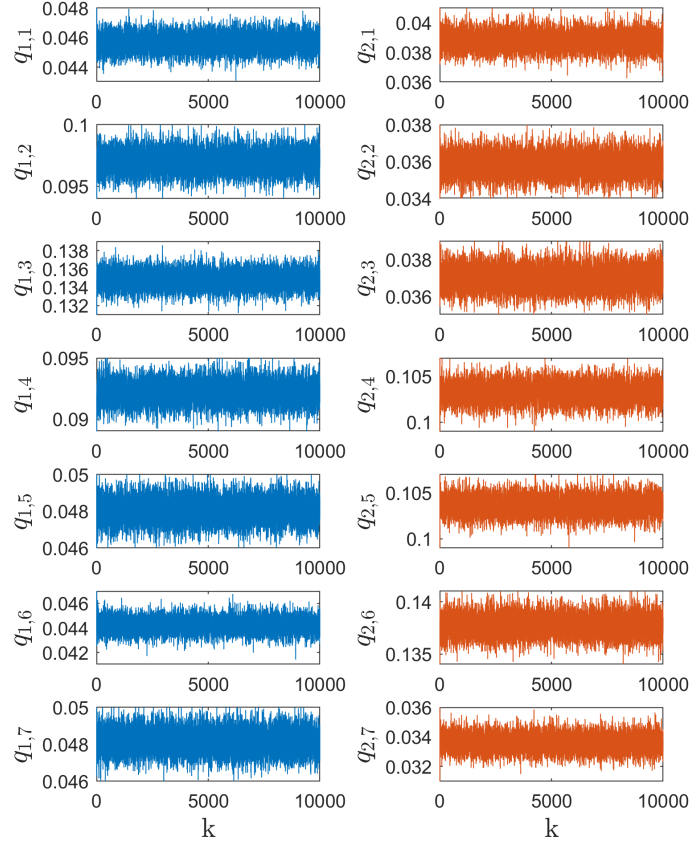


Figure 1: Time series of $q^{(2)}$.

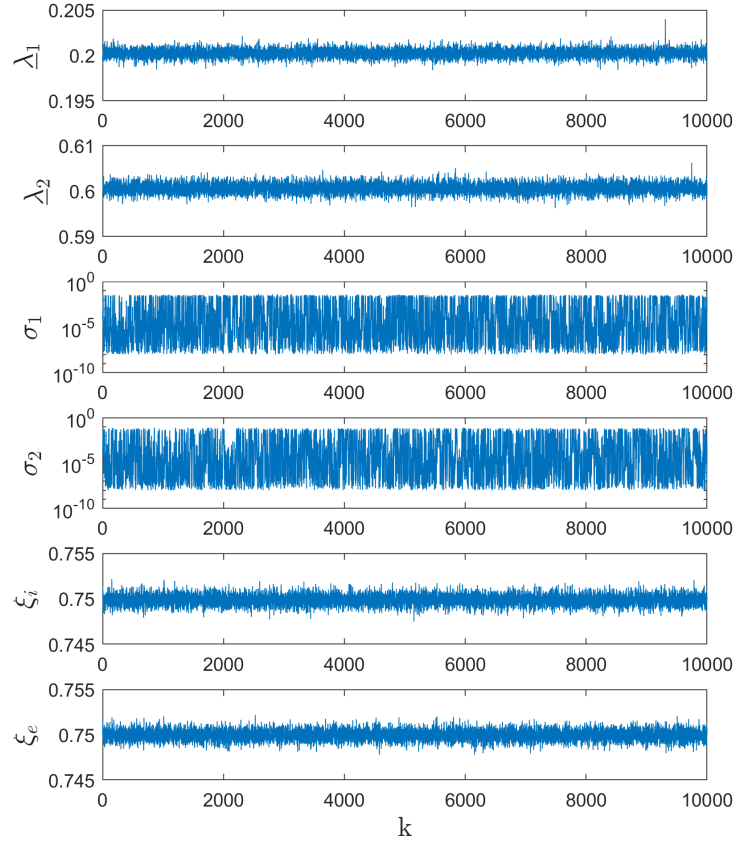


Figure 2: Time series of π . The real values are $\lambda_1 = 0.2$, $\lambda_2 = 0.6$, $\sigma_1 = 0$, $\sigma_2 = 0$, $\xi_i = 0.75$, $\xi_e = 0.75$. The algorithm can only evaluate that σ_j is very small but cant pin it down because in this regime it has no influence.

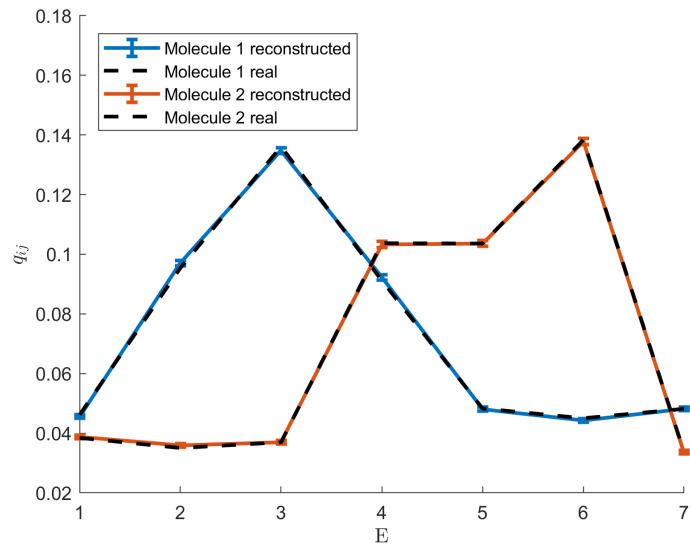


Figure 3: Reconstructed spectrum $q^{(2)}$.

5 Troubleshooting

5.1 Reconstruction is not converged

If the result is not converged the last entry of the time series of $q^{(1)}$, $q^{(2)}$ and π can be used as start parameter for a new computation. Extend the Monte Carlo chain until the desired results are converged.

5.2 C file in MATLAB

In the downloaded folder are pre-compiled c files (tested with *Windows10* and *Debian 9.6* - MATLAB version 2017b, 2018b). If the pre-compiled files dont work see https://de.mathworks.com/help/matlab/matlab_external/what-you-need-to-build-mex-files.html, follow the instructions and compile the c file on your computer.

References

- [1] M. Rumetshofer, P. Heim, B. Thaler, W. E. Ernst, M. Koch, and W. von der Linden. Analysis of femtosecond pump-probe photoelectron-photoion coincidence measurements applying bayesian probability theory. *Phys. Rev. A*, 97:062503, Jun 2018.
- [2] P. Heim, M. Rumetshofer, S. Ranftl, B. Thaler, W.E. Ernst, M. Koch, and W. von der Linden. Bayesian analysis of femtosecond pump-probe photoelectron-photoion coincidence spectra with fluctuating laser intensities. *Manuscript in preparation*.
- [3] V. Stert, W. Radloff, C.P. Schulz, and I.V. Hertel. Ultrafast photoelectron spectroscopy: Femtosecond pump-probe coincidence detection of ammonia cluster ions and electrons. *The European Physical Journal D-Atomic, Molecular, Optical and Plasma Physics*, 5(1):97–106, 1999.