## Transition Matrix Methods

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## Outline

- Review of PMD
- Multicanonical Algorithm
- Transition Matrix Methods
- Numerical results
- Conclusions


## Polarized Waveguide Modes

- For single mode waveguides with sufficient symmetry, light propagates in a superposition of two degenerate modes.
- In general the modal group velocities differ, yielding signal distortion in optical fibers (polarization mode dispersion).


## PMD Emulator Types



## B

## PMD Emulators

- We describe a general communication system by a set of system parameters $\vec{\alpha}$.
- For PMD emulators, these are e.g. the polarization rotation angles associated with polarization scramblers separated by concatenated birefringent sections or the optical path lengths of successive mutually rotated birefringent segments.


## System Simulation

- We generate random values of the local system parameters (rotation angles).

$$
\vec{\alpha}=\left(\theta_{1}, \phi_{1}, \theta_{2}, \phi_{2}, \ldots, \theta_{N}, \phi_{N}\right)
$$

- For each realization, we calculate one or more global system variables $\vec{O}$
(observables). In our case these are the PMD vectors $\tau, \tau_{\omega} \ldots$


## Control Variables

- Random control variables

$$
\vec{\alpha}=\left(\alpha_{1}, \alpha_{2} \ldots, \alpha_{N}\right)
$$

- Objective: Determine the probability distribution function (pdf) $p(\vec{O})$ of a set of observables

$$
\vec{O}=\vec{O}(\vec{\alpha})
$$

- Generate $N$ sets of control variables $\vec{\alpha}$ according to the physical distribution $p(\vec{\alpha})$
- Compute or measure $\vec{O}$ for each set $\vec{\alpha}$


## Monte Carlo Sampling

- If the function $I\left(\vec{O}_{k}, \vec{\alpha}^{(i)}\right)$ is one inside the $k$ :th histogram bin, then after $N$ realizations,

$$
p\left(\vec{O}_{k}\right) \approx \frac{1}{N} \sum_{i=1}^{N} I\left(\vec{O}_{k}, \vec{\alpha}^{(i)}\right)
$$

- Clearly few events arise in regions of small probability. Therefore, many samples are required to generate "worst case" events with small PMD statistics.


## Markov Chains

- To insure that lower probability states are sampled more often, Markov chains are employed.
- Additing a small perturbation to the current srate yields a proposed transition
- A rule governs the acceptance of this transition.


## Multicanonical Algorithm

- The transition rule should ensure:
- Equal sampling of equal O regions
- Rapid escape from local minima
- (Detailed balance)
- These require that the acceptance is a particular function of the ratio of the pdf of the initial and final states.
- However, the pdf is unknown and must be determined iteratively.


## Multicanonical Procedure

- A Markov chain leading to equal sampling of the system variable space, $O$.
- Starting from $p_{0}=1$, accept all transitions that decrease $p_{0}$ and those that increase $p_{0}$ with a probability $p_{0}\left(\vec{O}^{(\text {curr) }}\right) / p_{0}\left(\vec{O}^{(n e w)}\right)$ This gives the acceptance probability

$$
\min \left\{1, p_{0}\left(\vec{O}^{(\text {curr })}\right) / p_{0}\left(\vec{O}^{\text {new })}\right)\right\}
$$

- This yields a (Monte-Carlo) distribution $H_{1}$.
- Iterate with $p_{n}=p_{n-1} H_{n-1} / c_{n}$


## Result of Iterations

- In the tail regions regions, $p$ is initially 1 , yielding Monte-Carlo statistics and the number of events falls off as the PDF.
- As the iterations proceed, the states increasingly sample these low-probability regions.


## MC Enhancements

- Raise the intermediate result $p_{n}(E)$ to a power in the transition rule.
- Introduce a bias function $p_{n}(E) \rightarrow c_{n} p_{n}(E) H(E)$ into the transition rule and combine the results in overlapping regions.
- Interpolate the histogram in the transition rule.
- Employ different probability distributions


## $1^{\text {st }}$ Order PMD Results



Modified

-15 section PMD emulator
-Section DGD (| $\left.\vec{\tau}_{t} \mid=1 p s\right)$

- $\vec{\tau}_{l}$ rotates randomly on Poincare sphere
-500,000 samples/iteration
-8 iterations


## Joint PMD Calculation

- We employ a 15 section PMD emulator.
- 200,000 samples/iteration
- Number of histogram bins: $100 \times 100$
- We normalize the first order PMD to <PMD> and the second order PMD to <PMD>2.


## Joint pdf Results



## Experimental Results

- Determined the DGD distribution of a microheater-based PMD emulator.
-     * Measured the joint pdf of the first and second order PMD of a 8 section PMD emulator with General Photonics polarization controllers.
- Recorded the distribution of bit error rates in a recirculating fiber loop.


## PMD Emulator



Fig. 1. Experimental setup


Fig. 2. The pdf of the DGD obtained with a 20,000 sample experimental Monte Carlo measurement ( + ), a multicanonical measurement with ten iterations of 2,000 samples ( 0 ) and a numerical multicanonical simulation employing twenty 50,000 sample iterations (solid line)

## Biasing Method



Fig. 4. The $\mathcal{R}_{4} \equiv 24<\tau<32 p s$ subregion bias function


Fig. 5. The histograms obtained after each biased multicanonical iteration

## Full pdf



Fig. 6. The pdf obtained from the biased multicanonical method generated by combining results for all subregions $\mathcal{R}_{1} \ldots \mathcal{R}_{5}$.

## Transition Matrix Method

- For every accepted or rejected transition from the $n$ :th to the $m$ :th histogram bin in a biased calculation:
» Increment the corresponding element, $M_{n m}$, of an unbiased but unnormalized transition matrix, $M$, by unity.
» At the end normalize the rows of $T_{n m}$ to unity, yielding the transition matrix $T$.


## Rapid pdf updating

- By detailed balance, the pdf can be obtained from the recursion relation

$$
p\left(E_{i+1}\right)=\frac{T_{i+1, i}}{T_{i, i+1}} p\left(E_{i}\right)
$$

- This enables updates after every few steps.
- Alternatively transition only to final states visited less than the initial state.


## Detailed Balance Violation

- The preceding method violates detailed balance
- More system realizations enter a histogram bin from the high probability side than the low probability side.
- If transitions out of the bin are discarded until the physical pdf is established this problem can be resolved.


## Sampling Frequency



Fig. 2. Total number of times each histogram bin is visited for the standard multicanonical procedure ( $\triangle$ ), method $1(\bigcirc)$, and method 2 (dashed-dotted curve).

## Method Comparison



Fig. 1. Ratio between the numerical and analytic pdfs for the standard multicanonical procedure ( $\Delta$ ), our modified transition matrix procedure with a multicanonical acceptance rule (method 1,0 ), an acceptance rule that rejects transitions to more visited histogram bins (method 2, dashed-dotted curve), and a procedure that restricts transitions out of a recently visited bin (method 3, crosses) as functions of the normalized DGD for a $N_{\text {sec }}=10$ segment fiber emulator.


Fig. 3. Variation of the error, Eq. (2), weighted by the histogram bin probability as a function of the average DGD change over one Markov step for the standard multicanonical method ( $\Delta$ ), method $1(\mathrm{O})$, method 2 (dashed-dotted curve), and method 3 (crosses).

## Outage Time Formulation

- Initial state: $S_{n}=P\left(E_{n}\right)$ for $n$ in the nonoutage region, zero elsewhere
- Repeatedly multiply by the transition matrix.
- At each step, sum and then set to zero the histogram values of non-outage states.
- Alternatively, employ the outage region submatrix eigenfunctions and eigenvalues.


## Outage Time Calculation

- Outage conditions: (1) DGD > 2<DGD> and (2) DGD > 3<DGD>.
- Circles $-10^{9}$ step unbiased Markov chain.
- Solid line - Biased TM calculation with three 5,000K sample MC iterations
- Dashed line - Multiplication with the unbiased TM from twenty 200K MC iterations or submatrix eigenvalue method.


## Outage Time Results



## Fading Channels

- Gain of Rayleigh fading channel

$$
\begin{gathered}
g=\mu_{1}(t)+\mu_{2}(t) \\
\mu_{m}(t)=\sum_{n=1}^{N_{m}} \cos \left(2 \pi f_{\max } \sin \left(\frac{\pi(2 n-1)}{4 N_{m}}\right) t+\theta_{m n}\right)
\end{gathered}
$$

- The phases $\theta_{m}$ are the control variables.
- The durations of two successive fading events are employed as the observables.
- This yields relevant 1-dimensional pdf:s


## PDF of Fading Events



Fig. 1. The base 10 logarithm of the pdf of the number of fading events for $f_{\max }=10 \mathrm{~Hz}$ within $T_{F}=1 \mathrm{sec}$ for outage threshold levels of $R=5$ dB (dashed-dotted line), 10 dB (dashed line) and 15 dB (solid line) below the mean channel gain as evaluated with two $10^{5}$-sample iterations of the multicanonical method as well as for our modified transition matrix procedure $(+)$ with a $2 \times 10^{5}$ samples and a direct evaluation of Eq. (2) (0).

## PDF of Fade Duration



Fig. 2. The fade duration distribution for three $2 \times 10^{6}$-sample iterations of the multicanonical method (solid lines) with $N=100$ bins for outage levels of $R=5 \mathrm{~dB}$ and 15 dB below the mean channel gain, respectively, and for $6 \times 10^{6}$ samples of the the modified transition matrix procedure $(+)$ and Eq. (2) (o). The dashed line indicates the sampling bias introduced by the one-dimensional multicanonical calculation for $R=5 \mathrm{~dB}$

## Conclusions

- Transition matrix methods enable rapid prediction of the dynamic behavior of general communication systems.
- However, standard procedures are more accurate since all correlations between states are present.
- The technique must be carefully applied when long-time correlations exist.

