

# ALGORITHM FOR SAMPLING MAXIMAL COUPLINGS OF MARKOV CHAINS

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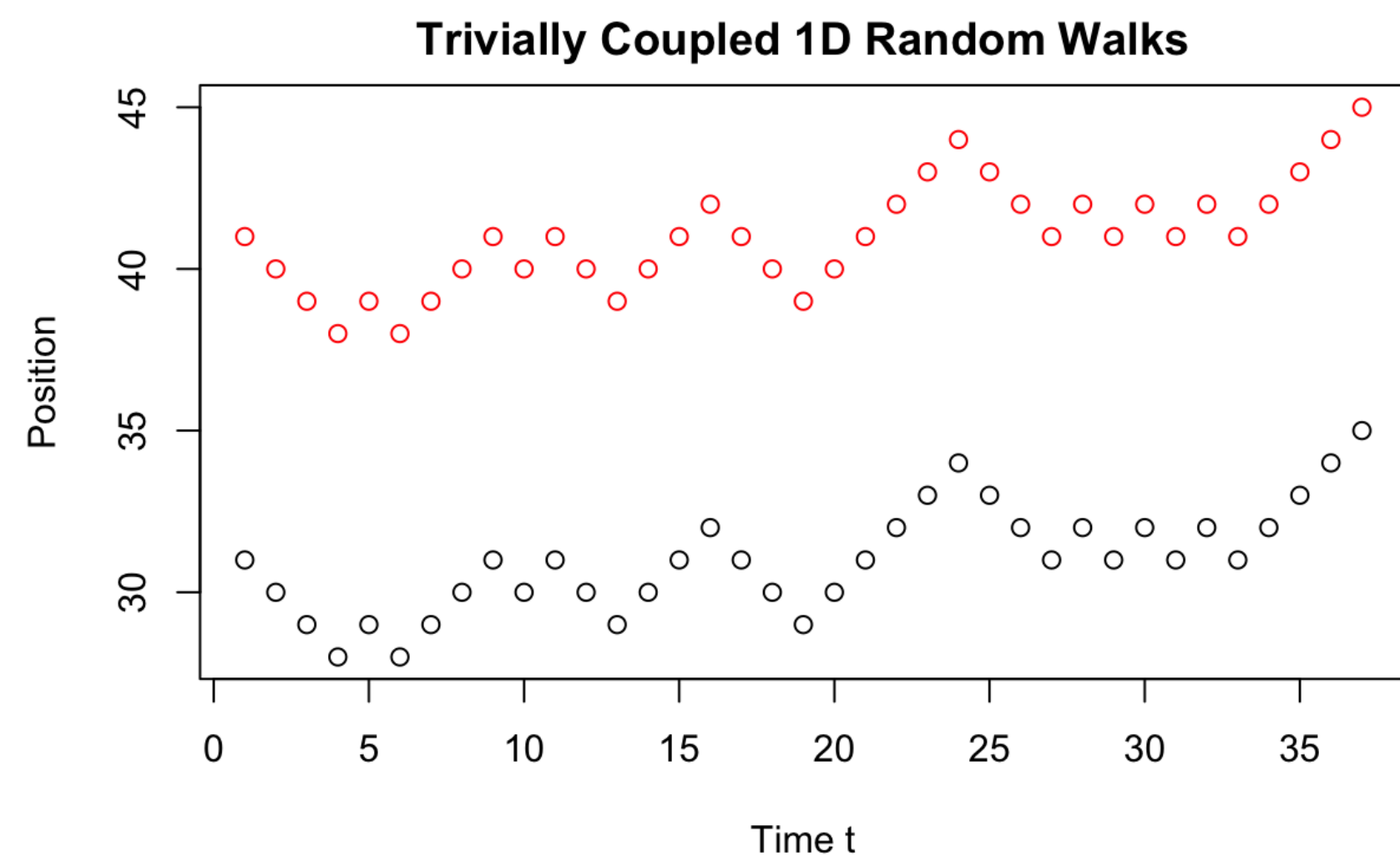
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## Defining a Maximal Coupling

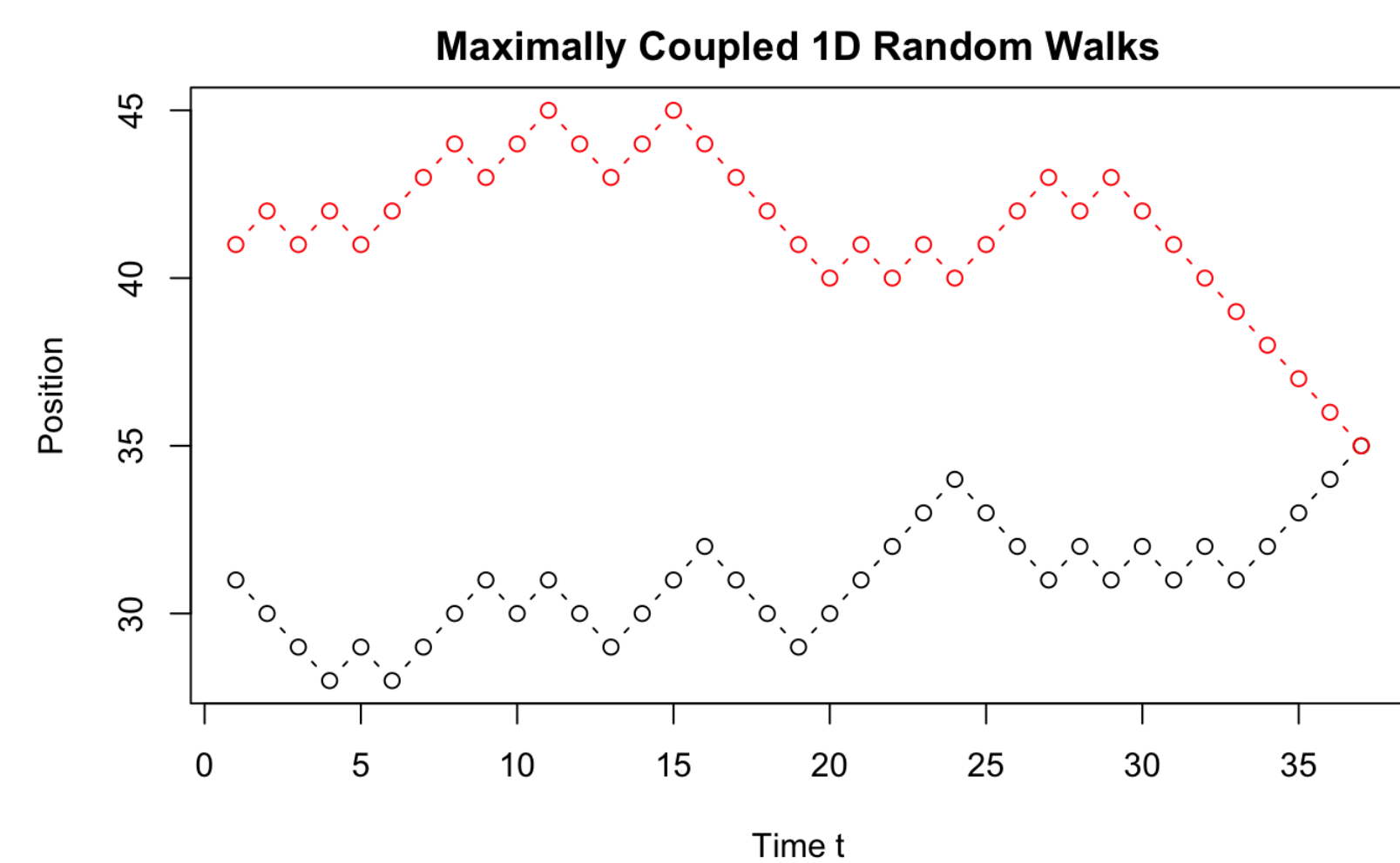
Given two Markov Chains  $X_t, Y_t$  with initial distributions  $\mu, \nu$  and shared transition function  $P$ , a *coupling* introduces some kind of dependence between  $X_t, Y_t$



Here  $X_t$  moves randomly, but  $Y_t$  depends on  $X_t$  and stays parallel



Here  $X_t$  and  $Y_t$  are independent; the path of  $X_t$  does not affect the path of  $Y_t$ .



$X_t$  and  $Y_t$  are **maximally coupled**; the dependence maximizes the rate at which they reach the same state.

## Constructing a Maximal Coupling

Let  $T$  be the first time the paths of  $X_t, Y_t$  meet (the coupling time)

**Theorem 1.** If  $X_t, Y_t$  are maximally coupled, we can recursively compute:

$$\begin{aligned} m_t(x) &= \mathbb{P}[X_T = Y_T = x, T = t] & \alpha_{\nu,t}(y) &= \mathbb{P}[Y_t = y, T > t] \\ \alpha_{\mu,t}(x) &= \mathbb{P}[X_t = x, T > t] & C_{\nu,t}(y) &= \mathbb{P}[T = t | Y_t = y, T \geq t] \\ C_{\mu,t}(x) &= \mathbb{P}[T = t | X_t = x, T \geq t] \end{aligned}$$

Probability that maximally coupled  $X$  takes on path  $\bar{x} = (x_0, \dots, x_t)$  and  $T = t$ :

$$\mathbb{P}(X_{[0,t]} = \bar{x}, T = t) = \mu(x_0) C_{\mu,t}(x_t) \prod_{s=0}^{t-1} (1 - C_{\mu,s}(x_s)) P(x_s, x_{s+1}) \quad (1)$$

## Time Reversal

Note the substitutions:

$$q_t(y, x) := \frac{\alpha_{\nu,t-1}(x) P(x, y) (1 - C_{\nu,t}(y))}{\alpha_t(y)} \quad (2)$$

$$Q_t(y, x) := \frac{\alpha_{\nu,t-1}(x) P(x, y) C_{\nu,t}(y)}{m_t(y)} \quad (3)$$

This reverses the time in our original path decomposition formula:

$$\mathbb{P}(Y_{[0,t]} = \bar{y} | T = t, X_t = Y_t = y_t) = Q_t(y_t, y_{t-1}) \prod_{n=1}^{t-1} q_n(y_n, y_{n-1}) \quad (4)$$

## Algorithm

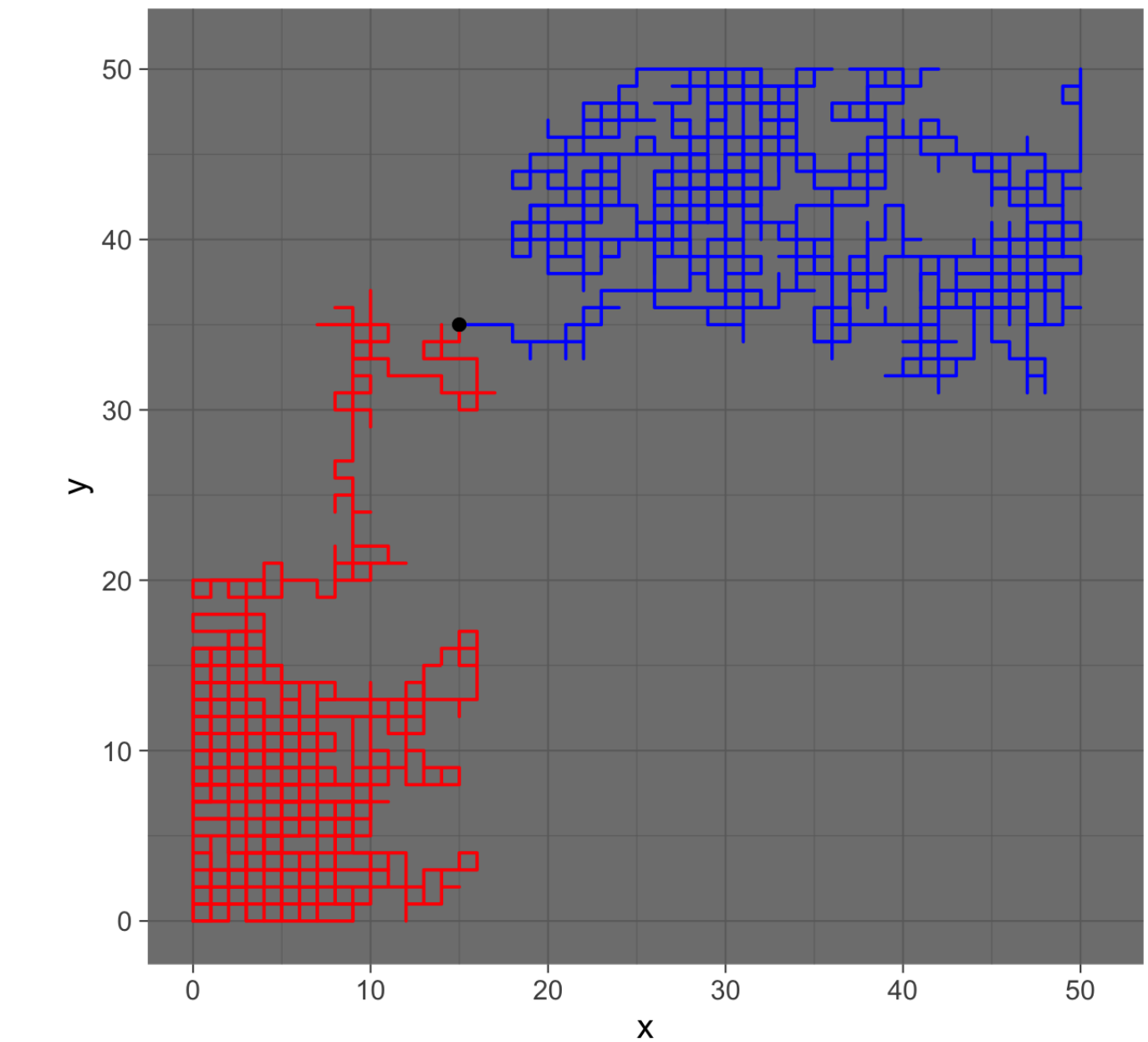
We first sample the  $X$  path, and then iterate backward to sample the  $Y$  path:

Sample a Maximal Coupling

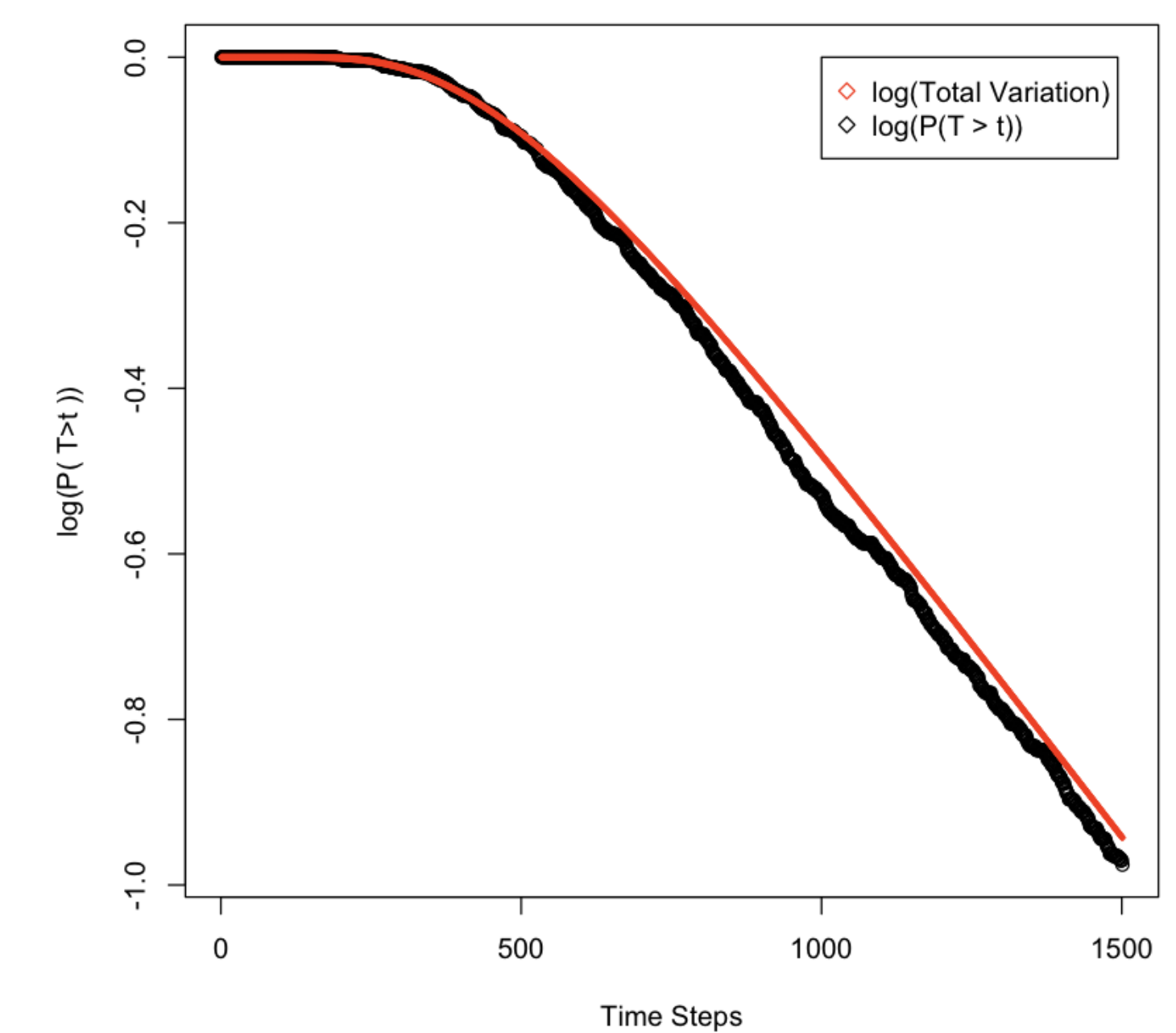
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0: procedure SAMPLE( $P, \mu, \nu$ )
0:    $t = 0$ 
0:    $X_0 \sim \mu$ 
0:   while Uniform[0, 1] <  $C_{\mu,t}(X_t)$  do
0:      $X_{t+1} \sim P(X_t, \cdot)$ 
0:      $t = t + 1$ 
0:   end while
0:    $Y_t = X_t$ 
0:    $Y_{t-1} \sim Q_t(Y_t, \cdot)$ 
0:    $t = t - 1$ 
0:   while  $t > 0$  do
0:      $Y_{t-1} \sim q_t(Y_t, \cdot)$ 
0:      $t = t - 1$ 
0:   end while
```

## Simulating a Maximal Coupling

Maximally Coupled Walks on a  $50 \times 50$  Grid



Theoretical TV vs. Empirical TV of Maximally Coupled Walks on a  $50 \times 50$  Grid



## References

### References

- [1] Levin, David A., Peres, Yuval and Wilmer, Elizabeth L.. Markov chains and mixing times. : American Mathematical Society, 2006.
- [2] Pitman, J.W. On coupling of Markov chains. Z. Wahrscheinlichkeitstheorie verw Gebiete 35, 315-322 (1976).