# Outline

### Markov processes

- 2 Classification of states
- 3 Classification of chains
- Stationary distributions and the limit theorem
   Stationary distributions
   Limit theorems

### 5 Reversibility

- 6 Chains with finitely many states
- 7 Branching processes revisited

## Reversed chain

### Theorem 42.

### Let

- X irreducible non-null persistent chain
- Transition for X is P, invariant measure is  $\pi$
- Hypothesis:  $X_n \sim \pi$  for all  $n \quad (\Leftarrow \times_0 \lor \intercal)$
- Set  $Y_n = X_{N-n}$  for  $0 \le n \le N$  (X with time run backwards)

Then

- Y is a Markov chain
- The transition for Y is

$$\mathbf{P}(Y_{n+1}=j|Y_n=i)=\frac{\pi_j}{\pi_i}p_{ji}$$

Recall  $Y_n = X_{N-n}$  for n = 0, ..., NQuantity to evoluate: Ynn = XN-(nn) = XN-n-1 Qn = P(Ynri=inri | Yn=in,..., Ys=is) = P(Ynn = inn, Yn=in, ..., Ys=is)  $P(Y_{a}=i_{n},..,Y_{o}=i_{o})$ = P(XN-n-1 = inri, XN-n= in,.., XN = is) P(XN-n=in, ..., XN=is) Parkor Frank. = Tlinn Pinn, in Pi, is Tin Pin, in-1 Pin, is Tim Pinniin 5 Ticn.

summary we have found P(Yn+1 = in+1 Yn=in,..., Ys=is) = Tlinn Pium, in Tlin Nov, with the same type of conputation P(Ynn=inn | Yn=in) = ITinn Pinn in Thus Y is a Markov chain with Mansihin No Poi

## Proof of Theorem 42

Computing conditional probabilities: We have

$$\mathbf{P}(Y_{n+1} = i_{n+1} | Y_n = i_n, \dots, Y_0 = i_0) \\
= \frac{\mathbf{P}(Y_0 = i_0, Y_1 = i_1, \dots, Y_{n+1} = i_{n+1})}{\mathbf{P}(Y_0 = i_0, Y_1 = i_1, \dots, Y_n = i_n)} \\
= \frac{\mathbf{P}(X_{N-n-1} = i_{n+1}, X_{N-n} = i_n, \dots, X_N = i_0)}{\mathbf{P}(X_{N-n} = i_n, \dots, X_N = i_0)} \\
= \frac{\pi_{i_{n+1}} p_{i_{n+1}i_n} p_{i_{n}i_{n-1}} \cdots p_{i_1i_0}}{\pi_{i_n} p_{i_ni_{n-1}} \cdots p_{i_1i_0}} \\
= \frac{\pi_{i_{n+1}} p_{i_{n+1}i_n}}{\pi_{i_n}} \\
= \mathbf{P}(Y_{n+1} = i_{n+1} | Y_n = i_n)$$

This gives the Markov property and the transition

Samy T. (Purdue)

= nar

< □ > < ---->

## **Reversed** chain

### Definition 43.

#### Let

- X irreducible non-null persistent chain
- Transition for X is P, invariant measure is  $\pi$

• Hypothesis: 
$$X_n \sim \pi$$
 for all  $n$ 

• Set 
$$Y_n = X_{N-n}$$
 for  $0 \le n \le N$   
then manifold  $T_i$  for  $0 \le n \le N$ 

•  $\dot{X}$  is said to be reversible if Y has transition P

# Vocabulary

#### Detailed balance: Let

- P transition matrix
- $\lambda$  distribution

Then  $P, \lambda$  are in detailed balance if

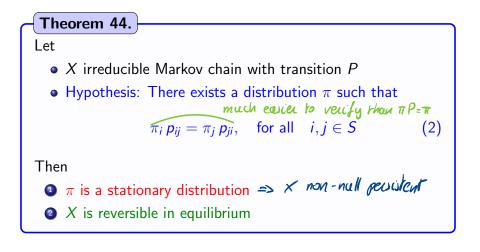
$$\lambda_i \, p_{ij} = \lambda_j \, p_{ji}, \quad ext{for all} \quad i,j \in S$$

### Reversible in equilibrium: If X is such that

•  $P, \pi$  are in detailed balance,

then X is said to be reversible in equilibrium

### Invariant measure and reversibility



Claim If Tipiz = Tip pic, then T is an invariant measure Most (TTP]; = Z Ti Pi; detailed bulance = Z T; P;i =  $\pi_i \sum_{i \in N} P_i$  = 1 (Pstechastic matrix) = Îl; Conclusion (TP); = Ti; => TT invariant

Proof of Theorem 44

Computation of  $\pi P$ : We have

$$(\pi P)_{j} = \sum_{i \in S} \pi_{i} p_{ij}$$
$$= \sum_{i \in S} \pi_{j} p_{ji}$$
$$= \pi_{j} \sum_{i \in S} p_{ji}$$
$$= \pi_{j}$$

### Conclusion:

- **1**  $\pi$  is invariant
- 2 X is reversible in equilibrium from (2)

э

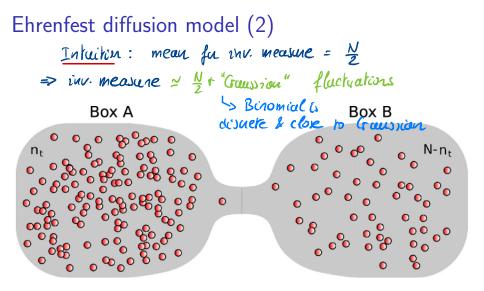
# Ehrenfest diffusion model (1)

Model: We consider

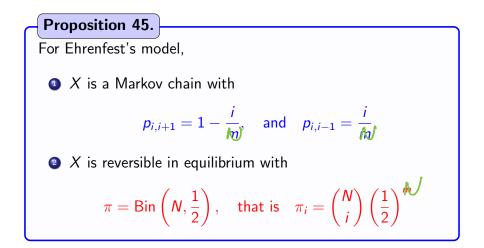
- Two boxes A and B
- Total of N gas molecules in  $A \cup B$
- At time *n*, one molecule is picked from the *N* molecules
- This molecule changes box

Process: We set

 $X_n \equiv \#$  molecules in Box A at time n



# Ehrenfest diffusion model (2)



Samv	T (	(Purdue)	
Jamy		(i uiuuc)	