



Efficiently Pricing European-Asian Options: Ultimate Implementation and Analysis of the AMO Algorithm

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Summary of This Talk

- **option**: typical financial derivative
- pricing **European-Asian option** on **binomial model**
 - - - difficult to compute accurately
 - approximation
- Aingworth, Motwani & Oldham (SODA00)
 - time: $O(kn^2)$, absolute error $O(nX/k)$
- Our Algorithm:
 - time: $O(kn^2)$, absolute error **$O(X/k)$**

n, X : problem parameters, k : time-error tradeoff param.

Option



- **option**: right to sell (or buy) some financial asset (e.g., stock) at some point in the future (**expiration date**) for a specified price (**strike price**)
- gain more benefit by investment
- hedge risk from the fluctuation of stock price

Payoff of Option

Example: option to buy a stock of Google Inc.
at the year-end at \$200

- stock price goes up to \$220 at the year-end
exercise option to buy the stock at \$200
sell it for \$220 gain \$20 (payoff)
- stock price goes down to \$170
do not exercise option payoff = \$0



Payoff of **European Option**:

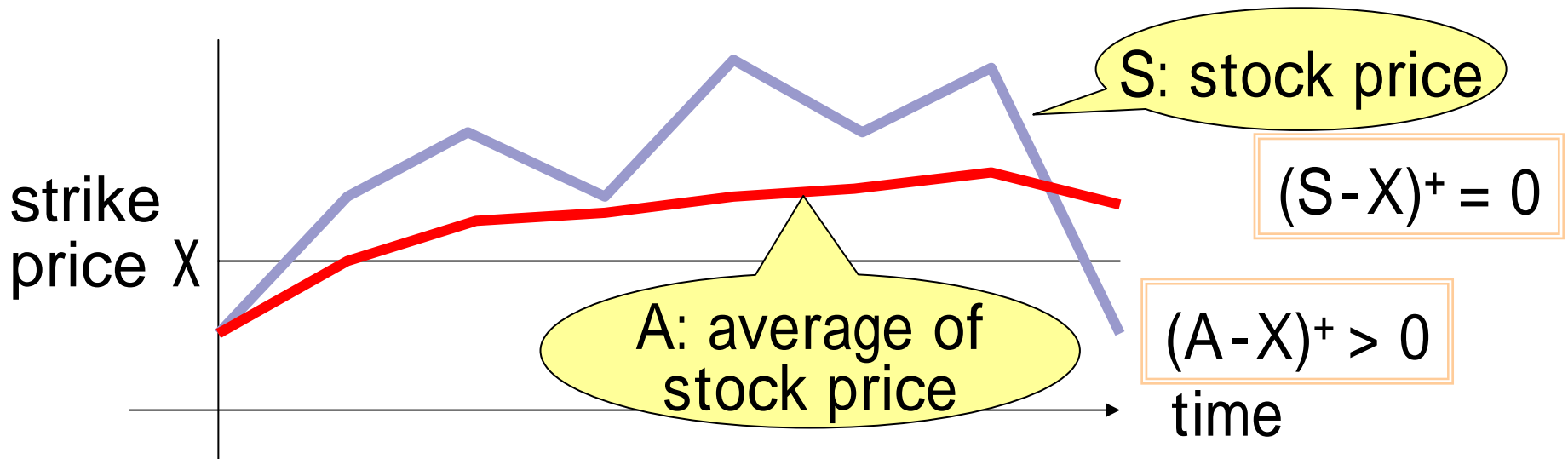
$$(S - X)^+ = \max\{S - X, 0\}$$

(S : stock price at expiration date, X : strike price)

European-Asian Option

- payoff of **European-Asian option** depends on **average of stock price A** during whole period

$$\text{payoff: } (A - X)^+ = \max\{A - X, 0\}$$

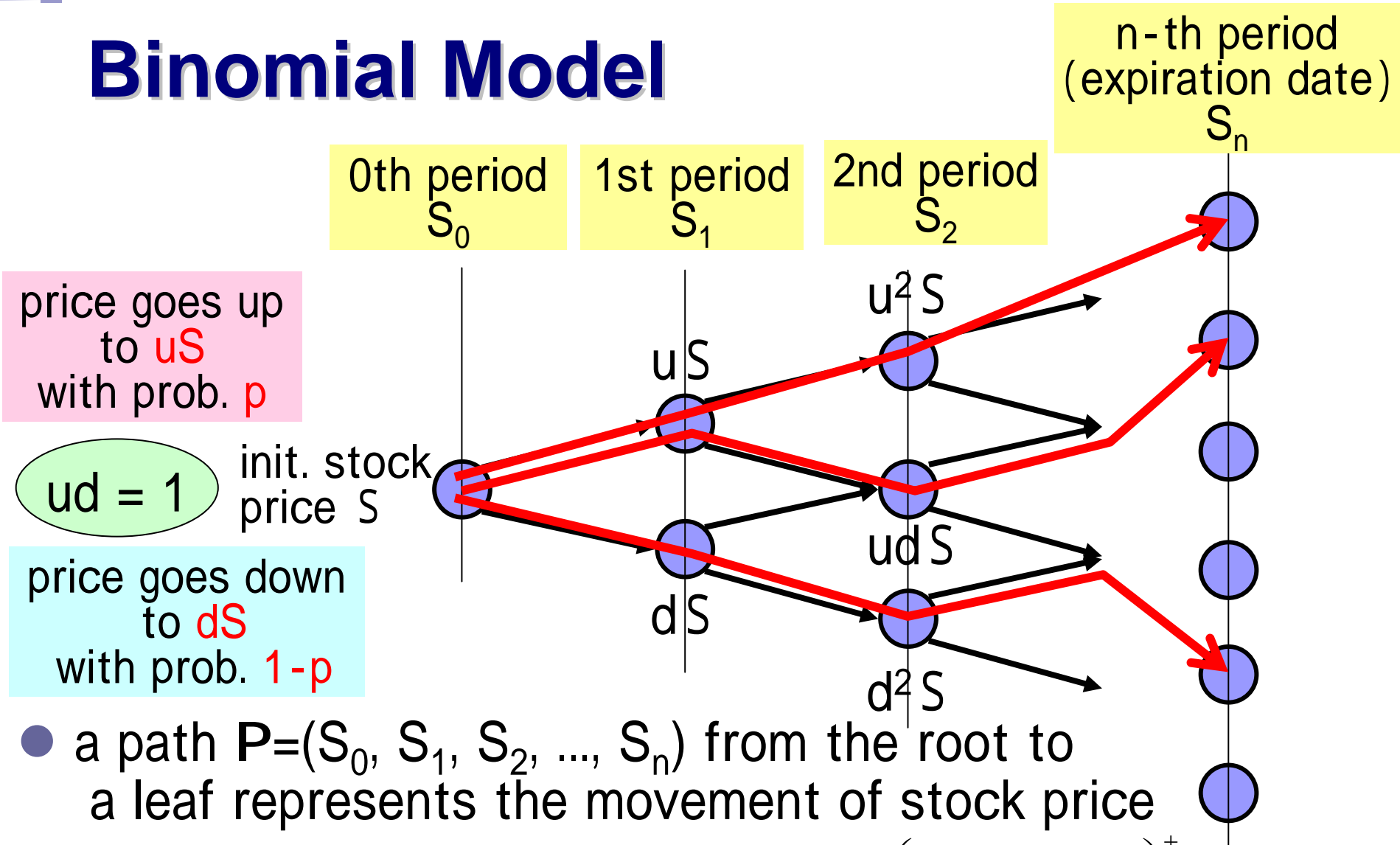


safe against fluctuation of stock price

Computation of Option Price

- price of option = discounted **expected value of payoff**
--- need to model the movement of stock price
- Our model: **binomial model** (discrete model)
 - proposed by Cox, Ross & Rubinstein (1979)
 - represent stock price movement
by a binomial tree
 - can compute exact option price by D P

Binomial Model



● payoff of European-Asian option = $\left(\frac{\sum_{i=0}^n S_i}{n+1} - X \right)^+$

Our Problem

compute the expected payoff
of European-Asian option
on the binomial model

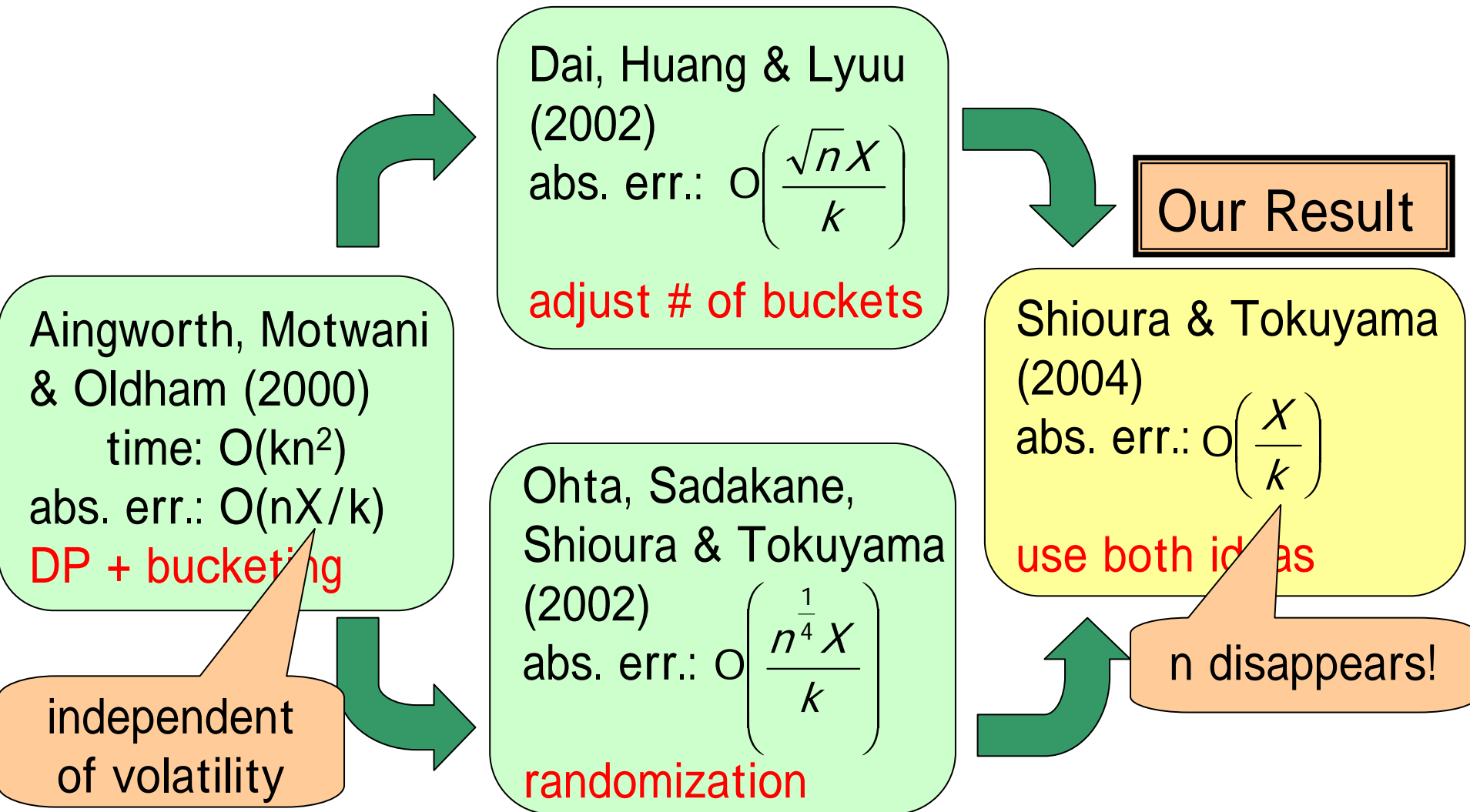
$$E \left(\left(\frac{\sum_{i=0}^n S_i}{n+1} - X \right)^+ \right)$$

- payoff is dependent on the path $\mathbf{P}=(S_0, S_1, S_2, \dots, S_n)$
(path-dependent option)
- payoff is nonlinear w.r.t. the running total $\sum_{i=0}^n S_i$
need enumeration of all the paths
exponential time

Approximation Algorithms for Pricing European-Asian Option

- Monte Carlo Method
 - based on path sampling
 - error bound depends on the volatility of stock price
- Other methods
 - based on heuristics
 - no theoretical error bound

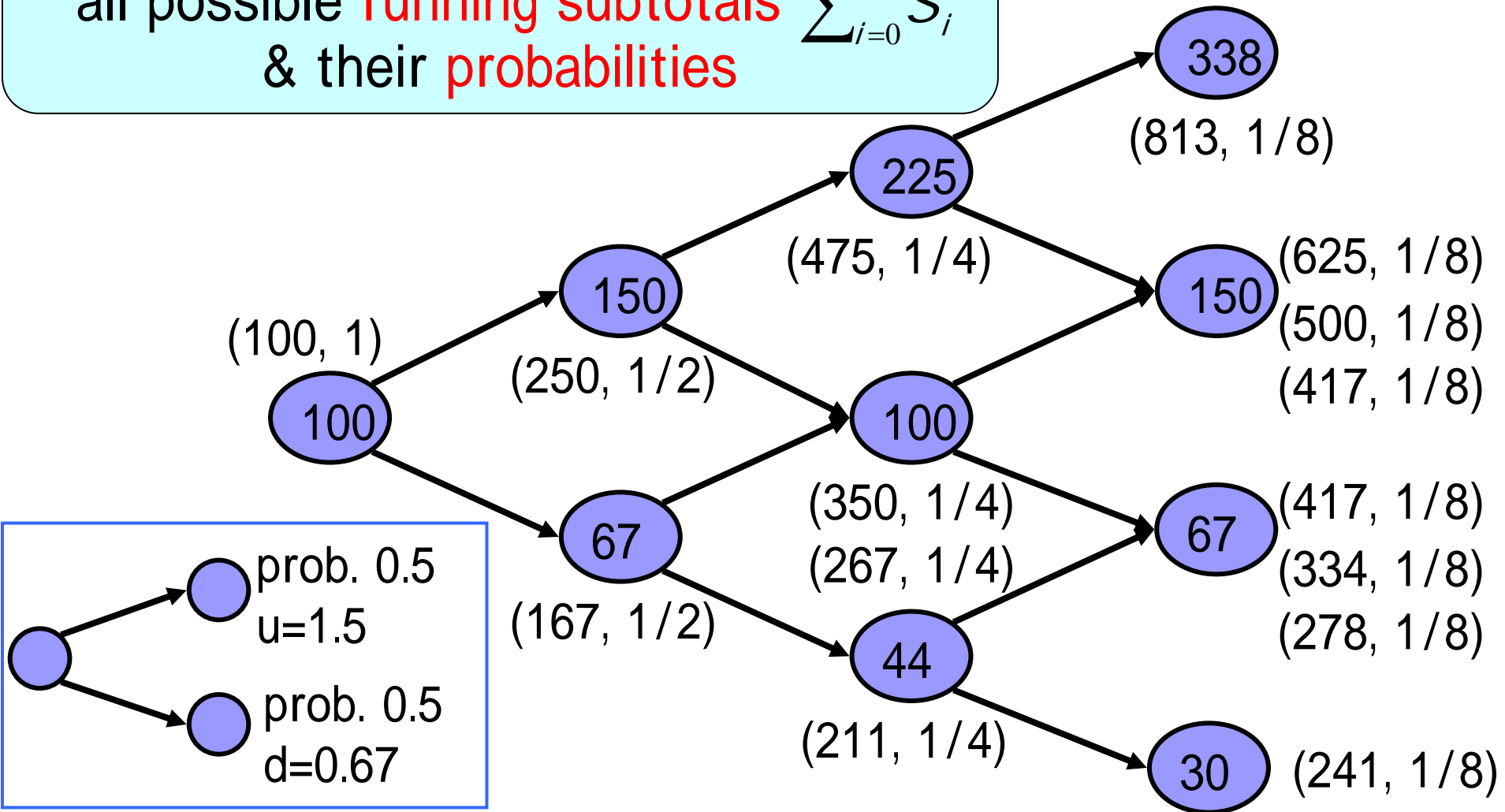
AMO Algorithm and its Variants



(n : depth of binomial tree, X : strike price, k : positive integer)

Exact Algorithm by DP

at each node of binomial tree, compute all possible **running subtotals** $\sum_{i=0}^t S_i$ & their **probabilities**



AMO Algorithm (1)

- # of running subtotals can be exponential
approximate running subtotals by bucketing

interval	running subtotal & probability
400 300	(310, 0.05)
300	(205, 0.15)
200	(240, 0.12)
200	(285, 0.20)
100	(170, 0.10)
100	(150, 0.10)
100	(110, 0.10)
100 0	(80, 0.05)
	(30, 0.01)

round up
running subtotals
&
sum up
probabilities
in each bucket



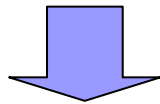
400 300	(400, 0.05)
300 200	(300, 0.47)
200 100	(200, 0.30)
100 0	(100, 0.06)

AMO Algorithm (2)

- k: # of buckets at each node
error bound **max. value of running subtotal/k**

Proposition:

running subtotal $\sum_{i=0}^t S_i$ is **$(n+1)X$**
at the t-th period

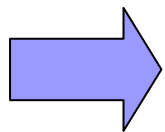


- option will be exercised at the expiration date
- conditional expectation of the payoff
can be computed easily

error bound of AMO algorithm = **$(n+1) X/k$**

Algorithm by Dai et al. (2002)

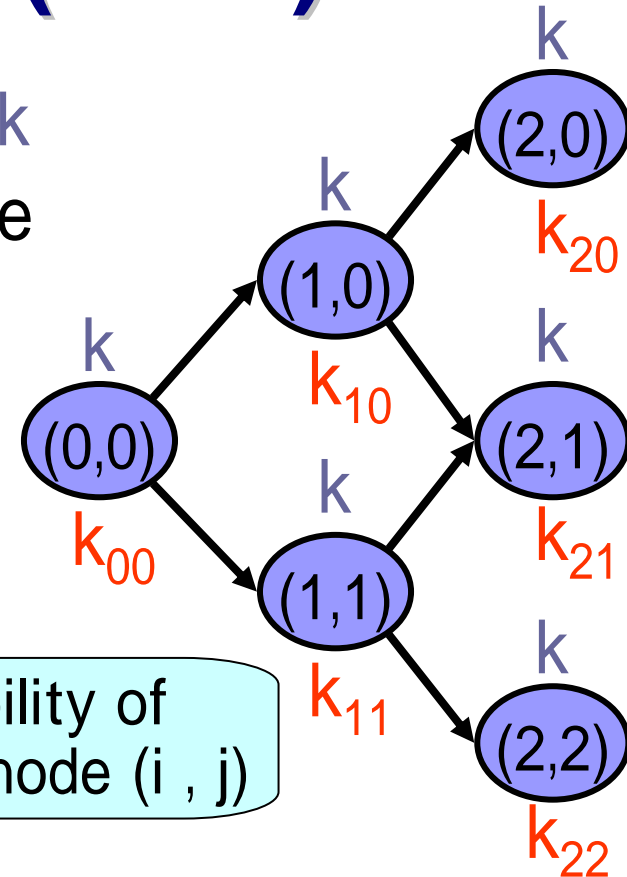
AMO algorithm: use the same number k of buckets at each node



set the number of buckets k_{ij} at the node (i, j) flexibly

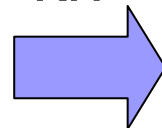
error bound
$$\sum_{i=0}^n \sum_{j=0}^i \frac{X}{k_{ij}} \omega(i, j)$$

probability of reaching node (i, j)



adjust # of buckets k_{ij} to minimize error bound under the condition

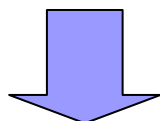
$$k_{ij} = kn^2$$



error bound
$$O\left(\frac{\sqrt{n}X}{k}\right)$$

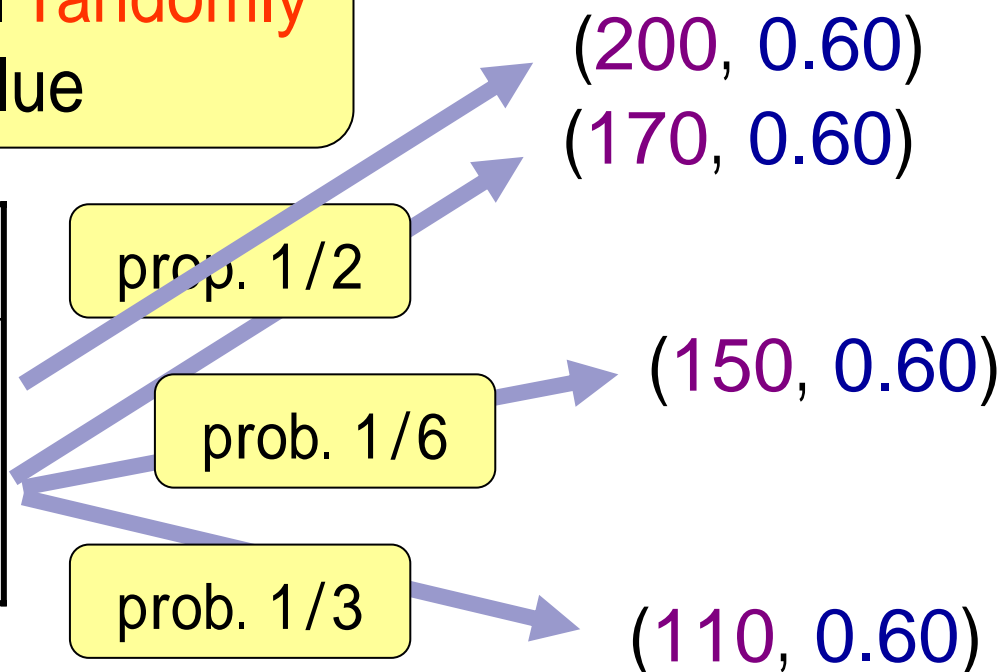
Algorithm by Ohta et al. (2002)

AMO algorithm: approximate running subtotals
in a bucket by **rounding-up**



choose a running subtotal **randomly**
as approximate value

interval	running subtotal & probability
200	(170, 0.30) (150, 0.10)
100	(110, 0.20)



Analysis of Ohta et al. (2002)

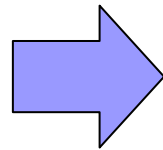
- regard the behavior of randomized algorithm as **stochastic process** **Martingale**
expectation of the error by random choice of running totals at a node = 0
apply Azuma's inequality (1967)

error bound $O\left(\frac{n^{\frac{1}{4}} X}{k}\right)$ (with high probability)

analysis is difficult

Our Algorithm

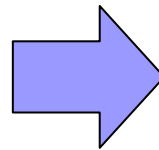
- set the **number of buckets** k_{ij} at node (i, j) **flexibly**
- **random choice of running subtotal**



error bound

$$O\left(X \sqrt{\sum_{i=1}^n \sum_{j=0}^i \left[\frac{\omega(i, j)}{k_{ij}} \right]^2}\right)$$

- **adjust # of buckets** k_{ij} to minimize error bound under the condition $k_{ij} = kn^2$



error bound $O\left(\frac{X}{k}\right)$

analysis is quite easy!

Open Problems

- derandomization of our algorithm with the same error bound
- approximation of American-Asian option
- analysis of error bound compared to exact price