

1 Algorithms for Lecture 11

Algorithm 1.1

LATTICE ALGORITHM FOR PREDICTION ERRORS		Number of operations/ equation		Number of operations/ Step
		★	/	+
1. Initialize at ($N = 0$) For $n = 0, \dots, n_{max} - 1$ $\alpha_n(0) = 0$ $\sigma_0(0) = \sigma_0^b(0) = 0$				$\mathcal{O}(n_{max}^2)$
2. Recursions in time For $N = 1, 2, \dots, N_{max}$				
2.1 Initialize recursions in order $e_0^f(N) = \varepsilon_0^f(N) = y(N)$ $e_0^b(N) = \varepsilon_0^b(N) = y(N - 1)$ $\sigma_0(N) = \sigma_0(N - 1) + \varepsilon_0^{f2}(N)$ (1) $\sigma_0^b(N) = \sigma_0(N - 1)$ (2)		1	0	1
2.2 Recursions in order: For $n = 0, \dots, n_{max} - 1$ $\alpha_n(N) = \alpha_n(N - 1) + e_n^f(N)\varepsilon_n^b(N)$ (3) $K_{n+1}^f(N) = -\alpha_n(N)/\sigma_n^b(N)$ (★) (4) $K_{n+1}^b(N) = -\alpha_n(N)/\sigma_n(N)$ (★) (5) $\sigma_{n+1}(N) = \sigma_n(N)(1 - K_{n+1}^f(N)K_{n+1}^b(N))$ (6) $\sigma_{n+1}^b(N) = \sigma_n^b(N - 1)(1 - K_{n+1}^f(N - 1)K_{n+1}^b(N - 1))$ (7) $e_{n+1}^f(N) = e_n^f(N) + K_{n+1}^f(N - 1)e_n^b(N)$ (8) $e_{n+1}^b(N) = e_n^b(N - 1) + K_{n+1}^b(N - 2)e_n^f(N - 1)$ (9) $\varepsilon_{n+1}^f(N) = \varepsilon_n^f(N) + K_{n+1}^f(N)\varepsilon_n^b(N)$ (10) $\varepsilon_{n+1}^b(N) = \varepsilon_n^b(N - 1) + K_{n+1}^b(N - 1)\varepsilon_n^f(N - 1)$ (11)		1	0	1
(★) : If $\sigma_n^b(N) = 0$ assign $K_{n+1}^f(N) = 0$ If $\sigma_n(N) = 0$ assign $K_{n+1}^b(N) = 0$				$\mathcal{O}(n_{max})$
Total: $8n_{max}$ multiplications; $2n_{max}$ divisions; $6n_{max}$ additions				

Extension of Algorithm 1.1

OPTIONAL SECTION FOR ORDER RECURSION OF PREDICTORS	Number of operations/ equation		
	★	/	+
<p>2.3 If the predictors are desired at time N</p> <p>Do:</p> <div style="border-left: 1px solid black; padding-left: 10px; margin-left: 20px;"> <p>2.3.1 Initialize order recursions:</p> $\theta_1(N) = -\alpha_0(N)/\sigma_0^b(N)$ $\theta_1^b(N) = -\alpha_0(N-1)/\sigma_0(N-1)$ $K_1^*(N) = -\varepsilon_0^b(N)/\sigma_0^b(N)$ <p>2.3.2 Order recursions:</p> <p>For $n = 1, \dots, n_{max} - 1$</p> $\theta_{n+1}(N) = \begin{bmatrix} \theta_n(N) \\ 0 \end{bmatrix} - \frac{\alpha_n(N)}{\sigma_n^b(N)} \begin{bmatrix} \theta_n^b(N) \\ 1 \end{bmatrix}$ $\theta_{n+1}^b(N) = \begin{bmatrix} 0 \\ \theta_n^b(N) \end{bmatrix} - \frac{\alpha_n(N-1)}{\sigma_n(N-1)} \begin{bmatrix} 1 \\ \theta_n(N) \end{bmatrix} -$ $- (e_n^b(N) - \frac{\alpha_n(N-1)}{\sigma_n(N-1)} e_n^f(N)) \begin{bmatrix} 0 \\ K_n^*(N) \end{bmatrix}$ $K_{n+1}^*(N) = \begin{bmatrix} K_n^*(N) \\ 0 \end{bmatrix} - \frac{\varepsilon_n^b(N)}{\sigma_n^b(N)} \begin{bmatrix} \theta_n^b(N) \\ 1 \end{bmatrix}$ </div>			
Total: $\mathcal{O}(n_{max}^2)$ multiplications; $\mathcal{O}(n_{max})$ divisions; $\mathcal{O}(n_{max}^2)$ additions			

Algorithm 1.2

LATTICE ALGORITHM FOR PREDICTION ERRORS AND PREDICTORS		Num. op.		
		★	/	+
1.	Initialize at time ($N = 0$)			
	For $n = 0, \dots, n_{max} - 1$ $\alpha_n(0) = 0$			
	$\sigma_0(0) = \sigma_0^b(0) = 0$			
2.	Time recursions			
	For $N = 1, 2, \dots, N_{max}$			
2.1	Initialize order recursions			
	$\varepsilon_0^f(N) = y(N); \quad \varepsilon_0^b(N) = y(N - 1)$			
	$\sigma_0(N) = \sigma_0(N - 1) + \varepsilon_0^{f2}(N); \quad \sigma_0^b(N) = \sigma_0(N - 1)$	1	0	1
	$\gamma_0^0(N) = 1$			
2.2	Recursions in order:			
	ntru $n = 0, \dots, n_{max} - 1$			
	$\alpha_n(N) = \alpha_n(N - 1) + \varepsilon_n^f(N)\varepsilon_n^b(N)/\gamma_n^0(N)$ (1)	1	1	1
	$K_{n+1}^f(N) = -\alpha_n(N)/\sigma_n^b(N)$ (★) (2)	0	1	0
	$K_{n+1}^b(N) = -\alpha_n(N)/\sigma_n(N)$ (★) (3)	0	1	0
	$\sigma_{n+1}(N) = \sigma_n(N)(1 - K_{n+1}^f(N)K_{n+1}^b(N))$ (4)	2	0	1
	$\sigma_{n+1}^b(N) = \sigma_n^b(N - 1)(1 - K_{n+1}^f(N - 1)K_{n+1}^b(N - 1))$ (5)	1	0	0
	$\varepsilon_{n+1}^f(N) = \varepsilon_n^f(N) + K_{n+1}^f(N)\varepsilon_n^b(N)$ (6)	1	0	1
	$\varepsilon_{n+1}^b(N) = \varepsilon_n^b(N - 1) + K_{n+1}^b(N - 1)\varepsilon_n^f(N - 1)$ (7)	1	0	1
	$\gamma_{n+1}^0(N) = \gamma_n^0(N) - \varepsilon_n^{b2}(N)/\sigma_n^b(N)$ ★ (8)	1	0	1
2.3	If the predictors are needed at time N			
	Then:			
	2.3.1 Initialize order recursions:			
	$\theta_1(N) = K_1^f(N); \quad \theta_1^b(N) = K_1^b(N - 1)$			
	$K_1^*(N) = -\varepsilon_0^b(N)/\sigma_0^b(N)$ (9)		1	
	2.3.2 Recursions in order:			
	For $n = 1, \dots, n_{max} - 1$			
	$\theta_{n+1}(N) = \begin{bmatrix} \theta_n(N) \\ 0 \end{bmatrix} + K_{n+1}^f(N) \begin{bmatrix} \theta_n^b(N) \\ 1 \end{bmatrix}$ (10)	n	0	n
	$\theta_{n+1}^b(N) = \begin{bmatrix} 0 \\ \theta_n^b(N) \end{bmatrix} + K_{n+1}^b(N - 1) \begin{bmatrix} 1 \\ \theta_n(N) \end{bmatrix} -$			
	$\quad - \frac{(\varepsilon_n^b(N) + K_{n+1}^b(N - 1)\varepsilon_n^f(N))}{\gamma_n^0(N)} \begin{bmatrix} 0 \\ K_n^*(N) \end{bmatrix}$ (11)	$2n+1$	1	$2n+1$
	$K_{n+1}^*(N) = \begin{bmatrix} K_n^*(N) \\ 0 \end{bmatrix} - \frac{\varepsilon_n^b(N)}{\sigma_n^b(N)} \begin{bmatrix} \theta_n^b(N) \\ 1 \end{bmatrix}$ (12)	n	1	n
(★) : If any of $\sigma_n(N), \sigma_n^b(N)$ sau $\gamma_n^0(N)$ are zero, set the corresponding division results to 0 0.				