

# Dynamic Programming with Shape Preservation

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# Shape Preservation

Goal: Suppose

- ▶ theory tells us that a value function is concave and increasing, and
- ▶ the data from value function iteration is consistent with the shape information

then

- ▶ we want the approximation to satisfy the shape information
- ▶ which presumably will improve stability, and
- ▶ presumably improve accuracy

Challenge #1: Neither interpolation nor regression preserves shape of the data

- ▶ These methods create approximations that minimize a norm of the errors, choosing among a vector space of functions
- ▶ Shapes define cones of functions: monotone, concave, etc.
- ▶ Cones do not have countable bases, not even if we limit ourselves to convex combinations of basis functions

Challenge #2: Shape constraints are infinitistic

- ▶  $f$  monotonically increasing means  $f'(x) > 0$  for  $a \leq x \leq b$
- ▶  $f$  concave means  $f''(x) < 0$  for  $a \leq x \leq b$
- ▶ Finding a function of a particular shape is a problem with an infinite number of constraints!

# Shape-preserving Chebyshev Interpolation: Attempt 1

LP model for shape-preserving Chebyshev Interpolation: Suppose

- ▶ Data from max step is  $(z_i, v_i)$ , and consistent with concavity and monotonicity, and
- ▶ we want an interpolating Chebyshev polynomial,  $\sum_{j=0}^{m-1} c_j T_j(z)$  with same shape

Idea: Add shape constraints to interpolation problem at shape points  $y_i, i = 1, \dots, m'$  where  $m' > m$ :

$$\begin{aligned} \min_{c_j} \quad & \sum_{j=0}^{m-1} c_j \\ \text{s.t.} \quad & \sum_{j=0}^n c_j T_j'(y_i) > 0 > \sum_{j=0}^n c_j T_j''(y_i), \quad i = 1, \dots, m', \\ & \sum_{j=0}^n c_j T_j(z_i) = v_i, \quad i = 1, \dots, m, \end{aligned}$$

## Shape-preserving Chebyshev Interpolation: Attempt 2

Add more basis functions, for a total of  $n > m$ , enough so that we are sure there will be a shape-preserving interpolating polynomial:

$$\begin{aligned} \min_{c_j} \quad & \sum_{j=0}^n c_j \\ \text{s.t.} \quad & \sum_{j=0}^n c_j T_j'(y_i) > 0 > \sum_{j=0}^n c_j T_j''(y_i), \quad i = 1, \dots, m', \\ & \sum_{j=0}^n c_j T_j(z_i) = v_i, \quad i = 1, \dots, m, \end{aligned}$$

Problem: Undetermined. Too many solutions, particularly since we don't know minimal number required extra basis functions.

## Shape-preserving Chebyshev Interpolation: Attempt 3

Select a nice solution by penalizing the high-order basis elements:

$$\begin{aligned} \min_{c_j} \quad & \sum_{j=0}^{m-1} c_j + \sum_{j=m}^n (j+1-m)^2 c_j \\ \text{s.t.} \quad & \sum_{j=0}^n c_j T_j'(y_i) > 0 > \sum_{j=0}^n c_j T_j''(y_i), \quad i = 1, \dots, m', \\ & \sum_{j=0}^n c_j T_j(z_i) = v_i, \quad i = 1, \dots, m, \end{aligned}$$

Problem: No theoretical problem - for fixed  $n$ , generally get a unique solution. (Note: different  $n$  could produce different function)

# Shape-preserving Chebyshev Interpolation: Success

LP model for shape-preserving Chebyshev Interpolation: Improve computational performance by

- ▶ using the pure interpolant,  $\hat{c}_j$ , as the initial guess, and
- ▶ decompose coefficients into positive and negative parts

$$\begin{aligned} \min_{c_j^+, c_j^-, c_j^-} \quad & \sum_{j=0}^{m-1} (c_j^+ + c_j^-) + \sum_{j=m}^n (j+1-m)^2 (c_j^+ + c_j^-) \\ \text{s.t.} \quad & \sum_{j=0}^n c_j T_j'(y_i) > 0 > \sum_{j=0}^n c_j T_j''(y_i), \quad i = 1, \dots, m', \\ & \sum_{j=0}^n c_j T_j(z_i) = v_i, \quad i = 1, \dots, m, \\ & c_j - \hat{c}_j = c_j^+ - c_j^-, \quad j = 0, \dots, m-1, \\ & c_j = c_j^+ - c_j^-, \quad j = m, \dots, n, \\ & c_j^+ \geq 0, \quad c_j^- \geq 0, \quad j = 1, \dots, n. \end{aligned}$$

# Optimal Growth Models

- ▶ Optimal Growth Problem:

$$V_0(k_0) = \max_{c,l} \sum_{t=0}^{T-1} \beta^t u(c_t, l_t) + \beta^T V_T(k_T),$$

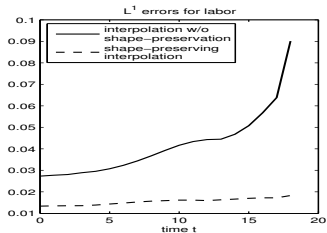
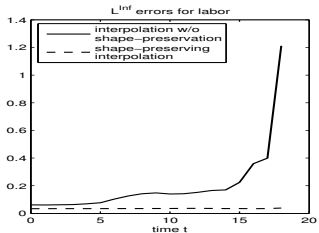
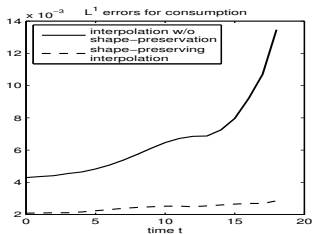
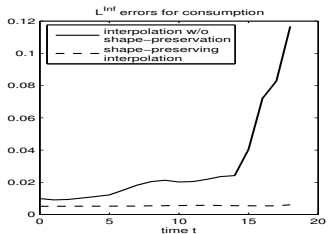
s.t.  $k_{t+1} = F(k_t, l_t) - c_t, \quad 0 \leq t < T$

- ▶ DP model of optimal growth problem:

$$V_t(k) = \max_{c,l} u(c, l) + \beta V_{t+1}(F(k, l) - c)$$



# Errors of NDP with Chebyshev interpolation (shape-preserving or not)



# Multi-Stage Portfolio Optimization

- ▶  $W_t$ : wealth at stage  $t$ ; stocks' random return:  $R = (R_1, \dots, R_n)$ ; bond's riskfree return:  $R_f$ ;
- ▶  $S_t = (S_{t1}, \dots, S_{tn})^\top$ : money in the stocks;  $B_t = W_t - e^\top S_t$ : money in the bond,
- ▶  $W_{t+1} = R_f(W_t - e^\top S_t) + R^\top S_t$
- ▶ Multi-Stage Portfolio Optimization Problem:

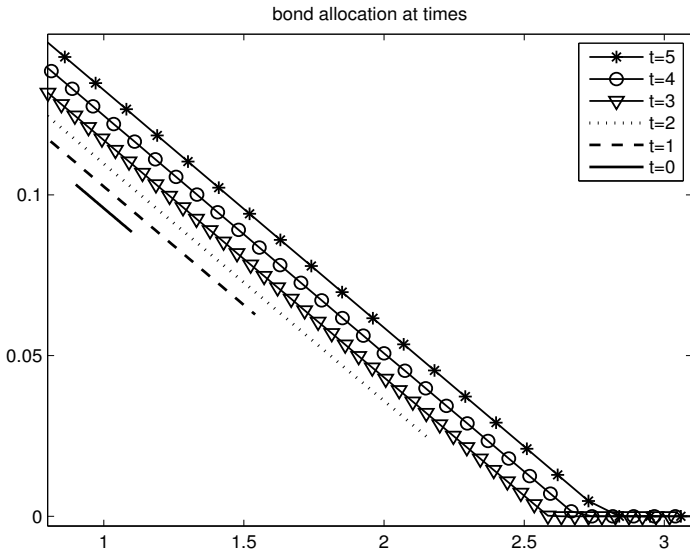
$$V_0(W_0) = \max_{x_t, 0 \leq t < T} E\{u(W_T)\}$$

- ▶ Bellman Equation:

$$V_t(W) = \max_S E\{V_{t+1}(R_f(W - e^\top S) + R^\top S)\}$$

$W$ : state variable;  $S$ : control variables.

# Exact optimal bond allocation



# Errors of Optimal Stock Allocations (shape-preserving or not)

