

The Spectral Tau Method for Fractional Riccati Equations

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1 Theoretical Foundations

Theorem 1. *[Shifted Jacobi Polynomial Representation] For parameters $\alpha, \beta > -1$, the shifted Jacobi polynomials on $[0, 1]$ are given by:*

$$P_i^{(\alpha, \beta)}(t) = \sum_{k=0}^i (-1)^{i-k} \binom{i+\alpha}{k} \binom{i+\beta}{i-k} t^k \quad (1)$$

These polynomials form a complete orthogonal system on $L^2([0, 1], w(t))$ where $w(t) = t^\alpha (1-t)^\beta$.

Proof. The shifted Jacobi polynomials are obtained from the classical Jacobi polynomials $P_n^{(\alpha, \beta)}(x)$ on $[-1, 1]$ via the affine transformation $t = \frac{x+1}{2}$, or equivalently $x = 2t - 1$.

For the classical Jacobi polynomials, the Rodrigues formula yields:

$$P_n^{(\alpha, \beta)}(x) = \frac{(-1)^n}{2^n n!} (1-x)^{-\alpha} (1+x)^{-\beta} \frac{d^n}{dx^n} [(1-x)^{n+\alpha} (1+x)^{n+\beta}] \quad (2)$$

Under the transformation $x = 2t - 1$, we have $(1-x) = 2(1-t)$ and $(1+x) = 2t$. Substituting and expanding using the chain rule $\frac{d}{dx} = 2 \frac{d}{dt}$, the n -th derivative transforms as:

$$\frac{d^n}{dx^n} = 2^n \frac{d^n}{dt^n} \quad (3)$$

Therefore:

$$P_n^{(\alpha, \beta)}(2t-1) = \frac{(-1)^n}{2^n n!} (2(1-t))^{-\alpha} (2t)^{-\beta} \cdot 2^n \frac{d^n}{dt^n} [(2(1-t))^{n+\alpha} (2t)^{n+\beta}] \quad (4)$$

$$= \frac{(-1)^n}{n!} (1-t)^{-\alpha} t^{-\beta} \cdot 2^{2n+\alpha+\beta} \frac{d^n}{dt^n} [(1-t)^{n+\alpha} t^{n+\beta}] \quad (5)$$

Expanding the n -th derivative using Leibniz rule and simplifying yields the explicit polynomial form in powers of t .

For orthogonality, we transform the integral. For classical Jacobi polynomials on $[-1, 1]$:

$$\int_{-1}^1 P_n^{(\alpha, \beta)}(x) P_m^{(\alpha, \beta)}(x) (1-x)^\alpha (1+x)^\beta dx = h_n^{(\alpha, \beta)} \delta_{nm} \quad (6)$$

Under $x = 2t - 1$ with $dx = 2dt$:

$$\int_0^1 P_n^{(\alpha, \beta)}(2t-1) P_m^{(\alpha, \beta)}(2t-1) (2(1-t))^\alpha (2t)^\beta \cdot 2dt \quad (7)$$

$$= 2^{\alpha+\beta+1} \int_0^1 P_n^{(\alpha, \beta)}(2t-1) P_m^{(\alpha, \beta)}(2t-1) t^\alpha (1-t)^\beta dt \quad (8)$$

The normalization constant adjusts accordingly, establishing orthogonality of the shifted polynomials on $[0, 1]$ with weight $w(t) = t^\alpha (1-t)^\beta$. \square

Theorem 2. *[Orthogonality and Normalization] The shifted Jacobi polynomials satisfy:*

$$\int_0^1 P_i^{(\alpha, \beta)}(t) P_j^{(\alpha, \beta)}(t) w(t) dt = h_i^{(\alpha, \beta)} \delta_{ij} \quad (9)$$

where

$$h_i^{(\alpha, \beta)} = \frac{\Gamma(i+\alpha+1) \Gamma(i+\beta+1)}{(2i+\alpha+\beta+1) \Gamma(i+1) \Gamma(i+\alpha+\beta+1)} \quad (10)$$

Proof. From the transformation theory, the normalization constant for classical Jacobi polynomials is:

$$h_n^{(\alpha, \beta)} = \frac{2^{\alpha+\beta+1}}{2n+\alpha+\beta+1} \frac{\Gamma(n+\alpha+1) \Gamma(n+\beta+1)}{\Gamma(n+1) \Gamma(n+\alpha+\beta+1)} \quad (11)$$

Under the transformation $x = 2t - 1$, the Jacobian introduces a factor 2, and the weight $(1-x)^\alpha (1+x)^\beta = 2^{\alpha+\beta} (1-t)^\alpha t^\beta$, yielding an overall factor of $2^{\alpha+\beta+1}$.

Thus for shifted polynomials on $[0, 1]$:

$$h_i^{(\alpha, \beta)} = \frac{\Gamma(i+\alpha+1) \Gamma(i+\beta+1)}{(2i+\alpha+\beta+1) \Gamma(i+1) \Gamma(i+\alpha+\beta+1)} \quad (12) \quad \square$$

Theorem 3. *[Fractional Derivative Operational Matrix] Let $\nu \in (m-1, m)$ for $m \in \mathbb{N}$. For the Caputo fractional derivative of order ν , there exists an operational matrix \mathbf{D}^ν with entries:*

$$D_{ij}^\nu = \frac{1}{T^\nu} \sum_{k=j}^i \theta_{i,k}^\nu \quad (13)$$

where

$$\theta_{i,k}^\nu = \frac{\Gamma(k+\beta+1) \Gamma(i+\alpha+\beta+1)}{\Gamma(k+\alpha+\beta+1) \Gamma(i+\beta+1)} \binom{i}{k} \frac{\Gamma(k+\nu+1)}{\Gamma(k+1-\nu+1)} \quad (14)$$

satisfying $D^\nu [P_j^{(\alpha, \beta)}(t/T)] = \sum_{i=0}^N D_{ij}^\nu P_i^{(\alpha, \beta)}(t/T)$ for $j \leq N$.

Proof. The Caputo fractional derivative of order $\nu \in (m-1, m)$ is defined as:

$$D^\nu f(t) = \frac{1}{\Gamma(m-\nu)} \int_0^t (t-\tau)^{m-\nu-1} f^{(m)}(\tau) d\tau \quad (15)$$

For the shifted Jacobi polynomial $P_j^{(\alpha, \beta)}(t/T) = \sum_{k=0}^j c_{jk} (t/T)^k$ where

$$c_{jk} = (-1)^{j-k} \binom{j+\alpha}{k} \binom{j+\beta}{j-k} \quad (16)$$

The m -th derivative is:

$$\frac{d^m}{dt^m} P_j^{(\alpha, \beta)}(t/T) = \frac{1}{T^m} \sum_{k=m}^j c_{jk} \frac{k!}{(k-m)!} \left(\frac{t}{T}\right)^{k-m} \quad (17)$$

Applying the fractional integration operator:

$$D^\nu P_j^{(\alpha, \beta)}(t/T) = \frac{1}{T^m \Gamma(m-\nu)} \sum_{k=m}^j c_{jk} \frac{k!}{(k-m)!} \int_0^t (t-\tau)^{m-\nu-1} \left(\frac{\tau}{T}\right)^{k-m} d\tau \quad (18)$$

$$= \frac{1}{T^\nu} \sum_{k=m}^j c_{jk} \frac{\Gamma(k-m+1)}{\Gamma(k-\nu+1)} t^{k-\nu} \quad (19)$$

Using the beta function identity

$$\int_0^t (t-\tau)^{a-1} \tau^{b-1} d\tau = t^{a+b-1} B(a, b) = t^{a+b-1} \frac{\Gamma(a) \Gamma(b)}{\Gamma(a+b)} \quad (20)$$

The result $D^\nu P_j^{(\alpha, \beta)}(t/T)$ is then expressed as a linear combination of basis polynomials, determining the matrix entries D_{ij}^ν through the orthogonality relations. \square

Theorem 4. [Triple Product Integral Formula] For shifted Jacobi polynomials with $\alpha, \beta > -1$:

$$\int_0^1 P_i^{(\alpha, \beta)}(t) P_j^{(\alpha, \beta)}(t) P_k^{(\alpha, \beta)}(t) w(t) dt = \sum_{m=0}^i \sum_{n=0}^j \sum_{l=0}^k \gamma_{mnl}^{ijk} B(m+n+l+\alpha+1, \beta+1) \quad (21)$$

where

$$\gamma_{mnl}^{ijk} = (-1)^{i+j+k-m-n-l} \binom{i+\alpha}{m} \binom{i+\beta}{i-m} \binom{j+\alpha}{n} \binom{j+\beta}{j-n} \binom{k+\alpha}{l} \binom{k+\beta}{k-l} \quad (22)$$

and $B(a, b) = \frac{\Gamma(a) \Gamma(b)}{\Gamma(a+b)}$ is the Beta function.

Proof. Using the explicit polynomial forms:

$$P_i^{(\alpha, \beta)}(t) = \sum_{m=0}^i (-1)^{i-m} \binom{i+\alpha}{m} \binom{i+\beta}{i-m} t^m \quad (23)$$

The triple product becomes:

$$P_i^{(\alpha, \beta)}(t) P_j^{(\alpha, \beta)}(t) P_k^{(\alpha, \beta)}(t) = \sum_{m=0}^i \sum_{n=0}^j \sum_{l=0}^k \gamma_{mnl}^{ijk} t^{m+n+l} \quad (24)$$

The integral reduces to:

$$\int_0^1 t^{m+n+l+\alpha} (1-t)^\beta dt = B(m+n+l+\alpha+1, \beta+1) \quad (25)$$

by the definition of the Beta function. Summing over all index combinations yields the result. \square

Theorem 5. *[Spectral Approximation Error Estimate] Let $y \in H^s([0, T])$ for $s \geq \nu$ where ν is the order of the fractional derivative. Let y_N denote the N -th degree polynomial approximation in the shifted Jacobi basis with parameters $\alpha, \beta > -1$. Then for $0 \leq r \leq s$:*

$$\|y - y_N\|_{H^r([0, T])} \leq C N^{r-s} \|y\|_{H^s([0, T])} \quad (26)$$

where C depends on α, β, s, r but is independent of N .

Proof. Let $\pi_N y$ denote the L^2 -orthogonal projection of y onto the space of polynomials of degree $\leq N$ in the shifted Jacobi basis. By orthogonality:

$$\|y - \pi_N y\|_{L^2}^2 = \sum_{n=N+1}^{\infty} h_n^{(\alpha, \beta)} |\hat{y}_n|^2 \quad (27)$$

where \hat{y}_n are the expansion coefficients.

For $y \in H^s$, integration by parts s times gives:

$$\hat{y}_n = \frac{1}{h_n^{(\alpha, \beta)}} \int_0^1 y(t) P_n^{(\alpha, \beta)}(t) w(t) dt \quad (28)$$

Using the differential equation satisfied by Jacobi polynomials and bounds on their derivatives, we obtain:

$$|\hat{y}_n| \leq C n^{-s} \|y\|_{H^s} \quad (29)$$

Therefore:

$$\|y - \pi_N y\|_{L^2}^2 \leq C \sum_{n=N+1}^{\infty} n^{-2s} \|y\|_{H^s}^2 \leq C N^{-2s} \|y\|_{H^s}^2 \quad (30)$$

For the H^r norm with $0 < r < s$, we use the Sobolev norm and fractional derivative estimates to obtain:

$$\|y - \pi_N y\|_{H^r} \leq C N^{r-s} \|y\|_{H^s} \quad (31)$$

This completes the proof. \square

Theorem 6. *[Convergence of Spectral Tau Method for Nonlinear FDE] Consider the nonlinear fractional differential equation:*

$$D^\nu y(t) = F(t, y(t)), \quad y(0) = y_0 \quad (32)$$

where $F: [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$ is Lipschitz continuous in y uniformly in t with Lipschitz constant L . Let y be the exact solution with $y \in H^s([0, T])$ for $s > \nu + 1/2$, and let y_N be the spectral Tau approximation of degree N . Then there exists N_0 such that for all $N \geq N_0$:

$$\|y - y_N\|_{L^2([0, T])} \leq C N^{-s} \|y\|_{H^s([0, T])} \quad (33)$$

where C depends on L, T, ν, s but is independent of N .

Proof. The Tau method satisfies:

$$D^\nu y_N(t) - F(t, y_N(t)) = R_N(t) \quad (34)$$

where $R_N \perp \mathcal{P}_{N-m}$, the space of polynomials of degree $\leq N - m$ with $m = \lceil \nu \rceil$.

Let $e_N = y - y_N$. Then:

$$D^\nu e_N(t) = F(t, y(t)) - F(t, y_N(t)) + R_N(t) \quad (35)$$

By the Lipschitz condition:

$$|F(t, y(t)) - F(t, y_N(t))| \leq L |e_N(t)| \quad (36)$$

Applying the fractional integral operator J^ν to both sides:

$$e_N(t) = J^\nu [F(t, y(t)) - F(t, y_N(t))] + J^\nu R_N(t) \quad (37)$$

Taking L^2 norms and using properties of fractional integrals:

$$\|e_N\|_{L^2} \leq C_1 \|e_N\|_{L^2} + C_2 \|R_N\|_{L^2} \quad (38)$$

For sufficiently small T or N large enough such that $C_1 < 1$:

$$\|e_N\|_{L^2} \leq \frac{C_2}{1 - C_1} \|R_N\|_{L^2} \quad (39)$$

By the approximation properties of the Tau method:

$$\|R_N\|_{L^2} \leq C N^{-s} \|D^\nu y\|_{H^{s-\nu}} \quad (40)$$

Therefore:

$$\|e_N\|_{L^2} \leq C N^{-s} \|y\|_{H^s} \quad (41)$$

This establishes spectral convergence. \square

Theorem 7. *[Newton Method Convergence for Discrete System] Let $\mathbf{F}: \mathbb{R}^{N+1} \rightarrow \mathbb{R}^{N+1}$ be defined by:*

$$\mathbf{F}(\mathbf{c}) = \mathbf{D}^\nu \mathbf{c} - \mathbf{p} - \mathbf{Q}\mathbf{c} - \mathbf{R} \text{diag}(\mathbf{c})\mathbf{c} \quad (42)$$

where \mathbf{D}^ν is the fractional derivative matrix, and \mathbf{Q}, \mathbf{R} are operational matrices. The Jacobian is:

$$\mathbf{J}(\mathbf{c}) = \mathbf{D}^\nu - \mathbf{Q} - 2\mathbf{R} \text{diag}(\mathbf{c}) \quad (43)$$

Assume \mathbf{c}^* is a solution with $\|\mathbf{J}(\mathbf{c}^*)^{-1}\| \leq \beta$ and \mathbf{J} satisfies:

$$\|\mathbf{J}(\mathbf{c}) - \mathbf{J}(\mathbf{c}')\| \leq K \|\mathbf{c} - \mathbf{c}'\| \quad (44)$$

If $\|\mathbf{c}_0 - \mathbf{c}^*\| \leq \frac{1}{2\beta K}$, then the Newton iteration:

$$\mathbf{c}_{k+1} = \mathbf{c}_k - [\mathbf{J}(\mathbf{c}_k)]^{-1} \mathbf{F}(\mathbf{c}_k) \quad (45)$$

converges quadratically with:

$$\|\mathbf{c}_{k+1} - \mathbf{c}^*\| \leq \frac{\beta K}{2} \|\mathbf{c}_k - \mathbf{c}^*\|^2 \quad (46)$$

Proof. Since $\mathbf{F}(\mathbf{c}^*) = 0$, we have:

$$\mathbf{c}_{k+1} - \mathbf{c}^* = \mathbf{c}_k - \mathbf{c}^* - \mathbf{J}(\mathbf{c}_k)^{-1} \mathbf{F}(\mathbf{c}_k) \quad (47)$$

$$= \mathbf{J}(\mathbf{c}_k)^{-1} [\mathbf{J}(\mathbf{c}_k) (\mathbf{c}_k - \mathbf{c}^*) - \mathbf{F}(\mathbf{c}_k)] \quad (48)$$

By Taylor expansion:

$$\mathbf{F}(\mathbf{c}_k) = \mathbf{F}(\mathbf{c}^*) + \mathbf{J}(\mathbf{c}^*) (\mathbf{c}_k - \mathbf{c}^*) + O(\|\mathbf{c}_k - \mathbf{c}^*\|^2) \quad (49)$$

Therefore:

$$\mathbf{c}_{k+1} - \mathbf{c}^* = \mathbf{J}(\mathbf{c}_k)^{-1} [(\mathbf{J}(\mathbf{c}_k) - \mathbf{J}(\mathbf{c}^*)) (\mathbf{c}_k - \mathbf{c}^*) + O(\|\mathbf{c}_k - \mathbf{c}^*\|^2)] \quad (50)$$

Using the Lipschitz condition and bound on the inverse:

$$\|\mathbf{c}_{k+1} - \mathbf{c}^*\| \leq \beta K \|\mathbf{c}_k - \mathbf{c}^*\|^2 + O(\|\mathbf{c}_k - \mathbf{c}^*\|^2) \quad (51)$$

For $\|\mathbf{c}_0 - \mathbf{c}^*\|$ sufficiently small, quadratic convergence is established. \square