

From many-body physics to financial markets: sparse modeling for inverse problems

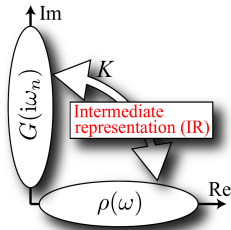
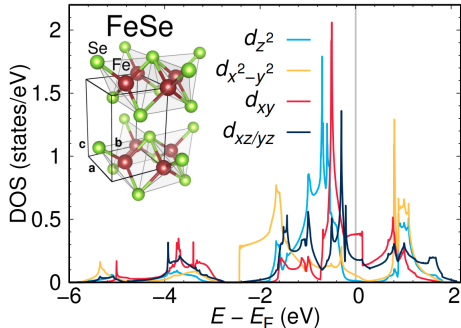
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Density of states and Matsubara Green's function

- ▶ Electronic density of states describes behavior of material, e.g. spectroscopy
- ▶ Start with $\rho(E)$ from effective non-interacting theory
- ▶ Matsubara Green's function
$$G(i\omega_n) = \int_{-\infty}^{\infty} dE \frac{\rho(E)}{i\omega_n - E}$$
with $\omega_n = (2n + 1)\pi/\beta$, $n \in \mathbb{Z}$
- ▶ Calculate interacting $G_{\text{int}}(i\omega_n)$ using $G(i\omega_n)$ and Feynman diagrams
- ▶ Compare $\rho_{\text{int}}(E)$ to experiments
- ▶ How to obtain $\rho_{\text{int}}(E)$ from $G_{\text{int}}(i\omega_n)$?

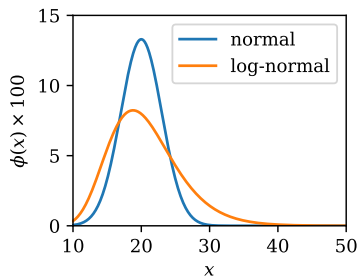
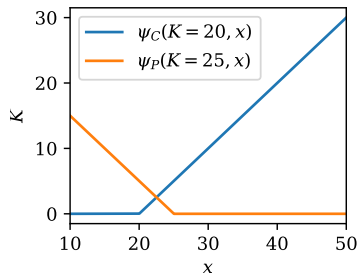


Terminal density and plain-vanilla option price

- Price for option with payoff $\psi(K, x)$ and known terminal density $\phi(x)$:

$$\Pr(K) = \int_{-\infty}^{\infty} dx \psi(K, x) \phi(x)$$

- Call option: $\psi_C(K, x) = \max(0, x - K)$
- Put option: $\psi_P(K, x) = \max(0, K - x)$
- Bachelier model: normal $\phi(x)$
- Black-Scholes model: log-normal $\phi(x)$
- Simple models do not match the market
- How to imply continuous $\phi(x)$ from discrete set of market prices $\Pr(K)$?

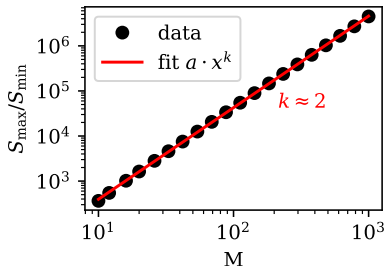


Linearization and singular value decomposition

- ▶ $\text{Pr}(\mathbf{K}) = \int_{-\infty}^{\infty} dx \psi(\mathbf{K}, x) \phi(x)$
- ▶ M known prices
- ▶ N equidistant points in x -direction
- ▶ Trapezoidal approximation for integral
- ▶ $g_i(x) = \Delta x \cdot \psi_{C/P}(\mathbf{K}_i, x)$

$$\text{Pr} = \begin{pmatrix} \text{Pr}_1 \\ \vdots \\ \text{Pr}_M \end{pmatrix} = \begin{pmatrix} \frac{1}{2}g_1(x_1) & g_1(x_2) & \dots & g_1(x_{N-1}) & \frac{1}{2}g_1(x_N) \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ \frac{1}{2}g_M(x_1) & g_M(x_2) & \dots & g_M(x_{N-1}) & \frac{1}{2}g_M(x_N) \end{pmatrix} \begin{pmatrix} \phi(x_1) \\ \vdots \\ \phi(x_N) \end{pmatrix} = \mathbf{G}\phi$$

- ▶ In general \mathbf{G} is $M \times N$ matrix, with $M \leq N$
- ▶ \mathbf{G} is ill-conditioned, cannot invert $\text{Pr} = \mathbf{G}\phi$ to get ϕ
- ▶ Analyze kernel with singular value decomposition: $\mathbf{G} = \mathbf{U}\mathbf{S}\mathbf{V}^T$
- ▶ Singular values decay quadratically, makes inversion unstable

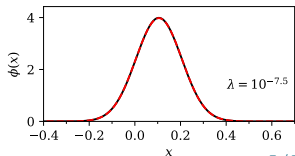
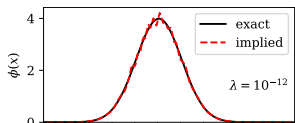
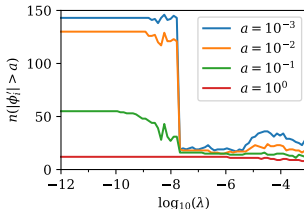
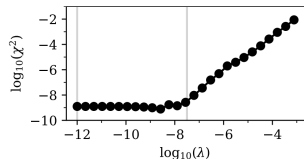


SVD, optimization, regularization

- ▶ SVD: $\text{Pr} = G\phi = \text{USV}^T\phi$
- ▶ $\text{Pr}' = \text{U}^T\text{Pr} = \text{SV}^T\phi = \text{S}\phi'$
- ▶ S is diagonal, $\text{Pr}'_i = \text{S}_{ii}\phi'_i = s_i\phi'_i$
- ▶ Keep **Q singular values**: $(\text{U}, \text{V}) \rightarrow (\tilde{\text{U}}, \tilde{\text{V}})$
- ▶ $\tilde{\text{U}}$ is $M \times Q$ with $Q \leq \min(M, N)$, approximate $G \approx \tilde{G} = \tilde{\text{U}}\tilde{\text{S}}\tilde{\text{V}}^T$
- ▶ **L_1 -regularized optimization problem**

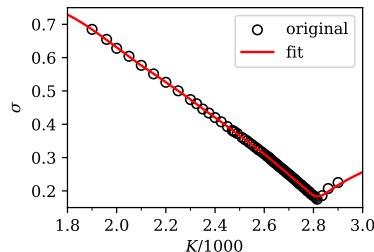
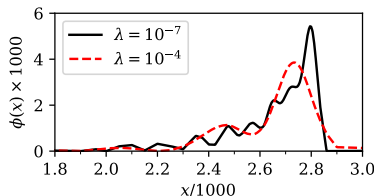
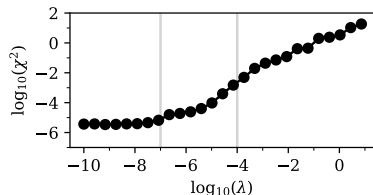
$$\text{F}(\phi'|\text{Pr}, \lambda) = \frac{1}{2}\|\text{Pr} - \tilde{\text{U}}\tilde{\text{S}}\phi'\|_2^2 + \lambda\|\phi'\|_1$$
- ▶ N discretization points for ϕ , but **optimize only $Q \leq \min(M, N)$ entries** of ϕ'
- ▶ additional conditions $\phi_i = (\tilde{\text{V}}\phi')_i \geq 0 \quad \forall i$

$$\text{and } 1 = \left(\frac{1}{2}(\phi_1 + \phi_N) + \sum_{i=2}^{N-1} \phi_i \right) \Delta x$$



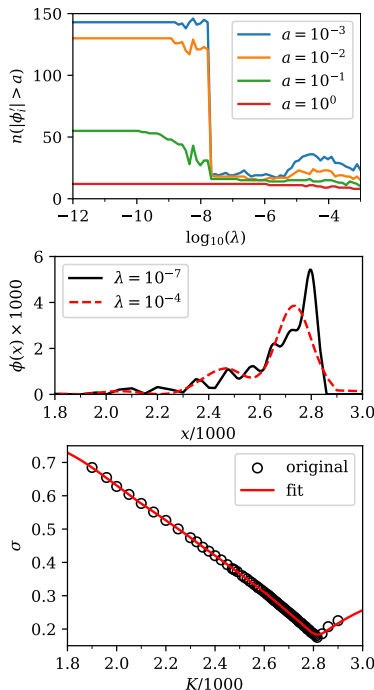
S&P 500 index options

- ▶ Perfect reproduction of normal (Bachelier), mixtures of normal and log-normal (Black-Scholes) models
- ▶ One-month S&P 500 index options on February 5th, 2018
- ▶ Calculate implied future distribution of stock index price
- ▶ Automatically correct implied volatility smiles
- ▶ Sensible inter- and extrapolation, currently no other general method available
- ▶ Performance easily tunable by choice of Q and N , execution time ~ 10 ms



Summary

- ▶ Want to find density $\phi(x)$ in integral
$$\Pr(K) = \int_{-\infty}^{\infty} dx \psi(K, x) \phi(x)$$
- ▶ Approximate integral using trapezoidal rule, write as matrix equation
- ▶ Perform SVD of kernel matrix
- ▶ Discard small singular values
- ▶ Reformulate as optimization problem in SVD-transformed domain
- ▶ Apply L_1 -regularization to parameters
- ▶ General recipe for ill-conditioned inverse problems of this form
- ▶ Solution to the problem of volatility smile interpolation and extrapolation



References

- ▶ Otsuki *et al.*, J. Phys. Soc. Japan **89**, 012001 (2020)
- ▶ Guterding & Jeschke, Comp. Phys. Commun. **231**, 114 (2018)
- ▶ Guterding, arXiv:2205.10865