

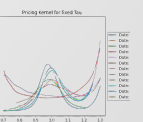
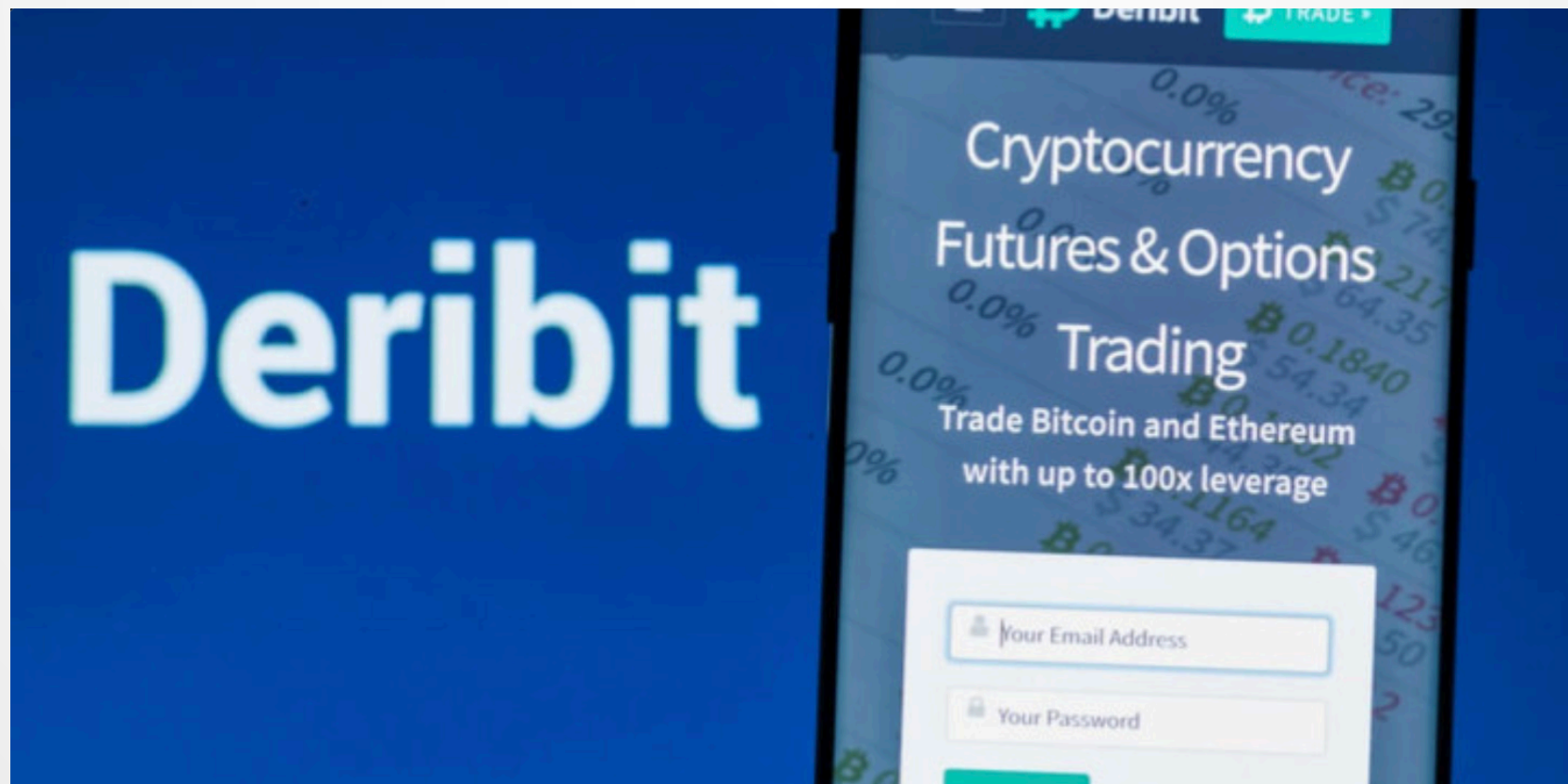
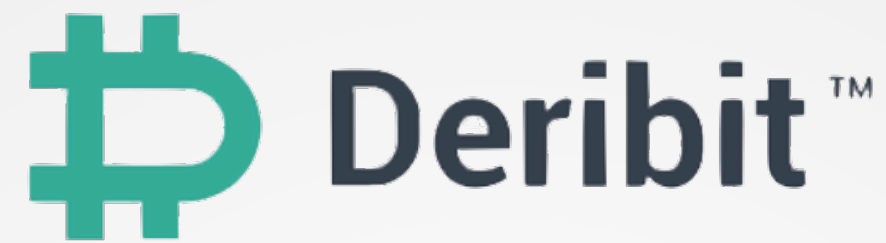


Bitcoin Pricing Kernels

Julian Winkel

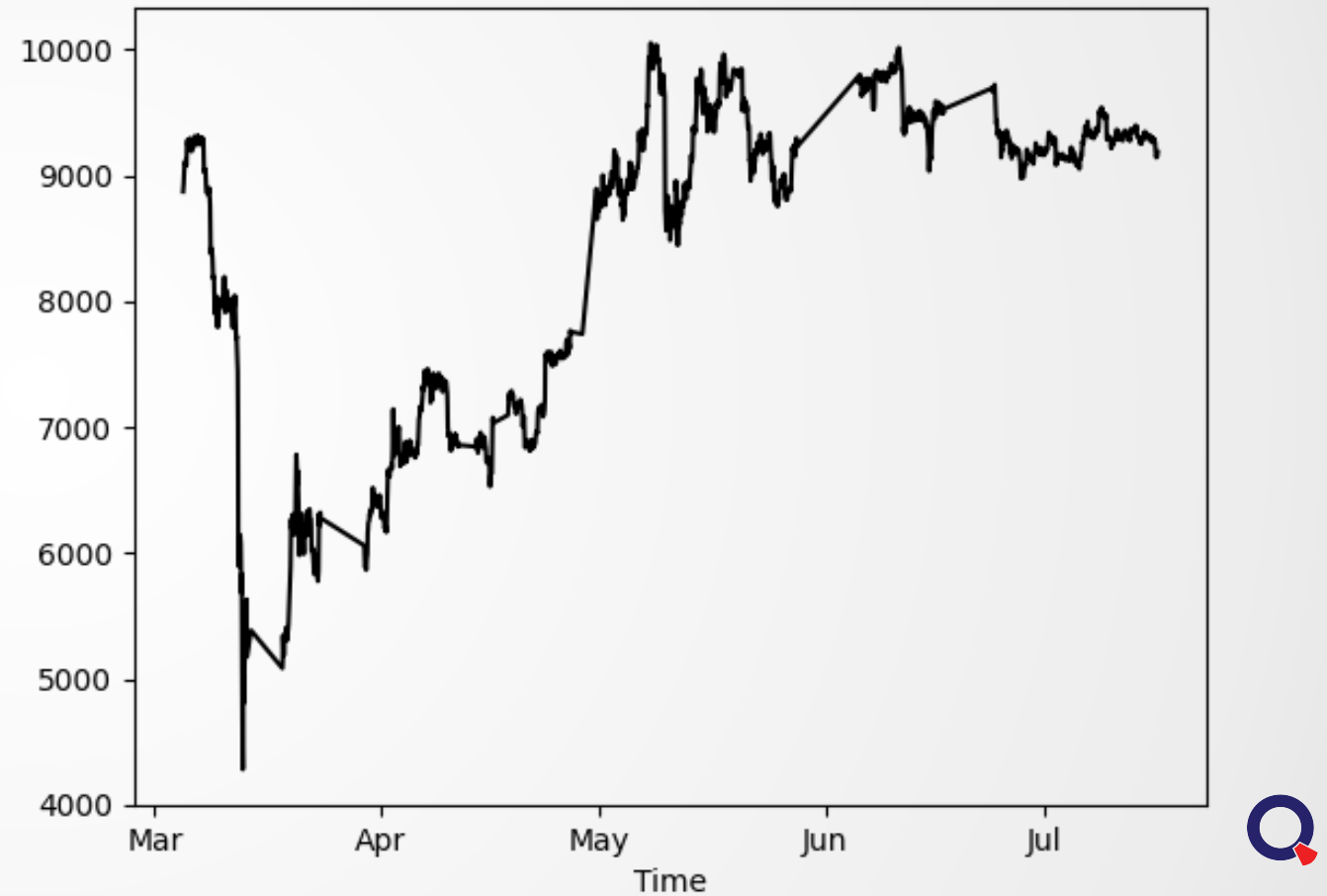
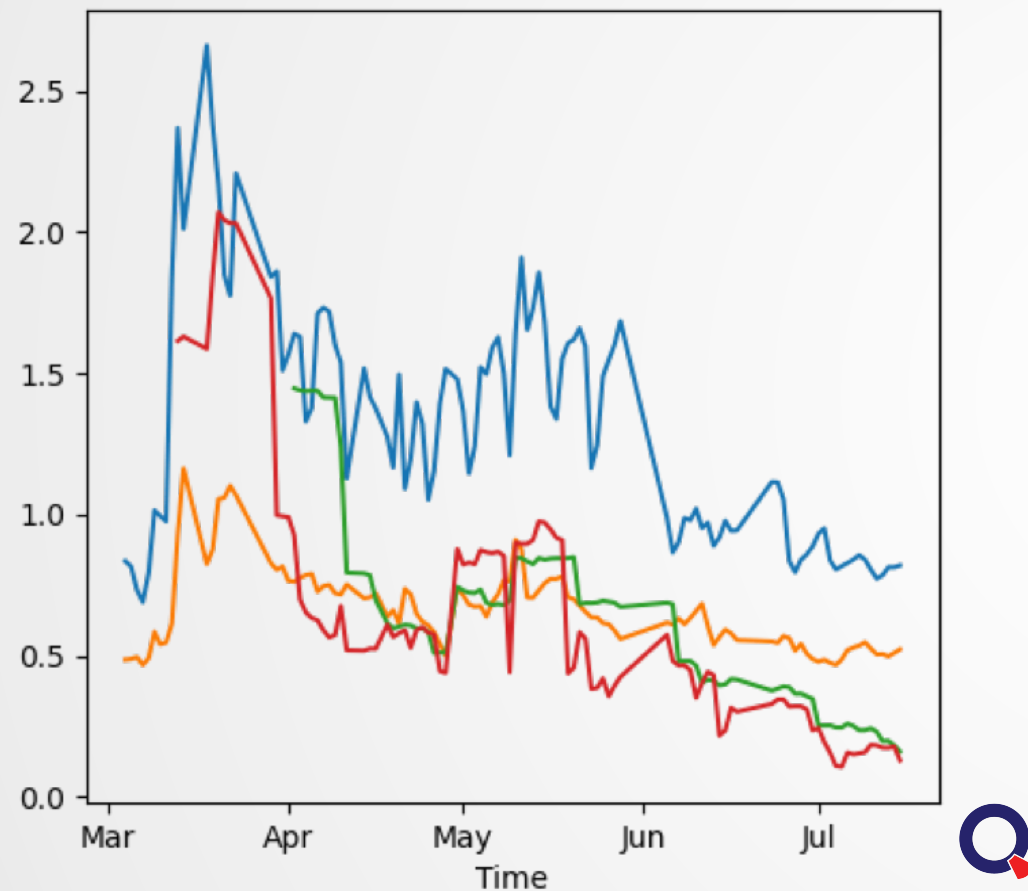
Ladislaus von Bortkiewicz Professor of Statistics
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BRC Blockchain Research Center
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Charles University, WISE XMU, NCTU 玉山学者



Volatile Times...

Is the market accurately pricing volatility?



Bid IV and Ask IV across all instruments and realized volatility on **daily** base and **21 day rolling window**. Deribit's Underlying, a Synthetic BTC USD Index.

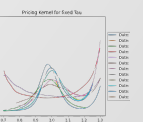
The Corona Shock

- Deribit Insurance Fund grows through liquidation orders
- Interesting data has been captured for the BRC
- Corona Shock caused spike in Option Writer defaults



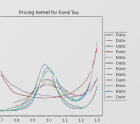
Date	Bankruptcies	Insurance Balance
21 Mar 2020	28	124.18843
20 Mar 2020	90	122.13254
19 Mar 2020	53	117.49559
18 Mar 2020	44	115.75211
17 Mar 2020	101	113.70538
16 Mar 2020	38	111.83231
15 Mar 2020	9	109.75960
14 Mar 2020	40	112.11395
13 Mar 2020	391	197.60795
12 Mar 2020	1273	211.77707

Deribit Insurance Fund denoted in Bitcoin



Outline

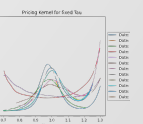
1. Motivation
2. Deribit
3. Pricing Kernels
4. State Price Densities
5. Physical Densities
6. Shape Invariant Models
7. Trading
8. Literature



Deribit

- www.deribit.com
- Largest Bitcoin Derivatives Exchange
- Controls around 80% of the BTC Derivatives Volume

Calls			Underlying: SYN.BTC-14DEC20(\$19192.28) 14 Dec 2020 Expires In 13 hours and 3 minutes							Puts								
Last	Size	IV	Bid	Ask	IV	Size	Vol	Δ Delta	Strike	Last	Size	IV	Bid	Ask	IV	Size	Vol	Δ Delta
-	-	-	-	-	-	-	-	1.00	16750	0.0005	-	-	-	0.0005	173.9%	17.5	5.0	0.00
-	-	-	-	-	-	-	-	1.00	17000	0.0005	-	-	-	0.0005	157.5%	0.6	47.0	0.00
-	-	-	-	-	-	-	-	1.00	17250	0.0005	-	-	-	0.0005	141.2%	1.5	33.0	0.00
-	-	-	-	-	-	-	-	1.00	17500	0.0005	-	-	-	0.0005	124.9%	1.0	77.6	0.00
-	-	-	-	-	-	-	-	1.00	17750	0.0005	-	-	-	0.0005	108.6%	0.5	47.8	0.00
-	-	-	-	-	-	-	-	0.99	18000	0.0005	-	-	-	0.0010	105.4%	0.6	55.5	-0.01
0.0185	5.0	0.0%	0.0290 \$556.58	0.0540 \$1036.38	139.2%	5.0	-	0.97	18250	0.0005	-	-	-	0.0010 \$19.19	87.3%	1.0	136.4	-0.03
0.0110	5.0	0.0%	0.0325 \$623.54	0.0440 \$844.18	142.7%	5.0	-	0.93	18500	0.0010	2.9	58.5%	0.0005 \$9.59	0.0010 \$19.19	68.5%	0.4	354.4	-0.07
0.0275	2.0	0.0%	0.0220 \$422.25	0.0280 \$537.41	88.2%	2.0	19.0	0.85	18750	0.0020	0.9	49.0%	0.0010 \$19.19	0.0025 \$47.97	66.0%	70.5	329.7	-0.15
0.0200	2.0	40.9%	0.0125 \$239.90	0.0155 \$297.48	63.2%	2.0	62.9	0.69	19000	0.0045	2.0	44.4%	0.0030 \$57.56	0.0045 \$86.35	55.6%	29.2	67.2	-0.31
0.0080	2.0	48.2%	0.0060 \$115.15	0.0075 \$143.94	58.0%	5.0	212.5	0.45	19250	0.0085	19.7	37.9%	0.0075 \$143.94	0.0105 \$201.51	57.7%	24.0	35.3	-0.55
0.0025	27.4	52.7%	0.0025 \$47.97	0.0035 \$67.16	61.2%	5.4	99.3	0.24	19500	0.0190	11.0	0.0%	0.0150 \$287.87	0.0225 \$431.81	83.1%	11.0	1.4	-0.76
0.0020	41.6	57.1%	0.0010 \$19.19	0.0020 \$38.38	70.0%	6.4	29.7	0.12	19750	-	0.1	0.0%	0.0215 \$412.49	-	-	-	-	-0.88
0.0025	14.6	64.5%	0.0005 \$9.59	0.0015 \$28.78	83.1%	40.2	59.9	0.08	20000	-	-	-	-	-	-	-	-	-0.92
0.0020	-	-	-	0.0015 \$28.78	100.4%	70.5	13.6	0.03	20250	-	-	-	-	-	-	-	-	-0.97
0.0015	-	-	-	0.0010 \$19.18	107.7%	40.5	24.1	0.01	20500	-	-	-	-	-	-	-	-	-0.99

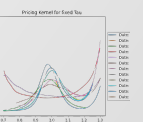


Deribit Data

- Every executed trade and order book change is saved on the BRC server
- Access with GUI via hu-berlin.de/brc
- Receive transactions via email

Exchange	Deribit	▼
Start date	01.04.2019 00:00:00	
End date	31.10.2020 23:59:59	
Frequency	Seconds	▼
Side	Buy	▼

REQUEST DATA



Pricing Kernels

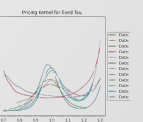
The value of a contingent claim at time t with maturity T for underlying S is

$$C_t = e^{-r\tau} \mathbb{E}_Q[\psi(S_T)] = e^{-r\tau} \int_{-\infty}^{\infty} \psi(S_T) q(S_T) dS_T$$

With option payoff $\Psi(S_T) = \max\{S_T - K, 0\}$, riskless rate r .

Risk-neutral pricing according to arbitrage theory

$$C_t = e^{-r\tau} \int_{-\infty}^{\infty} \psi(S_T) q(S_T) dS_T = e^{-r\tau} \int_{-\infty}^{\infty} \psi(S_T) p(S_T) \underbrace{\frac{q(S_T)}{p(S_T)}}_{Z(S_T)} dS_T$$

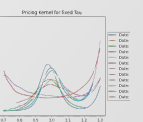


State Price Densities

- ▣ Derive the whole risk-neutral probability distribution of the Bitcoin price at maturity
- ▣ Use Rookley's method to derive SPDs
- ▣ Arbitrage-free option pricing for various instruments

$$C_t = e^{-r\tau} \mathbb{E}^Q[\psi(S_T)] = e^{-r\tau} \int_{-\infty}^{\infty} \psi(S_T) f_t^Q(S_T) dS_T$$

$$\hat{f}(S_T) = e^{r\tau} \left. \frac{\delta^2 C_t}{\delta K^2} \right|_{K=S_t}$$



Rookley's Method

- ▣ Estimate IV surface using time to maturity and moneyness
- ▣ Use IV estimates to calculate call prices
- ▣ Second derivative of call prices enables identification of state price densities

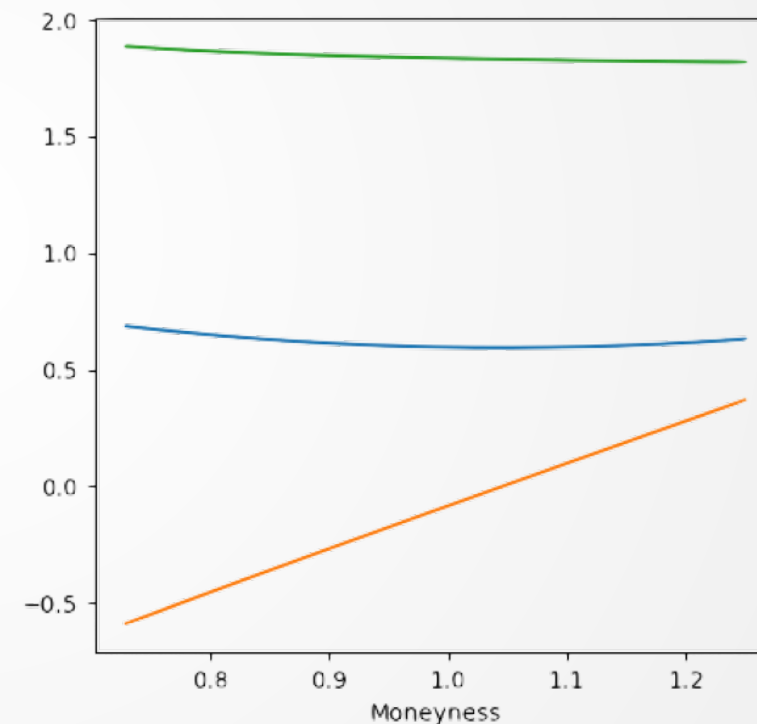
IV Decomposition

$$\sigma^{IV} = g(m, \tau) + \sigma(m, \tau)\varepsilon$$

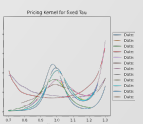
$$g(m, \tau) = g(m_0, \tau_0) + \frac{\partial g}{\partial M}(m - m_0) + \frac{1}{2} \frac{\partial^2 g}{\partial M^2}(m - m_0)^2$$

$$+ \frac{\partial g}{\partial \tau}(\tau - \tau_0) + \frac{1}{2} \frac{\partial^2 g}{\partial \tau^2}(\tau - \tau_0)^2$$

$$+ \frac{\partial^2 g}{\partial M \partial \tau}(m - m_0)(\tau - \tau_0)$$

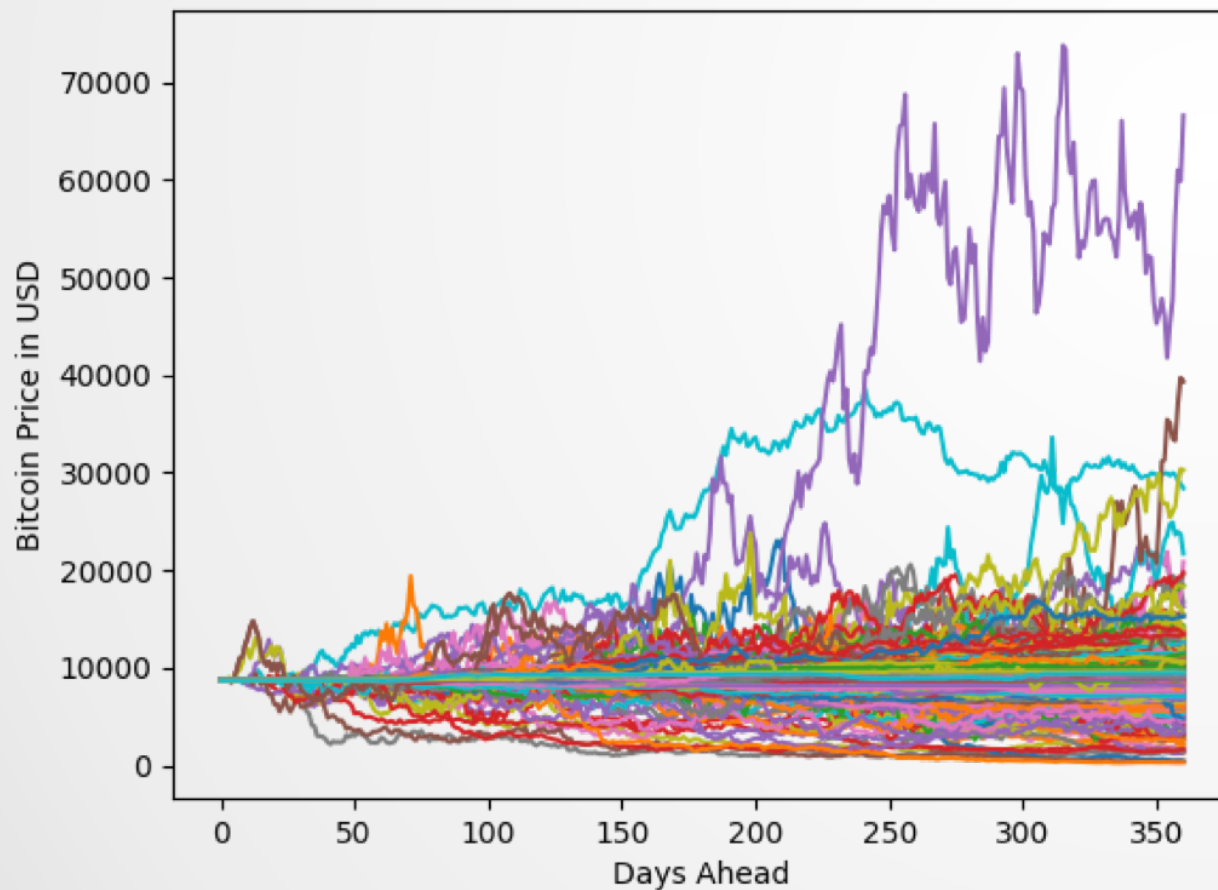


Volatility Smile,
First and Second Derivative



Physical Density

- ▣ Historical Density
- ▣ Dynamics of underlying assumed to follow SVCJ
- ▣ Alternatives: SV, SVJ, GARCH



$$\frac{dS_t}{S_t} = \mu dt \sqrt{V_t} W_t^{(s)} + Z_t^{(y)} dN_t$$

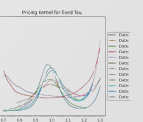
$$dV_t = \kappa(\theta - V_t)dt + \sqrt{V_t} W_t^{(v)} + Z^{(v)} dN_t$$

$$Cov(W^{(t)}, W^{(s)}) = \rho dt$$

$$P(dN_t = 1) = \lambda dt$$

	Mean	Standard Deviation
$\alpha = \kappa\tau$	α 0.0170	0.0040
$\beta = 1 - \kappa$	β -0.0570	0.0120
Jump Arrival Rate	λ 0.0150	0.0060
Log Return w/o Jumps	μ 0.0350	0.0140
	μ_v 1.4790	0.3860
	μ_y -0.0020	0.0430
Leverage Effect	ρ 0.0010	0.0340
	ρ_j -0.0000	0.0200
Volatility of Volatility	σ_v 0.0230	0.0050
	σ_y 0.4100	0.0540

SVCJ Simulation from 2020-03-01



Shape Invariant Models

Goal: Collapse different pricing kernels into a single population curve

$$Y_{tj} = K_t(u_j) + \varepsilon_t$$

$$K_t(u_j) = \theta_{t1} K_0\left(\frac{u - \theta_{t3}}{\theta_{t2}}\right) + \theta_{t4}$$

Initialization:

- 1) Estimate individual regression functions using Nadaraya-Watson-Estimator
- 2) Set starting values for Thetas for each point in time
- 3) Estimate average Kernel / Reference Kernel

$$K_0^{(0)}(u) = T^{-1} \sum_{t=1}^T \hat{K}_t(\theta_{t2}^{(0)} u + \theta_{t3}^{(0)})$$

Loop:

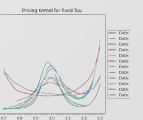
$$\arg \min_{\theta_{tj}} \int_{\mathbb{R}} \{ \hat{K}_t(\theta_{t2}^{(0)} u + \theta_{t3}^{(0)}) - \theta_{t1} K_0^{r-1}(u) - \theta_{t4} \}^2 w(u) du$$

Normalize Parameters

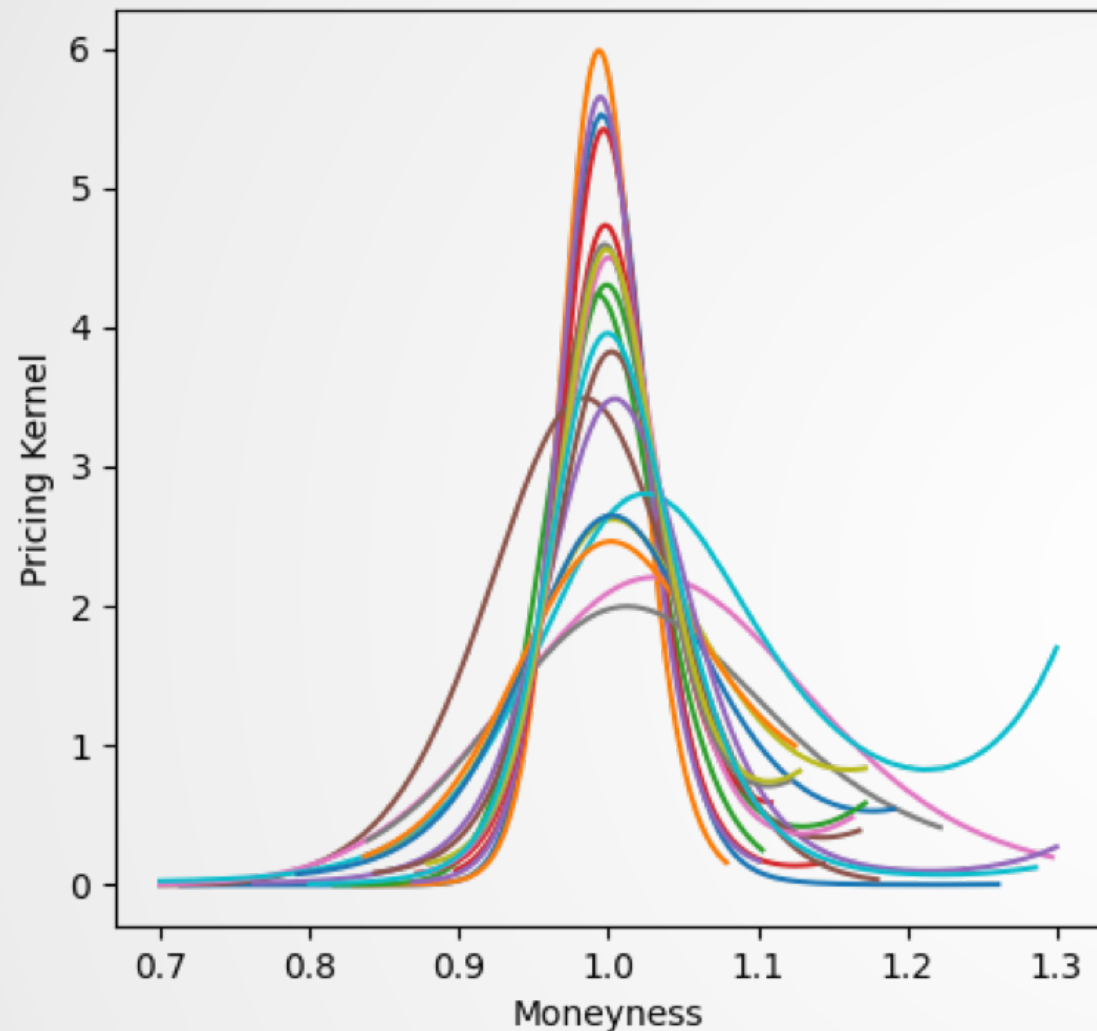
$$\theta_{tj}^r \leftarrow \frac{\theta_{tj}^r}{\sum_t \theta_{tj}^r} \quad \theta_{tj}^r \leftarrow \theta_{tj}^r - T^{-1} \sum_{t=1}^T \theta_{tj}^r$$

Update Reference Kernel

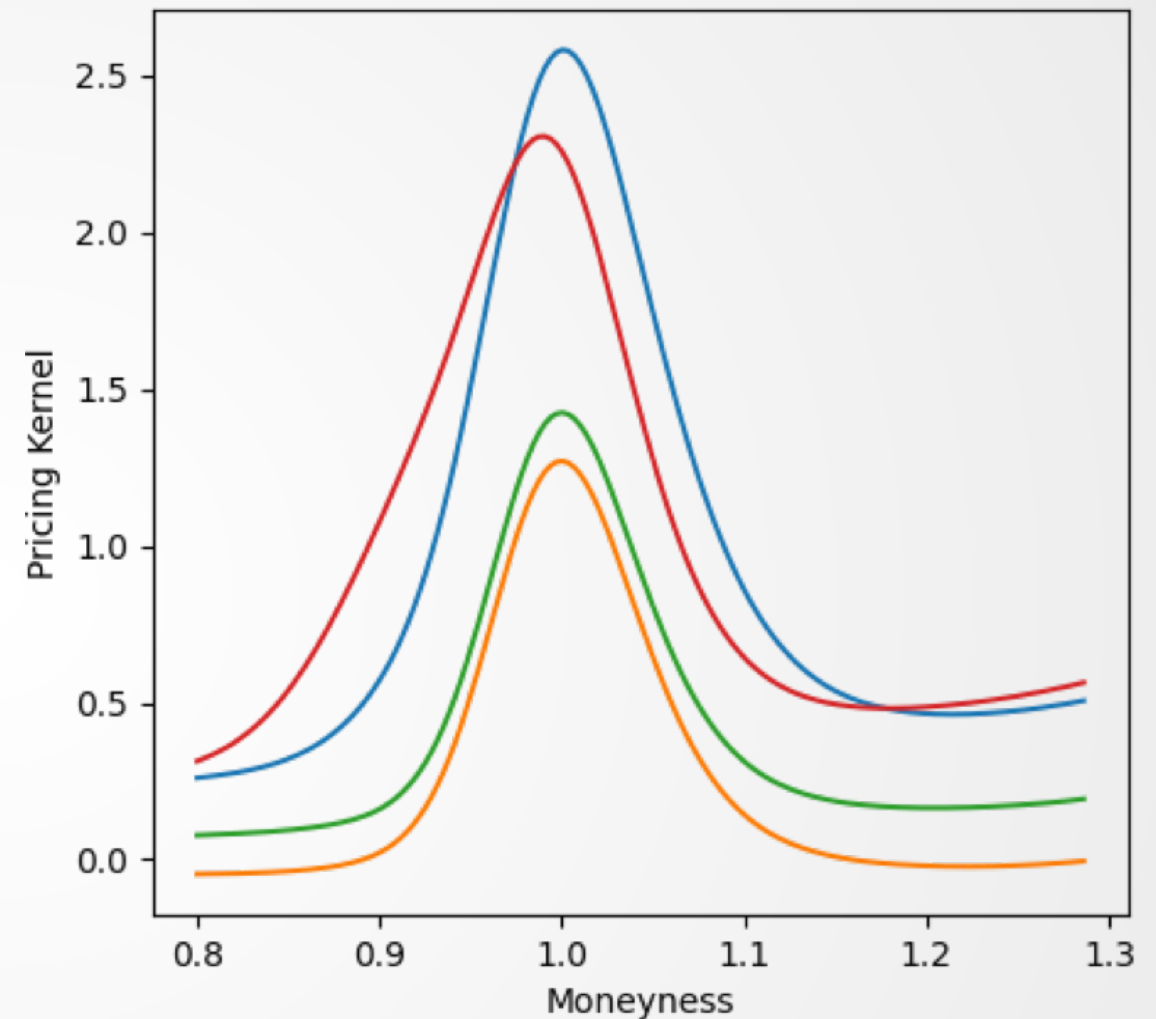
$$K_0^{(r)}(u) = T^{-1} \sum_{t=1}^T \hat{K}_t(\theta_{t2}^{(r)} u + \theta_{t3}^{(r)})$$



Shape Invariant Pricing Kernels

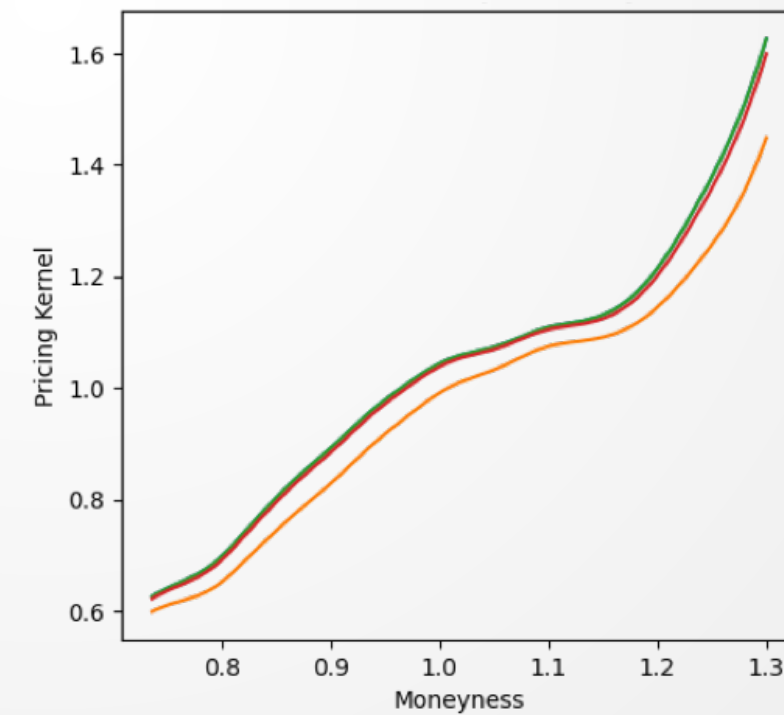
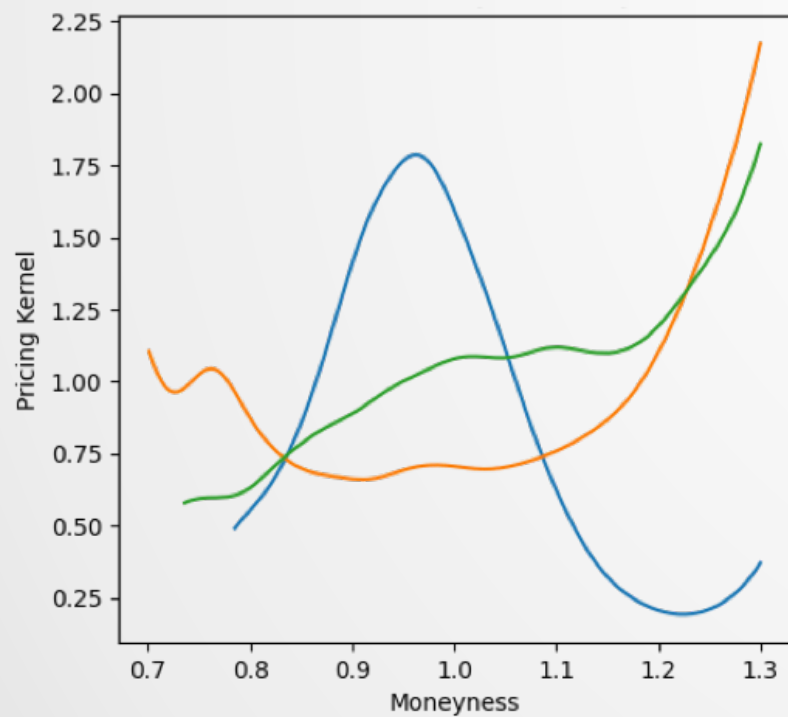
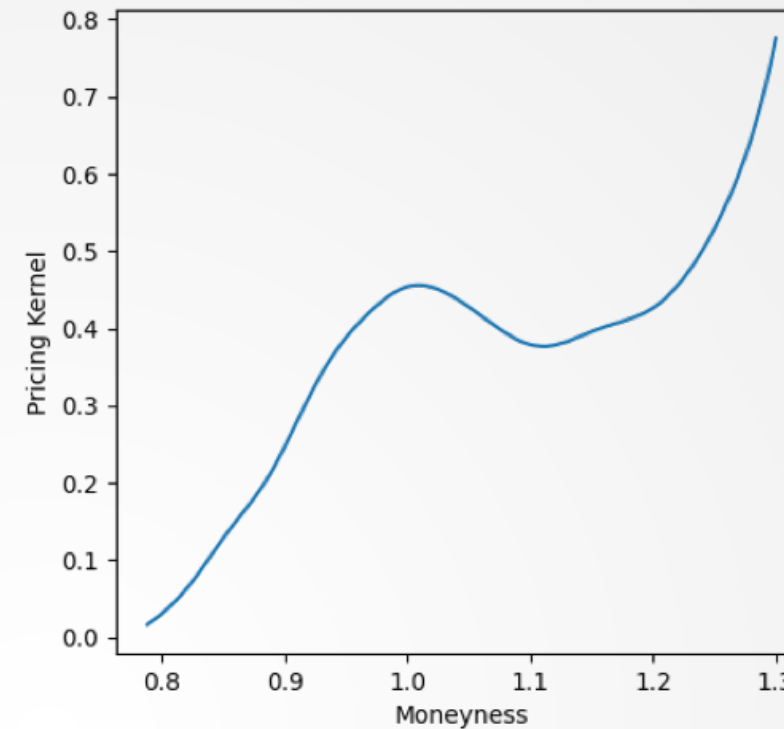
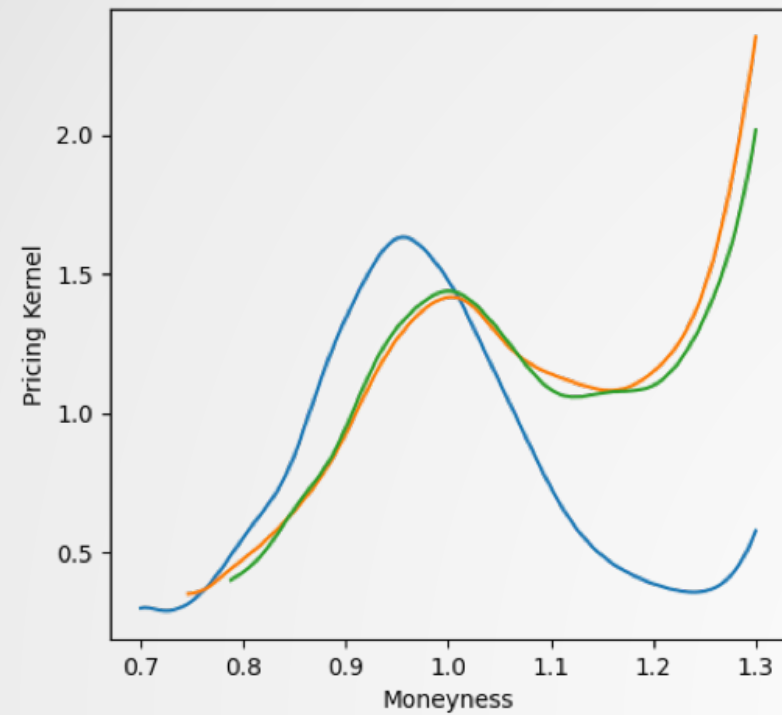


Pricing Kernels for Instruments with one day until maturity.



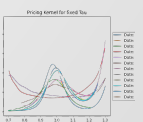
Shape Invariant Pricing Kernels across first, second, third and fourth iteration.

Pricing Kernels vs. Shape Invariant Pricing Kernels



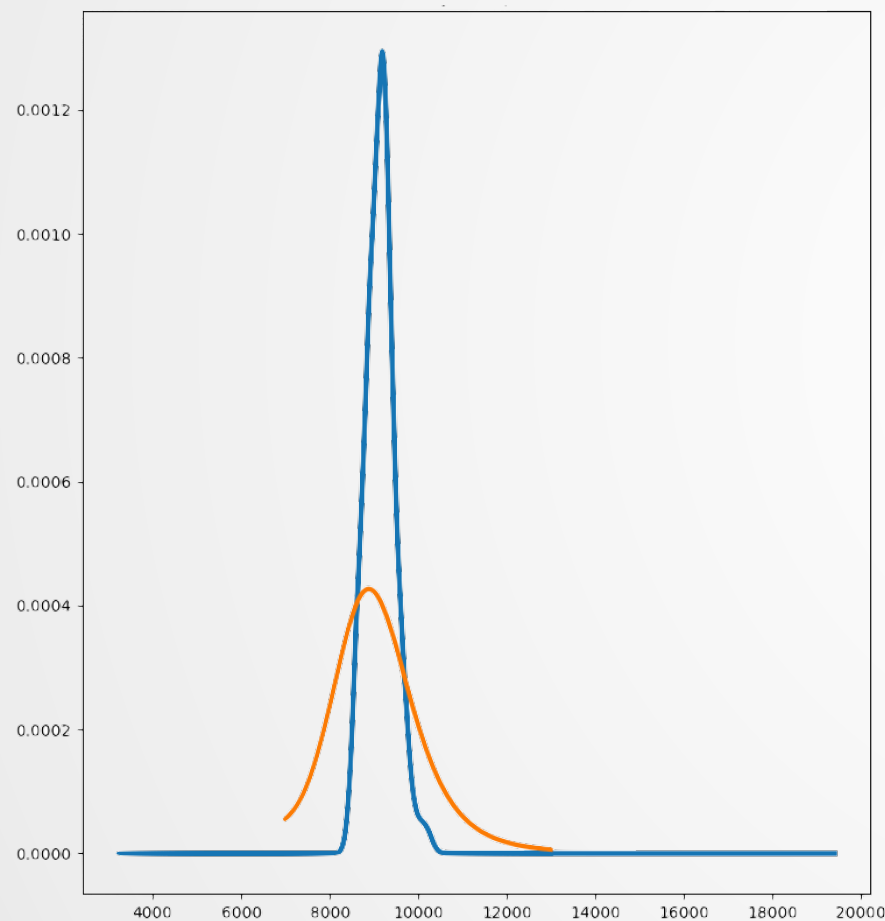
Maturity on [2020-03-20](#), [2020-04-10](#), [2020-04-17](#).

10 Days until Maturity in first row. 14 Days in second row.

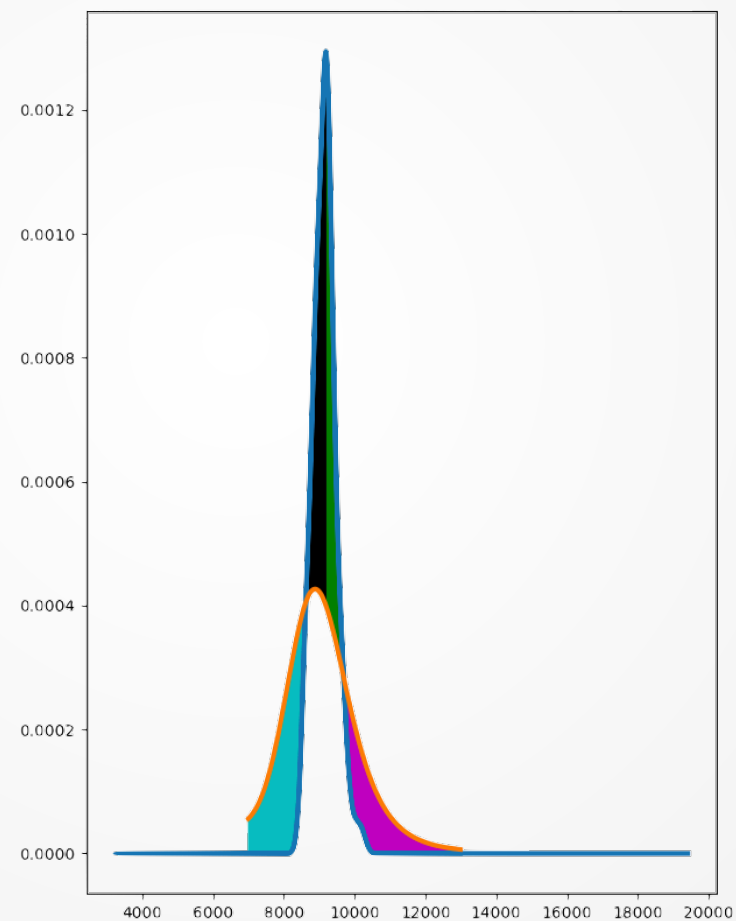


Trading on Density Deviations

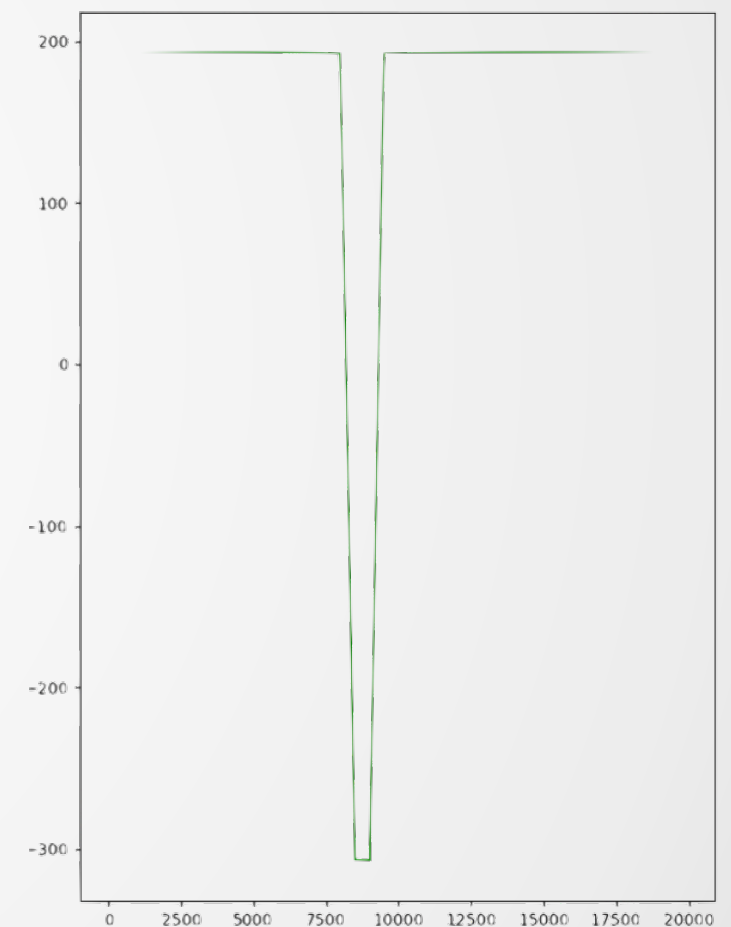
- ▣ Idea: If SPD deviates from PD, investors are mispricing certain events
- ▣ Correct mispricing by shorting overvalued contracts and longing undervalued ones
- ▣ Trade a call and a put spread and hold until maturity



State Price Density vs Physical Density.
Observed on 2020-03-06.



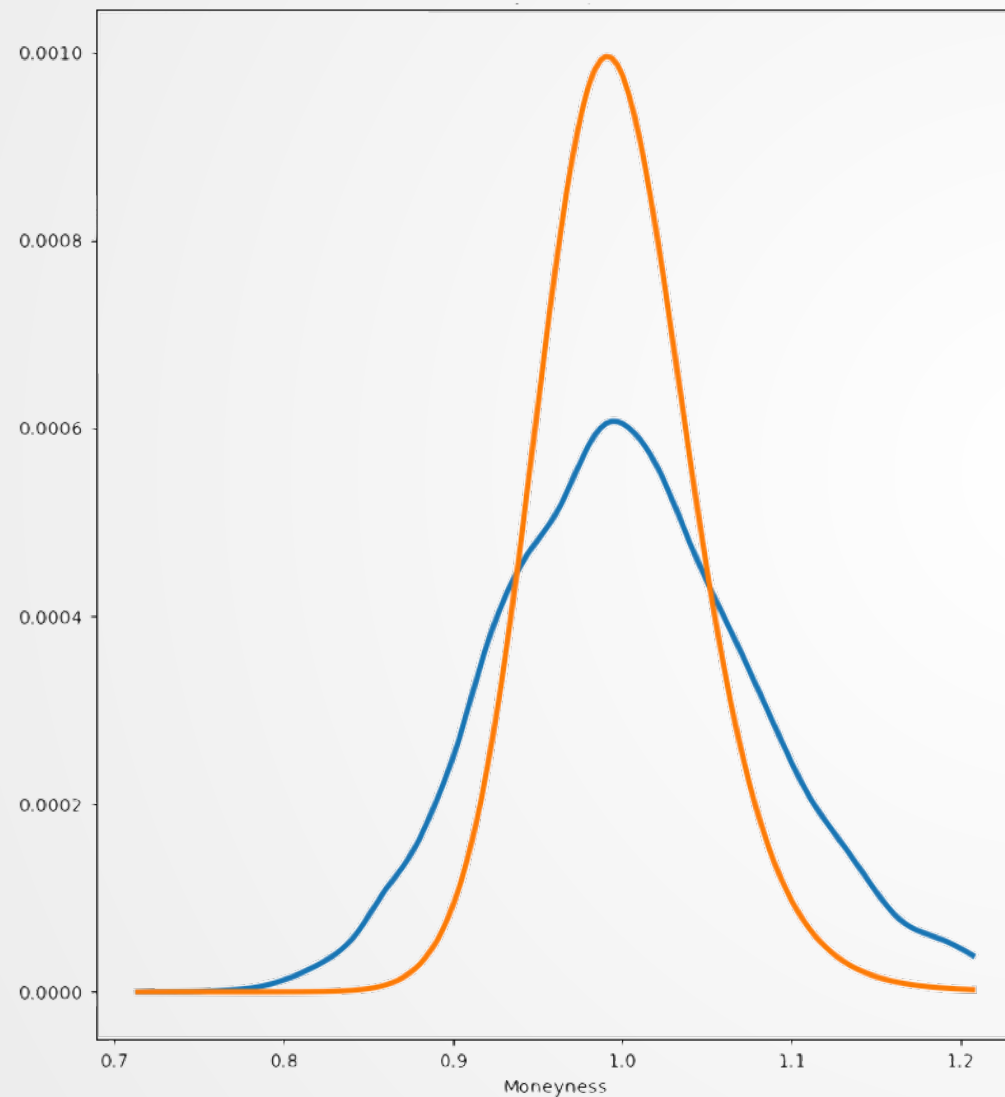
Trading Strategy
Long Put, Short Put,
Long Call, Short Call.



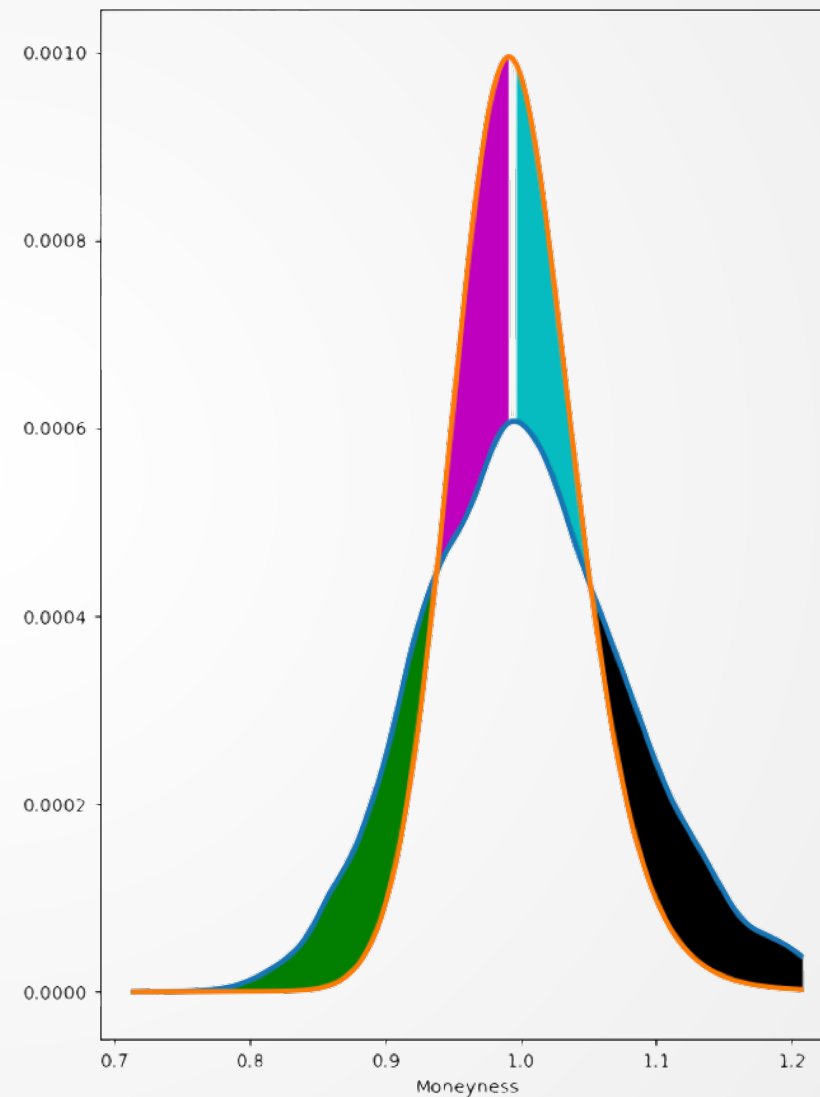
Portfolio Payoff

Pricing Kernel before the Corona Shock

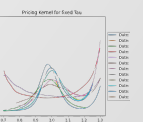
Compare pricing of 2-Days-until-Maturity instruments before and after the Corona Shock: Market was underestimating volatility.



State Price Density vs Physical Density.
Observed on 2020-03-04.

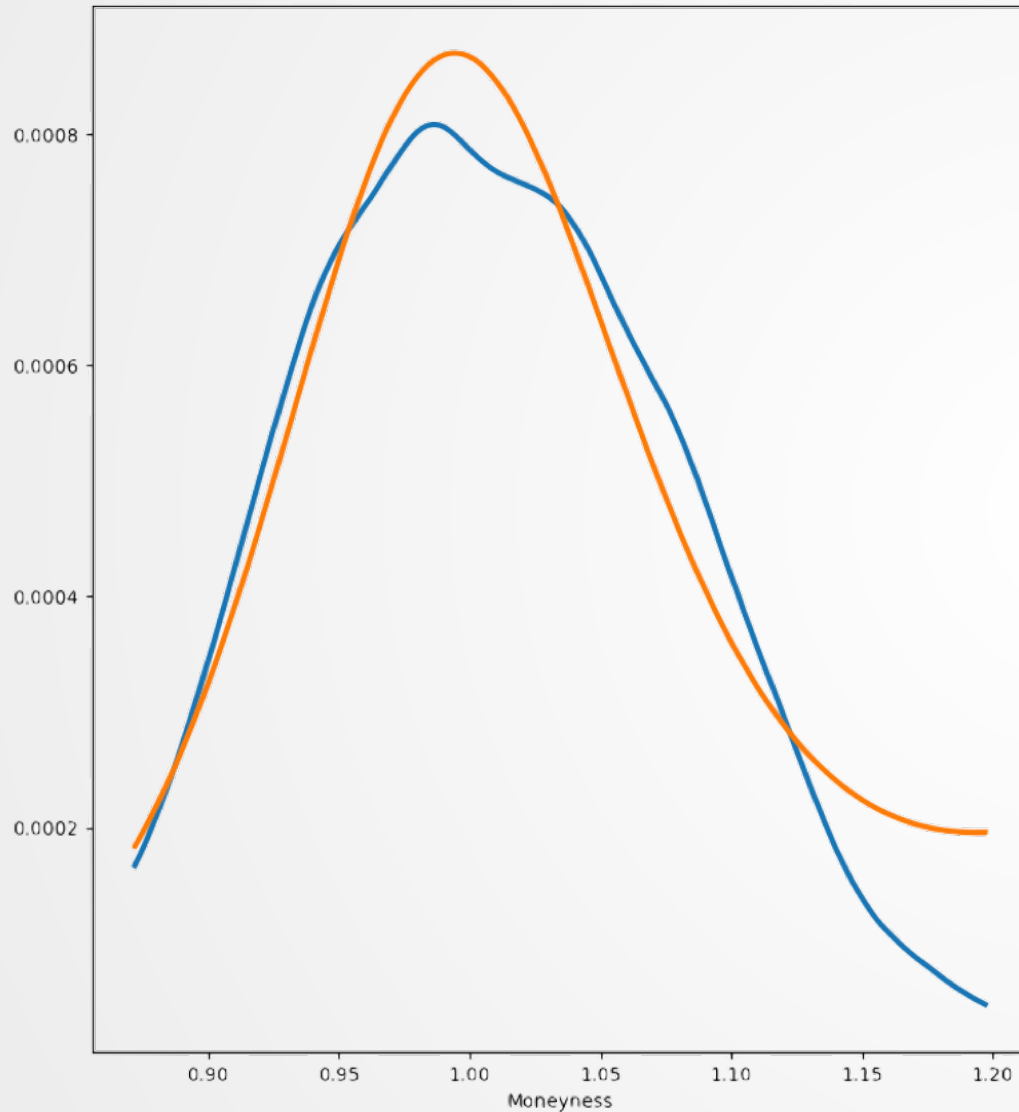


Trading Strategy
Long Put, Short Put, Long Call, Short Call.

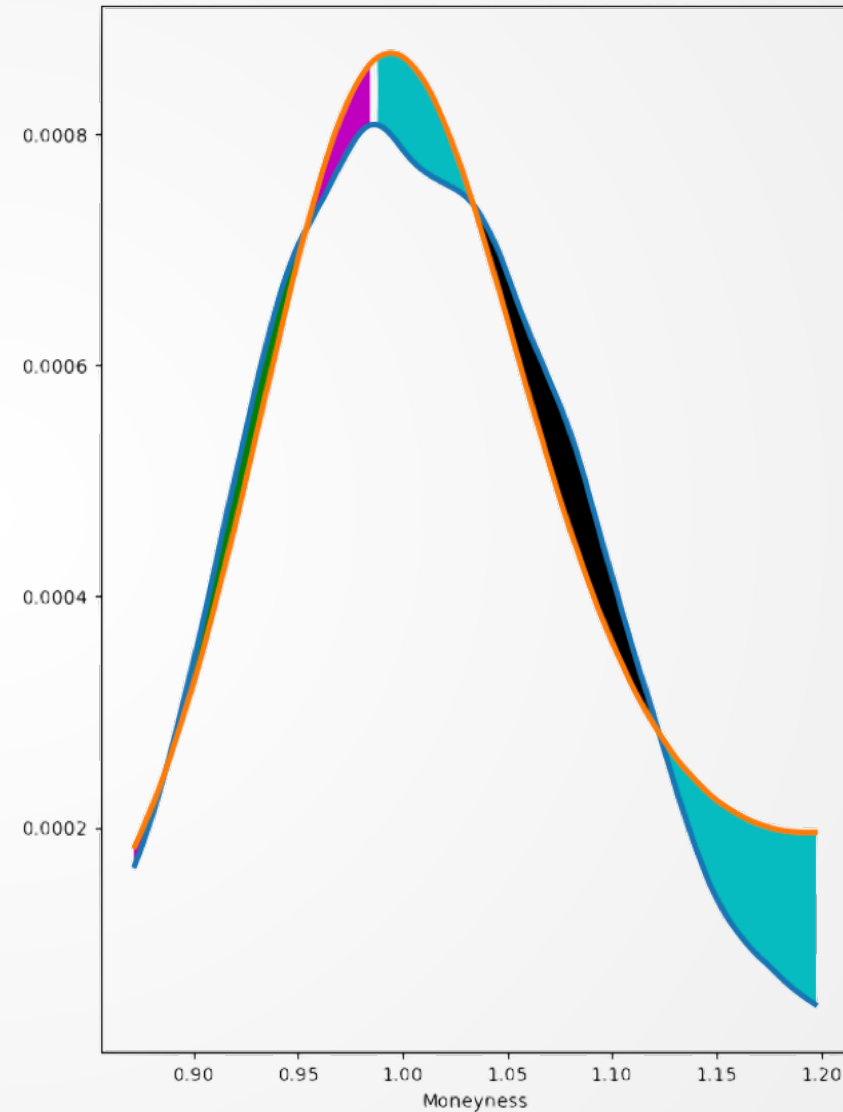


Pricing Kernels adjust quickly

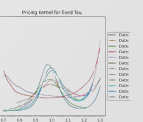
Market adjusted to our SVCJ simulation after the Corona Shock.



State Price Density vs Physical Density.
Observed on 2020-03-21.

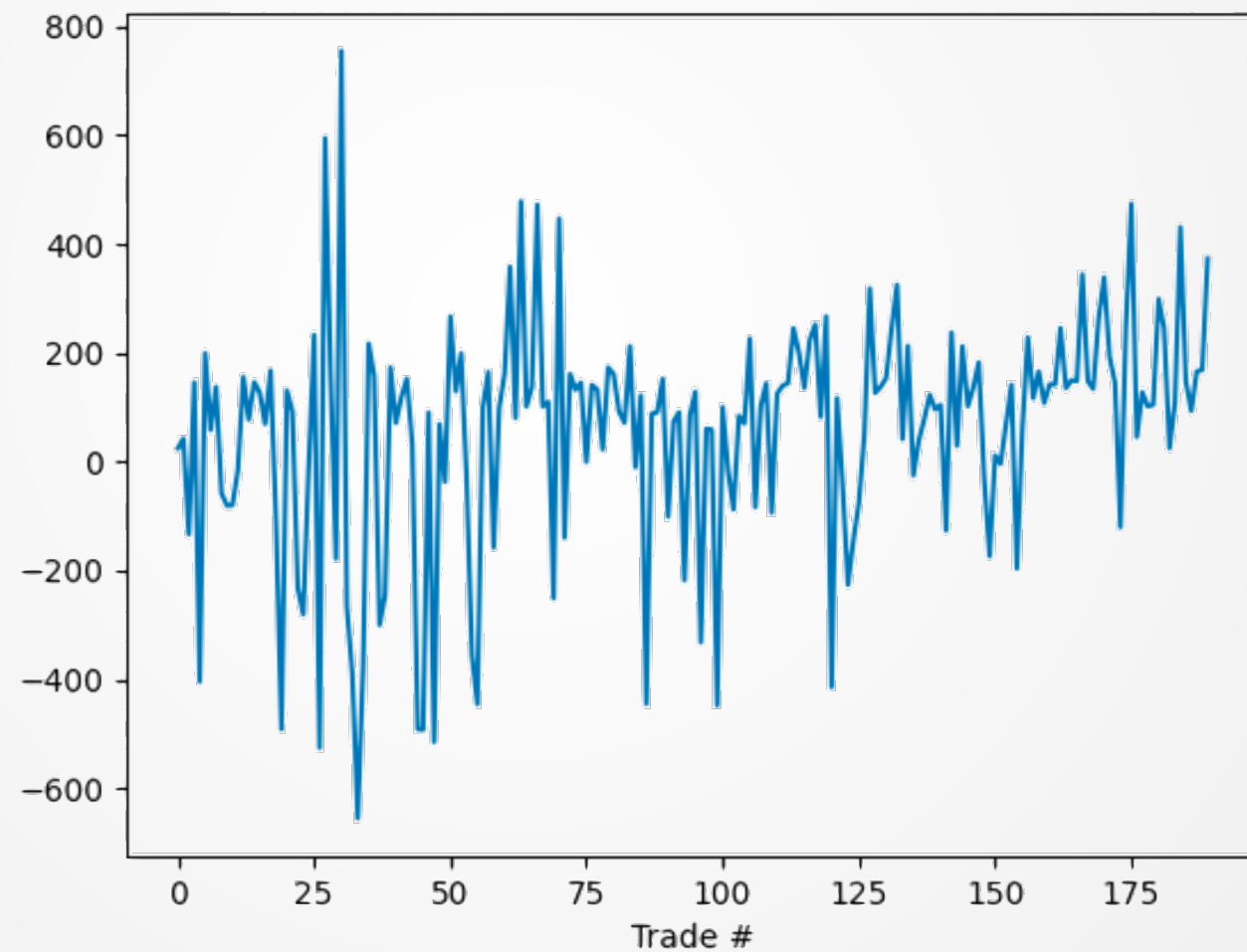


Trading Strategy
Long Put, Short Put, Long Call,
Short Call.

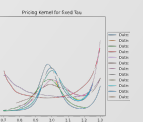


Trading Performance

- ▣ Trading Strategy is profitable
- ▣ Sharpe Ratio is only 0.22
- ▣ Too much directional risk



Daily Trading Returns



Literature

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