

# FE621 Homework 1 — Practice Solution Report

## Part 1 — Data Gathering

### 1. Data download program (Yahoo Finance)

### 2. DATA1 and DATA2 snapshots

This report contains computed results for DATA1. To complete DATA2, rerun the same script on the next consecutive trading day with tag DATA2.

tag	timestamp	asof_date	tsla_spot	spy_spot	vix_level	r_annual
DATA1	2026-02-15_20-27-20	2026-02-15	417.44	681.75	20.6	0.0364

### Why so many maturities exist

Exchanges list weekly, monthly, quarterly, and long-dated LEAPS expirations. Multiple maturities exist because different hedgers and speculators need exposure at different horizons. Market makers quote across maturities to provide liquidity along the volatility term structure.

### 3. Description of symbols

TSLA (Tesla, Inc.) represents a high-growth equity whose price dynamics are largely driven by firm-specific factors such as earnings announcements, delivery numbers, technological developments, and investor sentiment, which naturally leads to elevated realized and implied volatility. In contrast the SPY (SPDR S&P 500 ETF Trust) is an exchange-traded fund designed to track the performance of the S&P 500 index and therefore provides diversified exposure to hundreds of large U.S. companies, resulting in significantly lower volatility due to diversification across sectors and firms. Because SPY reflects broad market behavior rather than idiosyncratic company risk, its options are widely used by institutional investors for hedging and expressing macroeconomic views. The VIX (Cboe Volatility Index) complements this perspective by providing a market-implied measure of expected 30-day volatility derived from S&P 500 option prices. It is often interpreted as a gauge of overall market uncertainty or “fear.” In practice, the implied volatility of SPY options tends to track the VIX closely, since both are linked to aggregate market risk, whereas TSLA options exhibit substantially higher implied volatility and a more pronounced volatility smile due to concentrated, stock-specific uncertainty and the possibility of large price jumps. This distinction is clearly reflected in the empirical results, where TSLA displays much higher at-the-money implied volatility and steeper smiles compared to SPY. Together these three instruments illustrate the difference between idiosyncratic risk, systematic market risk and market-wide implied

volatility expectations. This highlights how option prices encode both company-level uncertainty and broader macroeconomic sentiment.

#### 4. Recorded items

## Part 2 — Analysis of the Data

### 5. Black–Scholes formulas

Call:  $C = S \cdot N(d1) - K \cdot e^{(-r\tau)} \cdot N(d2)$

Put:  $P = K \cdot e^{(-r\tau)} \cdot N(-d2) - S \cdot N(-d1)$

$d1 = [\ln(S/K) + (r + 0.5\sigma^2)\tau] / (\sigma\sqrt{\tau})$ ,  $d2 = d1 - \sigma\sqrt{\tau}$ .  $N(\cdot)$  is the standard normal CDF.

### 6. Implied volatility via Bisection (tolerance $1e-6$ )

Implied volatility solves  $f(\sigma) = \text{BS\_price}(\sigma) - \text{MidPrice} = 0$ .  $\text{MidPrice} = (\text{Bid} + \text{Ask}) / 2$  when both exist and  $\text{volume} > 0$ . ATM means strike  $K$  closest to  $S_0$ . Average IV near ATM used moneyness band  $S/K \in [0.95, 1.05]$ .

symbol	expiry	type	spot S	atm_strike	atm_iv_bisect	avg_iv_band_bisect	n_opts_used	n_in_band
SPY	2026-02-20	call	681.75	682	0.20248	0.198911	186	68
SPY	2026-02-20	put	681.75	682	0.198866	0.212719	194	68
SPY	2026-03-20	call	681.75	682	0.17345	0.17151	193	68
SPY	2026-03-20	put	681.75	682	0.182354	0.183538	202	68
SPY	2026-04-17	call	681.75	682	0.164241	0.161311	145	64
SPY	2026-04-17	put	681.75	682	0.179419	0.18168	150	65
TSLA	2026-	call	417.44	417.5	0.458069	0.460128	157	16

	02-20							
TSL A	2026-02-20	put	417.44	417.5	0.452839	0.454999	127	16
TSL A	2026-03-20	call	417.44	415	0.447146	0.446667	140	8
TSL A	2026-03-20	put	417.44	415	0.44369	0.443439	129	8
TSL A	2026-04-17	call	417.44	415	0.453661	0.45326	142	8
TSL A	2026-04-17	put	417.44	415	0.449674	0.44993	133	8

## 7. Newton method (using Vega) and speed comparison

Newton update:  $\sigma_{n+1} = \sigma_n - f(\sigma_n)/f'(\sigma_n)$ , with  $f(\sigma) = \text{Vega}$ .  $\text{Vega} = S \cdot \phi(d1) \cdot \sqrt{\tau}$ , where  $\phi$  is the normal PDF. Newton typically converges in far fewer iterations than bisection (often ~5–10 vs ~40–60 at  $1e-6$ ) but can be unstable if Vega is tiny or the initial guess is poor; bisection is guaranteed if a sign-change bracket exists.

## 8. Interpretation: TSLA vs SPY vs ^VIX; maturity and moneyness

On DATA1, TSLA ATM call IV is 0.458 (~45.8%), while SPY ATM call IV is 0.202 (~20.2%). The observed ^VIX level was 20.6. SPY implied vols are much closer to an index volatility benchmark than TSLA, which reflects large idiosyncratic stock risk.

Across strikes, implied volatility is not constant (smile/skew). Deep ITM/OTM options often have higher IV due to tail-risk and crash-protection demand. Across maturities, IV can change with the term structure: short-dated options may embed event risk more strongly, while longer maturities average risk over time.

## 9. Put–Call parity comparison

Put–Call parity (ignoring dividends) is  $C - P = S - K \cdot e^{(-r\tau)}$ . Parity-implied prices were compared to observed mids. Deviations arise from bid–ask spreads, stale/illiquid quotes,

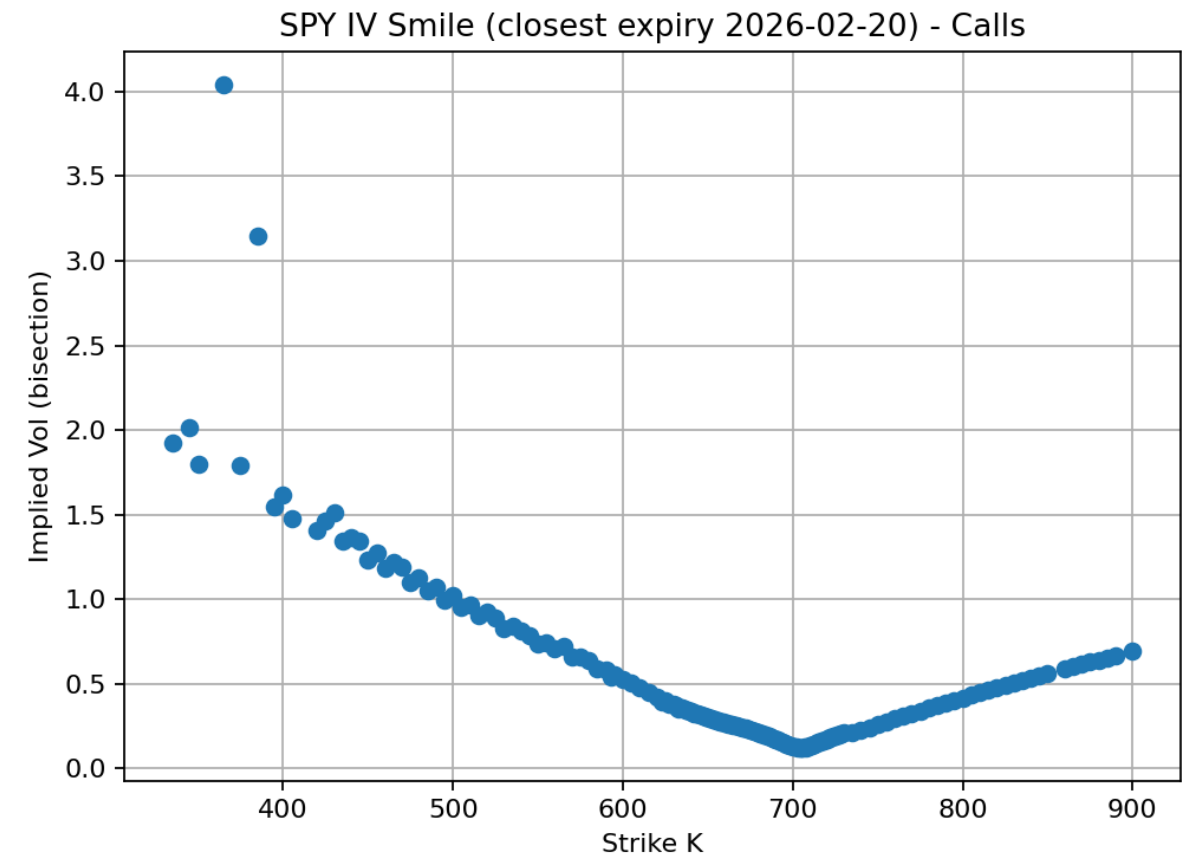
and dividends (SPY pays dividends; strict parity should adjust for PV(dividends) or use forwards).

Using mid prices, median  $|\text{parity error}| \approx 0.339$ , 95th percentile  $|\text{error}| \approx 16.110$ , and 99th percentile  $|\text{error}| \approx 19.085$ .

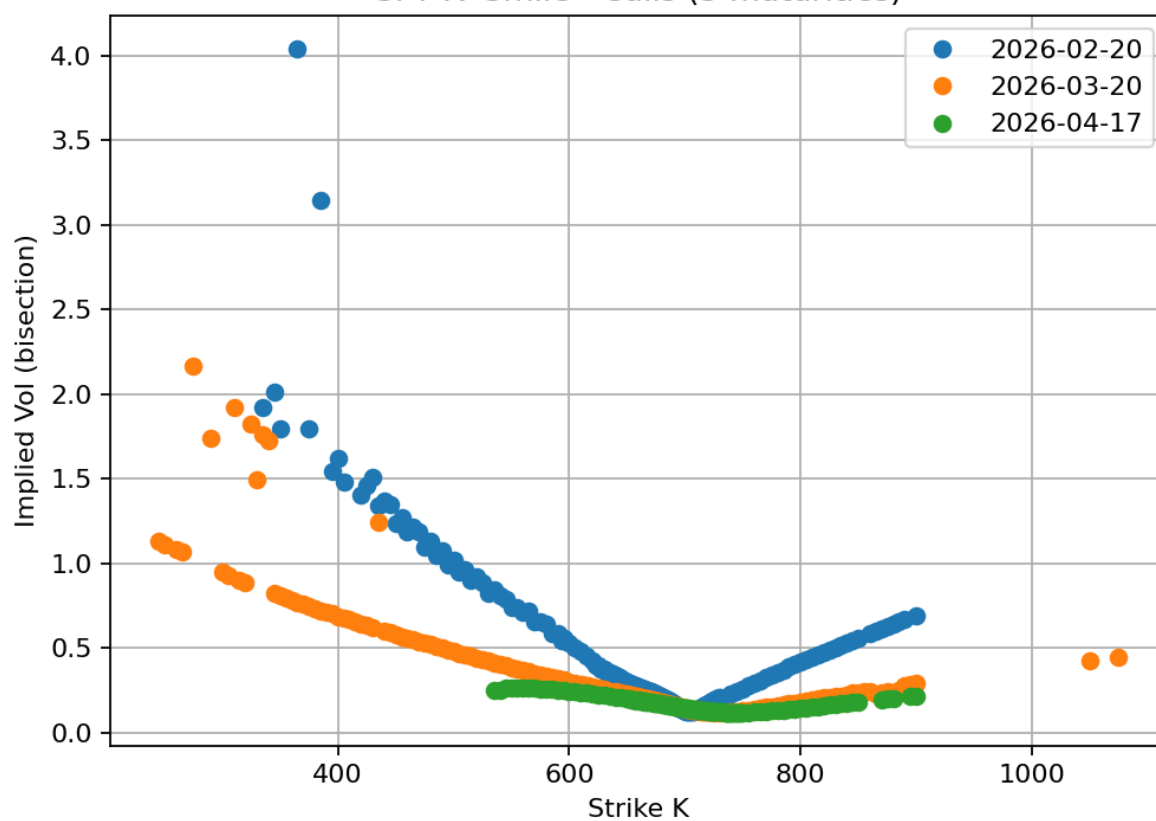
sym bol	exp iry	K	call	put	disc K	put_from call	call_fro m_put	call_parity error	put_parity error
TSL A	202 6- 02- 20	10 0	317. 575	0.0 05	99.9 501	0.085147	317.495	-0.080147	0.080147
TSL A	202 6- 02- 20	11 0	327	0.0 1	109. 945	19.5052	307.505	-19.4952	19.4952
TSL A	202 6- 02- 20	12 0	297. 55	0.0 05	119. 94	0.050176 9	297.505	- 0.0451769	0.0451769
TSL A	202 6- 02- 20	13 0	287. 55	0.0 1	129. 935	0.045191 8	287.515	- 0.0351918	0.0351918
TSL A	202 6- 02- 20	14 0	277. 575	0.0 1	139. 93	0.065206 7	277.52	- 0.0552067	0.0552067
TSL A	202 6- 02- 20	15 0	267. 575	0.0 1	149. 925	0.060221 7	267.525	- 0.0502217	0.0502217
TSL A	202 6- 02- 20	15 5	262. 575	0.0 1	154. 923	0.057729 2	262.527	- 0.0477292	0.0477292
TSL A	202 6- 02- 20	16 0	257. 575	0.0 05	159. 92	0.055236 6	257.525	- 0.0502366	0.0502366
TSL A	202 6- 02- 20	16 5	252. 6	0.0 1	164. 918	0.077744 1	252.532	- 0.0677441	0.0677441
TSL A	202 6-	17 0	247. 625	0.0 1	169. 915	0.100252	247.535	- 0.0902516	0.0902516

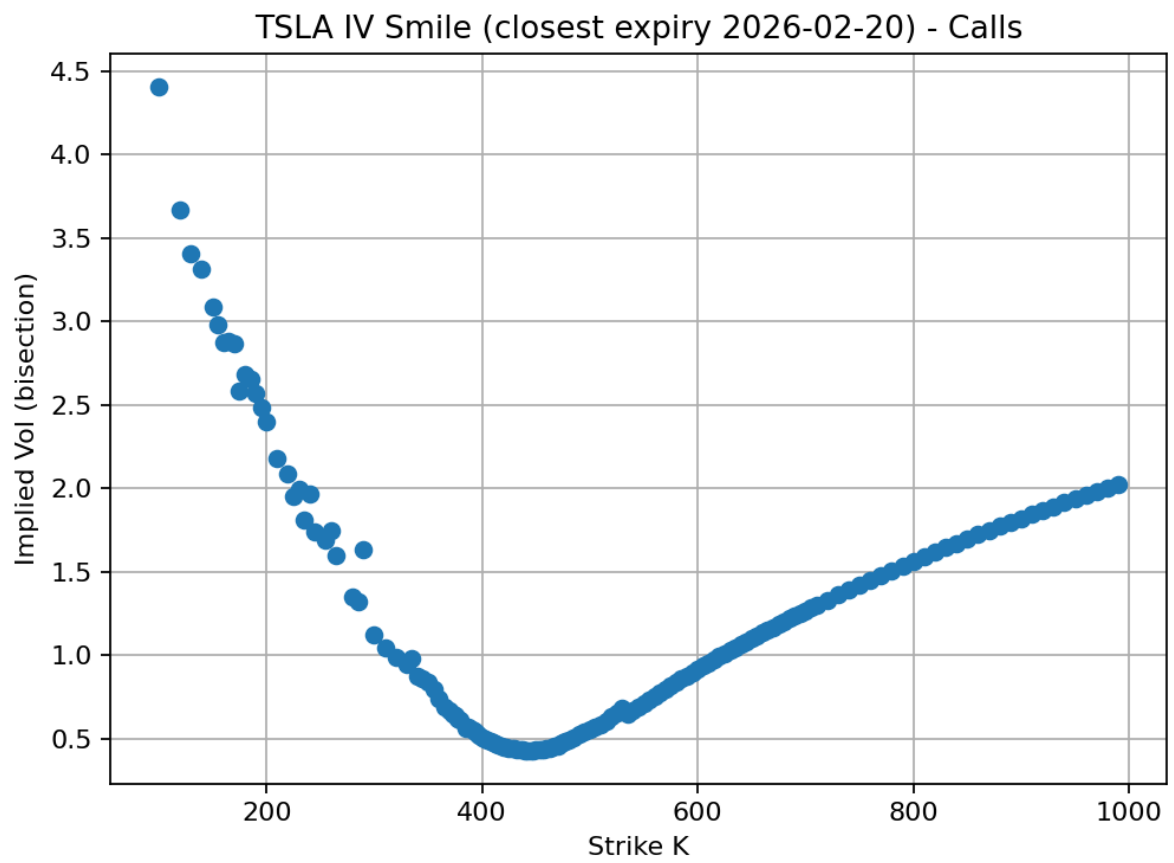
	02-20								
TSL A	2026-02-20	175	242.575	0.01	174.913	0.047759	242.537	-0.037759	0.037759
TSL A	2026-02-20	180	237.625	0.01	179.91	0.0952665	237.54	-0.0852665	0.0852665

10. Volatility smile plots (2D) and bonus 3D surface

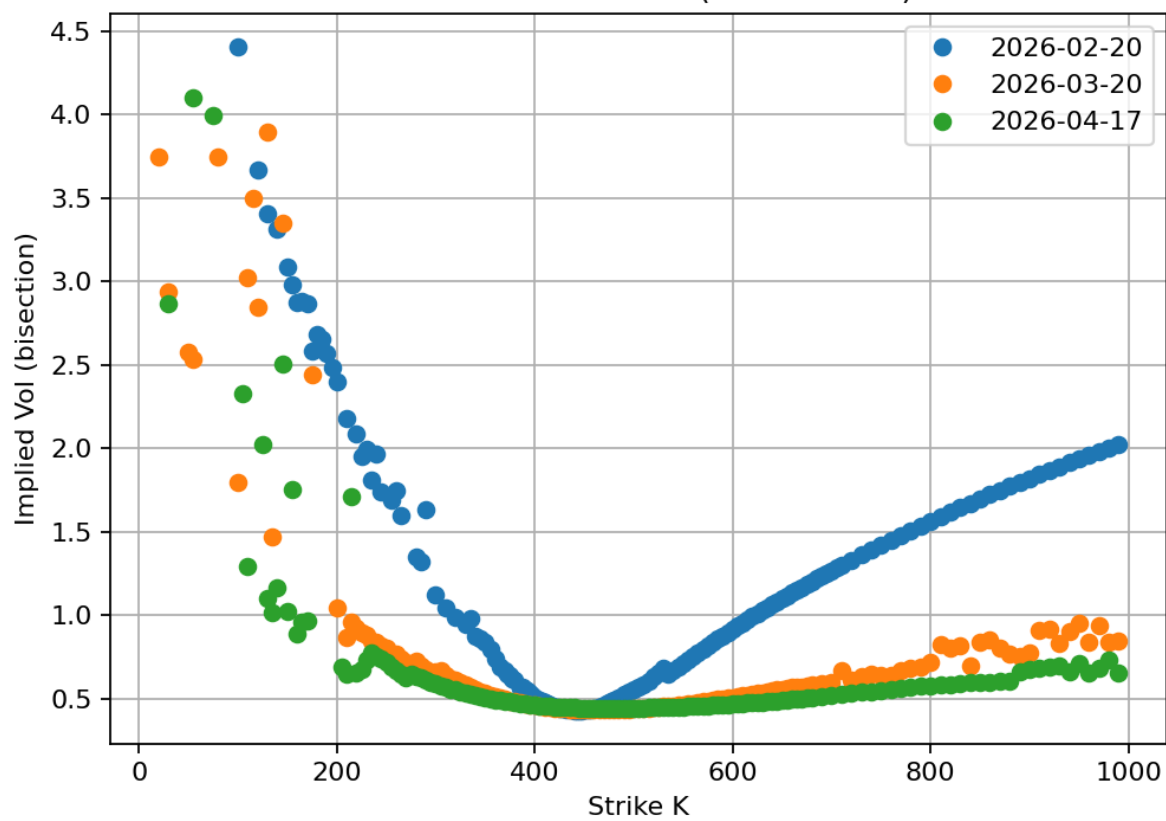


SPY IV Smile - Calls (3 maturities)

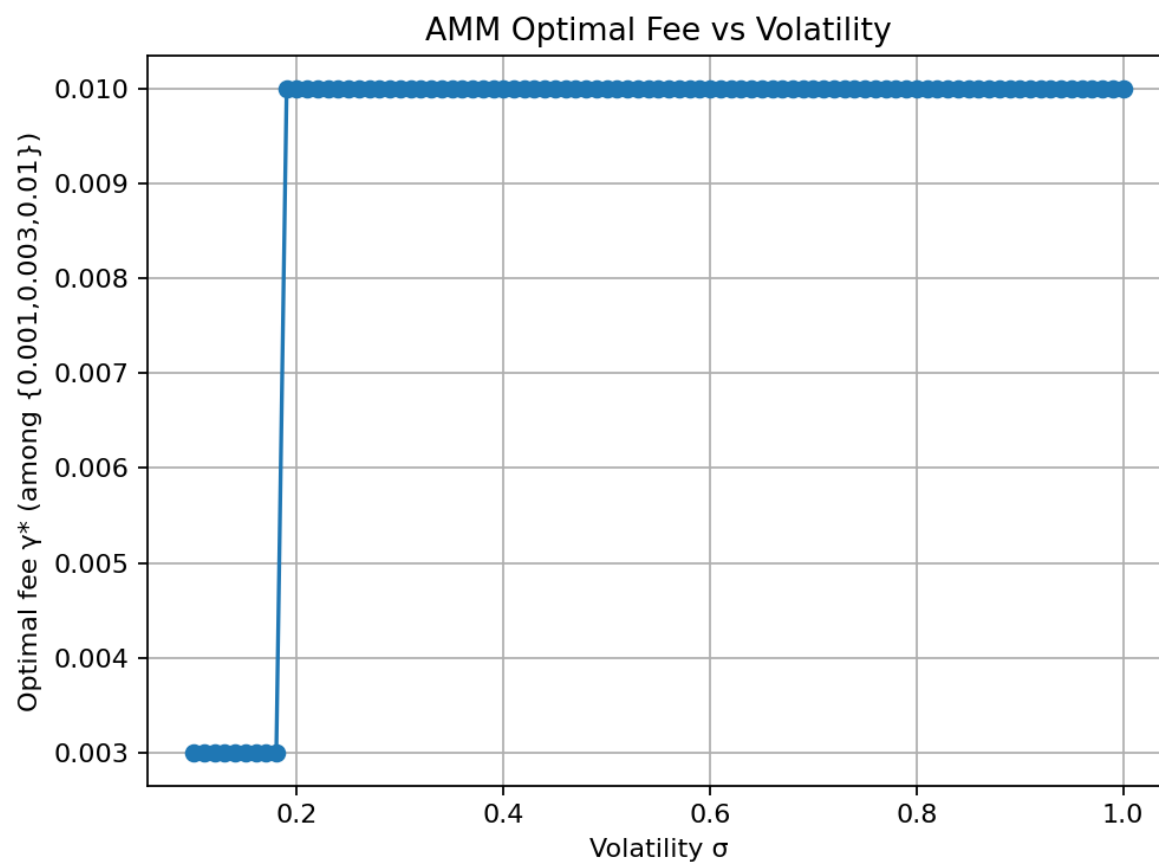




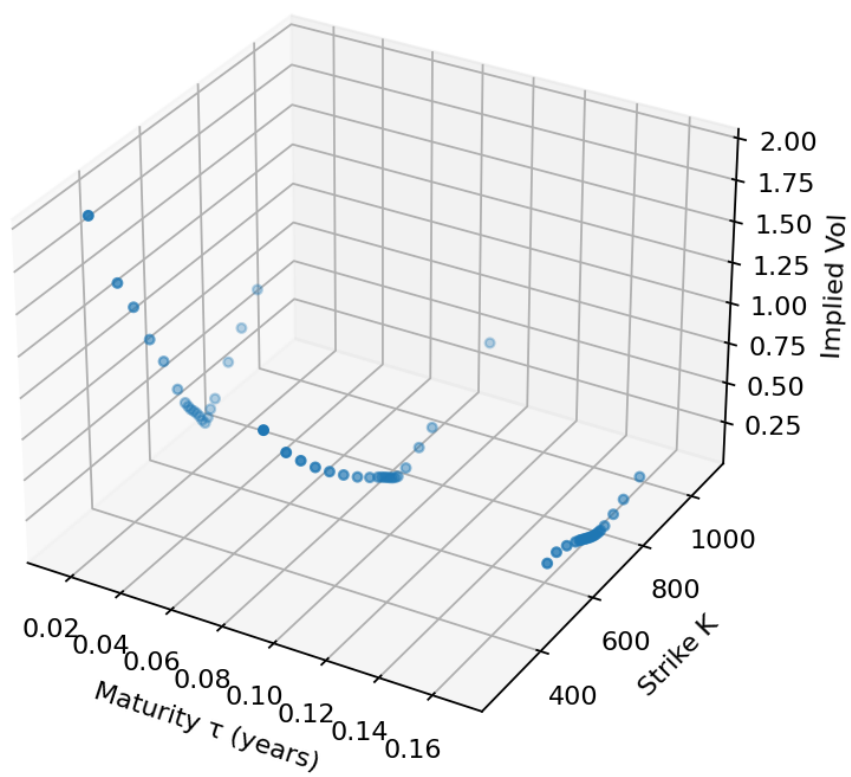
TSLA IV Smile - Calls (3 maturities)



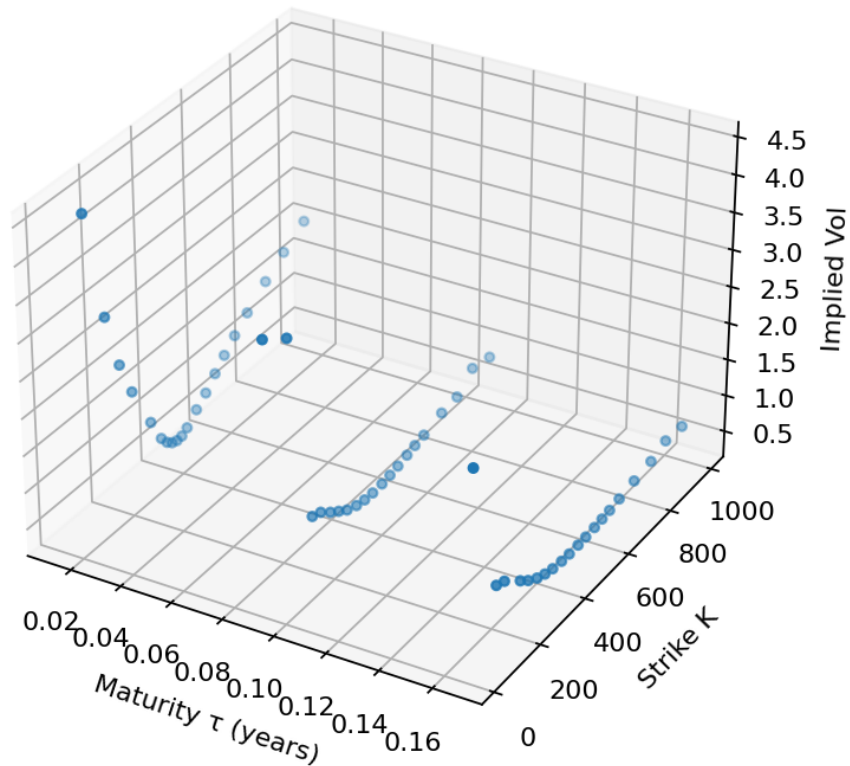




SPY Implied Vol Surface (Calls)



## TSLA Implied Vol Surface (Calls)



### 11. Greeks (analytic vs finite differences)

Analytic (call):  $\Delta = N(d1)$ ,  $\Gamma = \phi(d1)/(S\sigma\sqrt{\tau})$ ,  $\text{Vega} = S \cdot \phi(d1) \cdot \sqrt{\tau}$ . Finite differences used central differences. The table below shows close agreement for liquid strikes; discrepancies increase when numerical step sizes interact with curvature or when implied vol estimates are noisy.

symbol	expiry	K	tau	sigma	delta_analytic	gamma_analytic	vega_analytic	delt_a_fd	gamma_ma_fd	vega_a_fd	delt_a_diff	gamma_ma_diff	vega_a_diff
TSLA	2026-02-20	100	0.0136986	4.40377	0.998782	1.87539e-05	0.197143	0.998782	1.87532e-05	0.197144	-2.95775e-12	-7.1901e-10	7.0513e-10
TSLA	2026-02-20	102	0.0136986	3.66827	0.999094	1.71562e-05	0.150227	0.999094	1.71559e-05	0.150227	-9.5729e-10	-2.90986e-10	8.65002e-10

	- 20										e- 12		
TS L A	20 26 - 02 - 20	1 3 0	0.0 136 986	3.4 07 83	0.999 113	1.812 03e- 05	0.14 7404	0.9 991 13	1.81 211e -05	0.1 474 04	- 6.2 448 9e- 12	7.68 973e -10	9.4 561 2e- 10
TS L A	20 26 - 02 - 20	1 4 0	0.0 136 986	3.3 09 69	0.998 716	2.617 69e- 05	0.20 681	0.9 987 16	2.61 775e -05	0.2 068 1	- 7.4 632 5e- 12	6.01 925e -10	1.5 944 9e- 09
TS L A	20 26 - 02 - 20	1 5 0	0.0 136 986	3.0 85 87	0.998 718	2.803 09e- 05	0.20 6481	0.9 987 18	2.80 318e -05	0.2 064 81	- 1.0 506 6e- 11	8.30 893e -10	1.3 677 6e- 09
TS L A	20 26 - 02 - 20	1 5 5	0.0 136 986	2.9 79 46	0.998 723	2.894 44e- 05	0.20 5858	0.9 987 23	2.89 435e -05	0.2 058 58	- 1.3 306 6e- 11	- 8.52 324e -10	1.5 658 9e- 09
TS L A	20 26 - 02 - 20	1 6 0	0.0 136 986	2.8 76 38	0.998 729	2.984 38e- 05	0.20 4911	0.9 987 29	2.98 439e -05	0.2 049 11	- 1.0 872 6e- 11	1.26 083e -10	1.3 981 8e- 09
TS L A	20 26 - 02 - 20	1 6 5	0.0 136 986	2.8 81 81	0.998 261	3.964 99e- 05	0.27 2755	0.9 982 61	3.96 497e -05	0.2 727 55	- 1.6 648 8e- 11	- 1.44 754e -10	1.7 641 3e- 09
TS L A	20 26 - 02 - 20	1 7 0	0.0 136 986	2.8 66 65	0.997 791	4.952 94e- 05	0.33 8925	0.9 977 91	4.95 288e -05	0.3 389 25	- 1.9 537 6e- 11	- 5.47 519e -10	2.0 481 1e- 09

TS L A	20 26 - 02 - 20	1 7 5	0.0 136 986	2.5 84 93	0.998 763	3.240 77e- 05	0.19 9969	0.9 987 63	3.24 076e -05	0.1 999 69	- 1.3 679 4e- 11	- 1.64 27e- 10	2.4 277 5e- 09
TS L A	20 26 - 02 - 20	1 8 0	0.0 136 986	2.6 80 12	0.997 745	5.399 21e- 05	0.34 5422	0.9 977 45	5.39 919e -05	0.3 454 22	- 2.4 545 1e- 11	- 2.41 824e -10	2.3 753 5e- 09
TS L A	20 26 - 02 - 20	1 8 5	0.0 136 986	2.6 56 37	0.997 236	6.547 7e-05	0.41 5187	0.9 972 36	6.54 779e -05	0.4 151 87	- 2.5 667 8e- 11	9.49 599e -10	2.7 218 5e- 09
TS L A	20 26 - 02 - 20	1 9 0	0.0 136 986	2.5 68 8	0.997 197	6.856 67e- 05	0.42 0445	0.9 971 97	6.85 671e -05	0.4 204 45	- 2.7 152 3e- 11	3.80 165e -10	2.9 936 4e- 09
TS L A	20 26 - 02 - 20	1 9 5	0.0 136 986	2.4 83 57	0.997 159	7.179 84e- 05	0.42 5654	0.9 971 59	7.17 984e -05	0.4 256 54	- 2.8 534 5e- 11	- 2.88 532e -11	2.9 109 e- 09
TS L A	20 26 - 02 - 20	2 0 0	0.0 136 986	2.4 00 56	0.997 12	7.518 17e- 05	0.43 0814	0.9 971 2	7.51 814e -05	0.4 308 14	- 3.2 793 7e- 11	- 3.10 931e -10	3.3 543 3e- 09
TS L A	20 26 - 02 - 20	2 1 0	0.0 136 986	2.1 78 31	0.997 63	6.948 31e- 05	0.36 1296	0.9 976 3	6.94 831e -05	0.3 612 96	- 3.7 564 3e- 11	8.16 893e -12	4.1 491 8e- 09
TS L A	20 26 - 02	2 2 0	0.0 136 986	2.0 88 54	0.996 97	9.046 49e- 05	0.45 1013	0.9 969 7	9.04 657e -05	0.4 510 13	- 4.5 145	8.47 752e -10	4.6 196 3e- 09

	-										9e-		
	20										11		
TS	20	2	0.0	1.9	0.997	7.865	0.36	0.9	7.86	0.3	-	9.79	4.8
L	26	2	136	54	589	78e-	692	975	588e	669	4.2	732e	900
A	-	5	986	18		05		89	-05	2	916	-10	7e-
	02										1e-		09
	-										11		
	20												
TS	20	2	0.0	1.9	0.996	0.000	0.54	0.9	0.00	0.5	-	2.68	5.2
L	26	3	136	92	253	11473	5844	962	0114	458	5.9	956e	362
A	-	0	986	99		5		53	736	44	365	-10	5e-
	02										2e-		09
	-										11		
	20												
TS	20	2	0.0	1.8	0.997	8.541	0.36	0.9	8.54	0.3	-	5.33	5.8
L	26	3	136	13	569	3e-05	9651	975	135e	696	4.5	327e	552
A	-	5	986	02				69	-05	51	803	-10	6e-
	02										e-		09
	-										11		
	20												

### Part 3 — AMM Arbitrage Fee Revenue

#### (a) Swap sizes and fee revenue

Let  $P_t = y_t/x_t$  and  $k = x_t y_t$ . Case 1 ( $S_{t+1} > P_t/(1-\gamma)$ ):

$x_{t+1} = \sqrt{(k/(S_{t+1}(1-\gamma)))}$ ,  $y_{t+1} = \sqrt{(k S_{t+1}(1-\gamma))}$ ;  $\Delta x = x_t - x_{t+1}$ ,

$\Delta y = (y_{t+1} - y_t)/(1-\gamma)$ ,  $R = \gamma \Delta y$ . Case 2 ( $S_{t+1} < P_t/(1-\gamma)$ ):

$x_{t+1} = \sqrt{(k(1-\gamma)/S_{t+1})}$ ,  $y_{t+1} = \sqrt{(k S_{t+1}/(1-\gamma))}$ ;  $\Delta x = (x_{t+1} - x_t)/(1-\gamma)$ ,

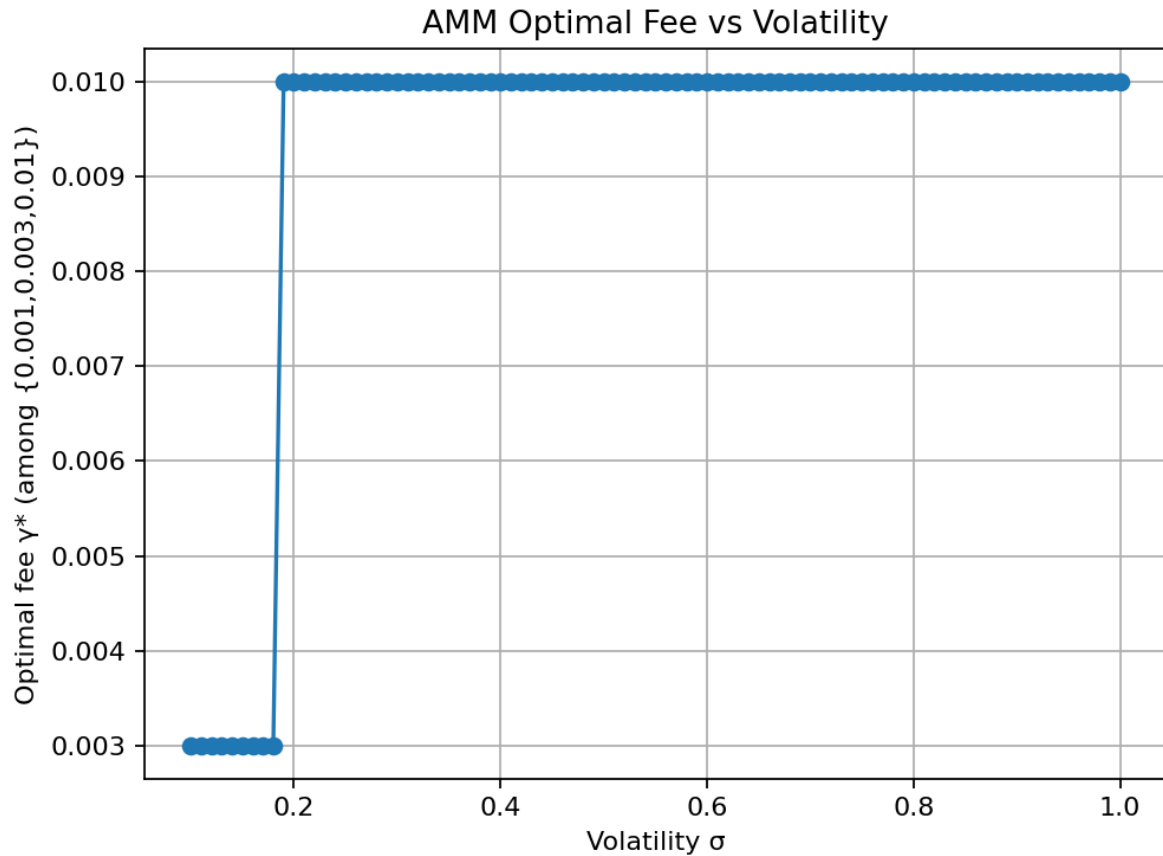
$\Delta y = y_t - y_{t+1}$ ,  $R = \gamma \Delta x \cdot S_{t+1}$ .

#### (b) Expected one-step fee revenue via trapezoidal integration

With  $x_t = y_t = 1000$ ,  $S_t = 1$ ,  $\Delta t = 1/365$ , and one-step GBM,  $S_{t+1}$  is lognormal.  $E[R]$  is computed as the sum of two integrals over the arbitrage regions, approximated numerically using the trapezoidal rule.

#### (c) Table of $E[R]$ and $\gamma^*(\sigma)$ ; $\sigma$ -grid plot

sigma	0.001	0.003	0.01
0.2	0.00368522	0.00852204	0.0094304
0.6	0.0119234	0.0329833	0.0810824
1	0.0200607	0.0573838	0.16069
sigma	gamma		E[R]
0.2	0.01		0.0094304
0.6	0.01		0.0810824
1	0.01		0.16069



Observation: as  $\sigma$  increases, expected arbitrage volume increases. In this run,  $\gamma^*=0.003$  for the lowest  $\sigma$  values on the grid, but for most  $\sigma \geq \sim 0.19$  the optimal fee among  $\{0.001, 0.003, 0.01\}$  is  $\gamma^*=0.01$ .

## Part 4 — Bonus (Double Integrals)

### 1. Analytic solutions

$$f1(x,y)=xy: \int_0^1 \int_0^3 xy \, dy \, dx = 9/4 = 2.25.$$

$$f2(x,y)=e^{\{x+y\}}: \int_0^1 \int_0^3 e^{\{x+y\}} \, dy \, dx = (e^3 - 1)(e - 1).$$

### 2. Numerical double integral using composite 2D trapezoidal rule

dx_req requested	dy_req requested	dx_ used	dy_ used	n	m	f1_ap prox	f1_e xact	f1_abs error	f2_ap prox	f2_e xact	f2_abs error
0.25	0.75	0.25	0.75	4	4	2.25	2.25	0	33.22 09	32.7 943	0.4266
0.2	0.6	0.2	0.6	5	5	2.25	2.25	4.4408 9e-16	33.06 74	32.7 943	0.2731 18
0.1	0.3	0.1	0.3	1 0	1 0	2.25	2.25	4.4408 9e-16	32.86 26	32.7 943	0.0683 11

0.05	0.15	0.05	0.15	2	2	2.25	2.25	1.3322	32.81	32.7	0.0170
				0	0			7e-15	14	943	797

Comment: f1 is integrated exactly (up to floating-point roundoff) for these grids. For f2, the error decreases as  $(\Delta x, \Delta y)$  decrease, consistent with convergence under grid refinement.



## 12. DATA2 Pricing Using DATA1 Implied Volatilities

A second market snapshot (DATA2) was collected later the same trading day. For each strike and maturity, Black–Scholes prices were recomputed using  $S(\text{DATA2})$ ,  $r(\text{DATA2})$ , and the implied volatilities obtained from DATA1. These theoretical prices were then compared to DATA2 market mid-quotes to evaluate predictive performance.

### DATA2 Market Snapshot

tag	timestamp	asof_date	tsla_spot	spy_spot	vix_level	r_annual
DATA2	2026-02-15	_20-48-05	2026-02-15	417.440002	681.75	20.6 0.05

### DATA2 Implied Volatility Summary

symbol	expiry	type	spot_S	atm_strike	atm_iv_bisect	avg_iv_band_bisect	n_opts_used	n_in_band
TSLA	2026-02-20	call	417.440002	417.5	0.456102	0.458025	157	16
TSLA	2026-02-20	put	417.440002	417.5	0.454864	0.457504	127	16
TSLA	2026-03-20	call	417.440002	415.0	0.441995	0.441596	138	8
TSLA	2026-03-20	put	417.440002	415.0	0.448792	0.448843	129	8
TSLA	2026-04-17	call	417.440002	415.0	0.446807	0.446505	131	8
TSLA	2026-04-17	put	417.440002	415.0	0.456767	0.457290	137	8
SPY	2026-02-20	call	681.750000	682.0	0.200489	0.196075	166	68
SPY	2026-02-20	put	681.750000	682.0	0.200864	0.218022	194	68
SPY	2026-03-20	call	681.750000	682.0	0.168143	0.165961	148	68
SPY	2026-03-20	put	681.750000	682.0	0.187335	0.189820	202	68
SPY	2026-04-17	call	681.750000	682.0	0.156855	0.153932	137	64
SPY	2026-04-17	put	681.750000	682.0	0.186096	0.188993	150	65

Observed pricing discrepancies are expected due to bid–ask noise, intraday spot movements, changes in volatility surface shape, and liquidity effects. This confirms that

implied volatility is time-varying and that yesterday's surface does not perfectly predict today's option prices.

Overall, the DATA2 repricing exercise demonstrates that Black–Scholes with frozen DATA1 volatility provides a reasonable first-order approximation, but systematic errors arise as the volatility smile and underlying price evolve.