

# MTHM059 Assignment 1

## Sentimental Returns: Extended Acoustics

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RAHUL SINGH, NOVEMBER 2025

750047133

# Premise

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- Better understanding of investor behaviour through unconventional data lenses.
- Edmans (2022)[1] showcased music as a groundbreaking dimension of investor behaviour.
- **Questions:**
  1. What role do mathematical models play?
  2. Can we enrich [1] for a more informative representation of investor behaviour?

# Role of Mathematical Modelling & Music

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- Unambiguous, communicable understanding of physical phenomena. Examples: Brownian motion, probability distributions, etc.

$$\begin{aligned} dS_t &= \mu S_t dt + \sqrt{\nu_t} S_t dW_t^s, \\ d\nu_t &= \kappa(\theta - \nu_t)dt + \xi \sqrt{\nu_t} dW_t^\nu \end{aligned} \quad P(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

- Better data (observation)  $\Leftrightarrow$  better mathematical models.
- Why music? Simple: individual participants make markets  $\rightarrow$  individuals use music to express mood,  $\therefore$  music influences the markets. A common idea, mathematised, provides a sharper understanding of a market dynamic.

# Research Recap: SWAV, Data, Findings

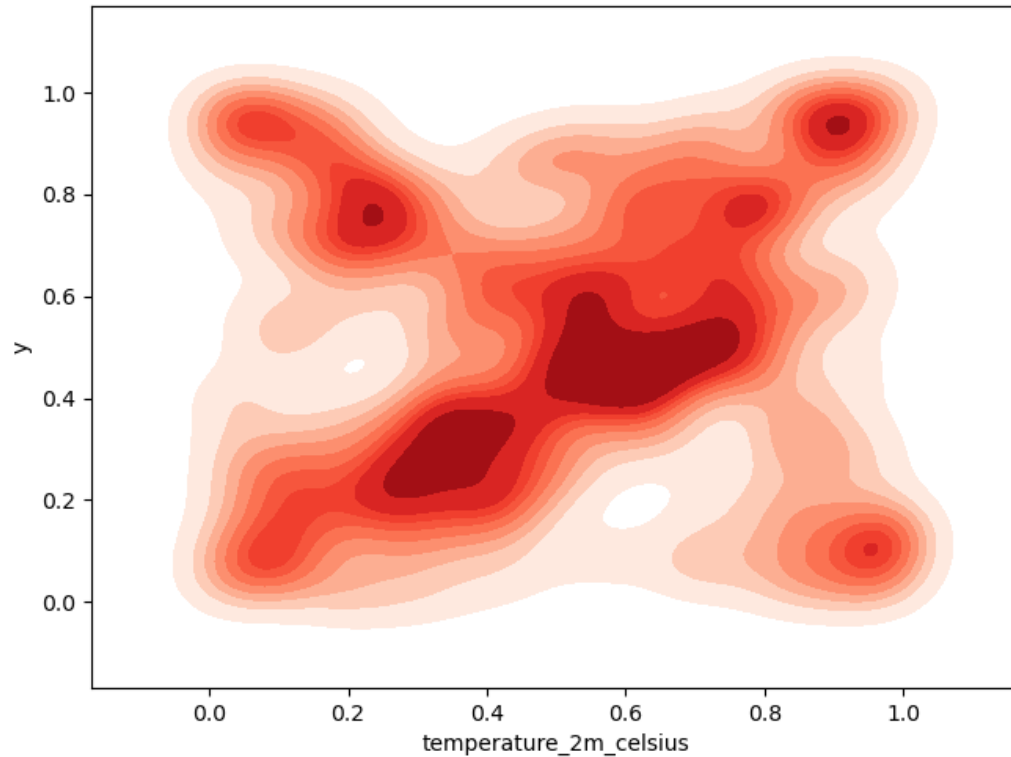
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- “Valence” proxies endogenous mood; [1] used stream-weighted average valence:

$$SWAV_{i,d} = \sum_{j=1}^{200} \left( \frac{Streams_{j,i,d}}{\sum_{j=1}^{200} Streams_{j,i,d}} \cdot Valence_{j,i,d} \right)$$

- Denoised with first-differencing, daily→weekly aggregation, and 2.5% winsorisation.  $\therefore MS_{i,t} = SWAV_{i,t} - SWAV_{i,t-1}$ .
- Preprocessed controls  $\supset$  weather, macroeconomics, EPU, volatility; local & global.
- Panel regressions & DID showed that a  $+1\sigma$  in  $MS \rightarrow +4.3\%$  annualised contemporaneous weekly return, with a next-week reversal of  $-3.7\%$ . Results held across markets, local controls, market caps, and arbitrage limits.

# Nonlinear Modelling – Copulae & TVP



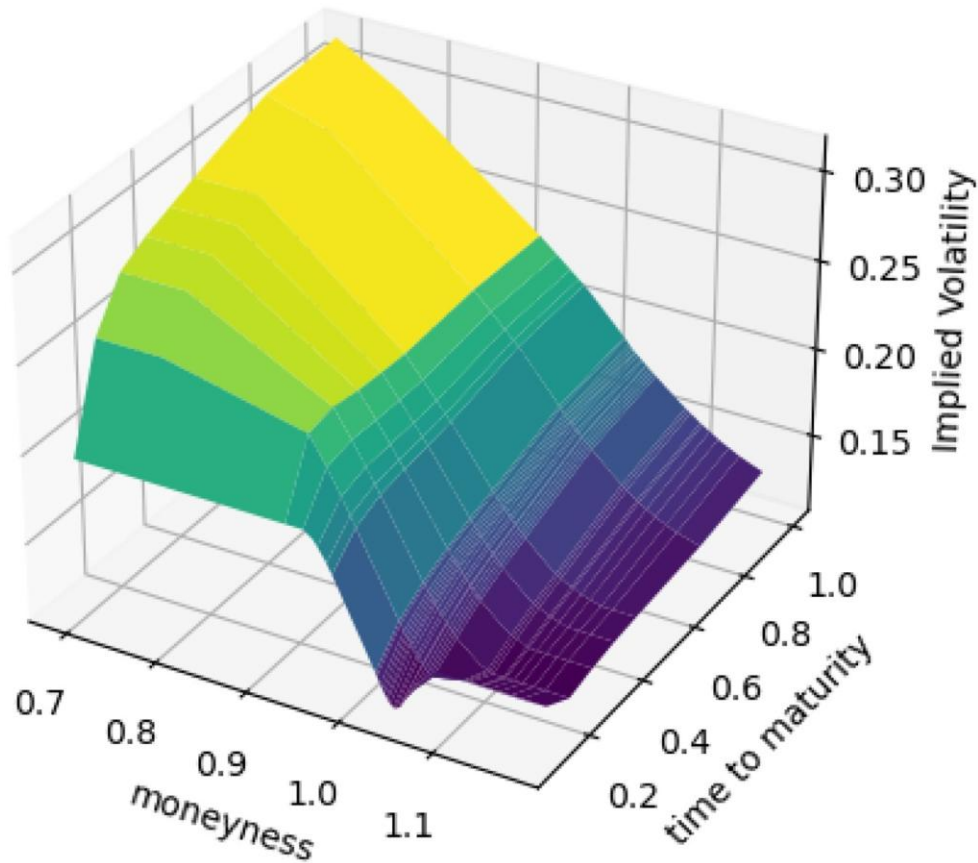
*Sample Copula: Beverage sales vs. weather.  
Borrowed from [9].*

**Nonlinear modelling:** Copulae uncover nonlinear relationships. Regime-switching models allow evolution of coefficients. Both reveal persistent & cascading effects [3, 4, 5, 6, 7, 8].

**Image:** Sample rank-based empirical copula showing regimes between weather (x-axis) and retail beverage sales (y-axis) after whitening & normalising. Dark red regions show high mass co-movement; lighter regions, lesser strength.

$$R_{i,t} = \alpha + \beta_1 MS_{i,t} + \sum \Gamma \text{Controls}_{i,t} + \varepsilon_{i,t} \rightarrow R_{i,t} = \alpha_t + \beta_{1,t} MS_{i,t} + \sum_{i=1}^N \Gamma_{i,t} \text{Controls}_{i,t} + \varepsilon_{i,t}$$

# Volatility Surface Expansion



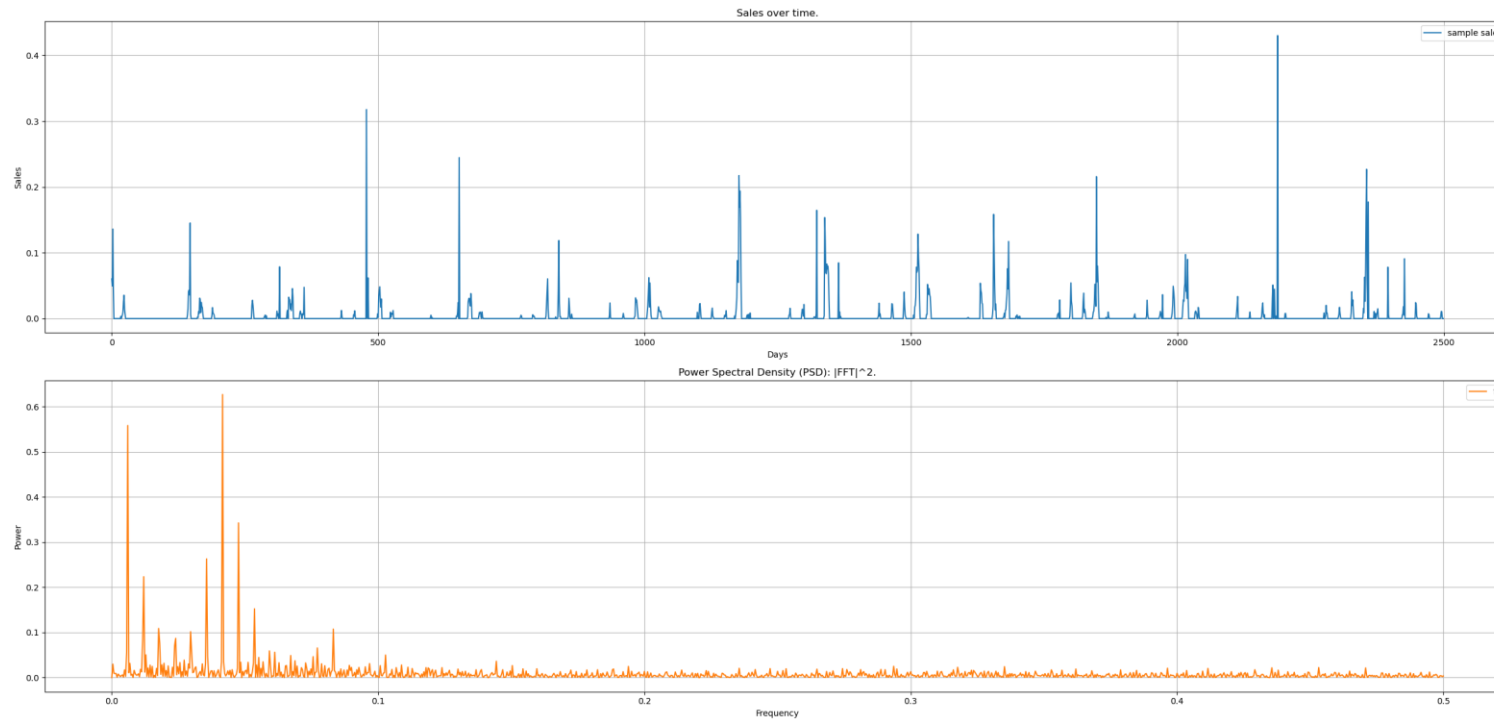
Sample IVS: SPX on 01/11/2021. Borrowed from [10].

**IV Surfaces:** Volatility is nonlinear, skewed, and shifts with market conditions. Research has highlighted a significant data-driven relationship between IVS, sentiment, and returns [4, 5].

**Image:** how Implied Volatility (IV) changes over different option prices (moneyness) relative to the current market level, and simultaneously over time in the market (time to maturity, TTM). Longer TTM increases IV because the market has more time to move. Market levels lower than current have the highest IV; higher than current have the lowest – nonlinear & time-varying.

Also a proxy for sentiment, because this spread indicates bearishness.

# Frequency Transformations



**Spectral features:** Fourier and EMD transformations shift time-signals into frequencies, highlighting periodic patterns. Literature shows that these refine nonlinear/surface analyses, and additional music descriptors beyond valence add further explanatory power [2, 5].

**Image:** Sales over time represented in frequencies, highlighting periodic behaviour. Example specifically chosen to accentuate periodicities.

*Sample FFT: Shows how sales in time (top) transforms to frequency (bottom).*

# References

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1. Alex Edmans et al (2022), "Music sentiment and stock returns around the world", Journal of Financial Economics, Volume 145, Issue 2, Part A, Pages 234-254, ISSN 0304-405X, doi:10.1016/j.jfineco.2021.08.014
2. Can TL, Le MD, Yu K (2024), "Music sentiment and the stock market in Vietnam". Journal of Asian Business and Economic Studies, Vol. 31 No. 1 pp. 74–83, doi:10.1108/JABES-07-2022-0170
3. Sinda Hadhri et al (2025), "Listening to the Market: Music sentiment and cryptocurrency returns", Journal of International Money and Finance, Volume 157, 103394, ISSN 0261-5606, doi:10.1016/j.jimonfin.2025.103394
4. Ksenija Doroslovacki et al (2024), "A Novel Market Sentiment Analysis Model for Forecasting Stock and Cryptocurrency Returns", IEEE TRANSACTIONS ON SYSTEMS, MAN, AND CYBERNETICS: SYSTEMS, VOL. 54, NO. 9, doi:10.1109/TSMC.2024.3402160



# References

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5. Jiahao Weng, Yan Xie (2024), "Degree of Irrationality: Sentiment and Implied Volatility Surface", arXiv cs.LG, doi:10.48550/arXiv.2405.11730
6. Salah A. Nusair, Jamal A. Al-Khasawneh (2022), "Impact of economic policy uncertainty on the stock markets of the G7 Countries:A nonlinear ARDL approach", The Journal of Economic Asymmetries, Volume 26, e00251, ISSN 1703-4949, doi:10.1016/j.jeca.2022.e00251
7. Kuang-Liang Chang (2009), "Do macroeconomic variables have regime-dependent effects on stock return dynamics? Evidence from the Markov regime switching model", Economic Modelling, Volume 26, Issue 6, Pages 1283-1299, ISSN 0264-9993, doi:10.1016/j.econmod.2009.06.003
8. Huang, Chai Liang et al (2023), "Nonlinear Effects of Temperature on Returns and Investor Optimism–Pessimism from Winner and Loser Stocks", Review of Pacific Basin Financial Markets and Policies, Vol. 26, No. 01, 2350003, doi:10.1142/S0219091523500030

# References

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9. Rahul Singh (2025), “Hospitality Sales: A Deep Dive — Part I”, Medium, [url:medium.com/@quantaco\\_rahuls/hospitality-sales-a-deep-dive-part-i-70da11e50fd8](https://medium.com/@quantaco_rahuls/hospitality-sales-a-deep-dive-part-i-70da11e50fd8)
10. Rama Cont & Milena Vuletić (2023), "Simulation of Arbitrage-Free Implied Volatility Surfaces", Applied Mathematical Finance Volume 30, Pages 94-121, doi:10.1080/1350486X.2023.2277960