

Optimize your marketing spend with Julia!

Introduction

Motivation

"Half the money I spend on advertising is wasted; the trouble is I don't know which half."

- John Wanamaker, 19th century

Today?

- Global Ad Spend >\$700bn (2021, estimate)
- Privacy-conscious consumers

Solution? Media mix modelling (MMM)!

- Quantify the benefits of your marketing activities
- Re-allocate marketing spend to maximize revenues

Challenges:

- Insufficient data
- Underspecified problem

Implementation:

- Bayesian Latent Variable Regression + Bayesian Decision Theory
- Pure Julia (DataFrames.jl + TableTransforms.jl + Turing.jl + Metaheuristics.jl and more)

Detailed methodology, practical tips and other resources can be found in the associated repository

Practical Example

Setting

- A local business using 3 marketing channels (total spend \$10,000):
 - TV ads (\$5,000)
 - Google Search Ads (\$2,500)
 - Facebook Ads (\$2,500)

Key Questions:

- What is marketing's **contribution towards revenues**?
- How to **re-allocate marketing \$\$\$\$** across TV, Search, Facebook to maximize revenues?

Marketing 101

ROAS (Return on Ad Spend) = Contributed revenue / Ad spend

Example: ROAS of 2.0x means that we captured twice as much revenue as we spent on ads, ie, 100% return

mROAS (marginal ROAS) = the additional revenue for the next \$1 spent on ads given our existing spending

Example: mROAS of 2.0x @ \$1,000 means that spending \$1,001 would earn +\$2 extra

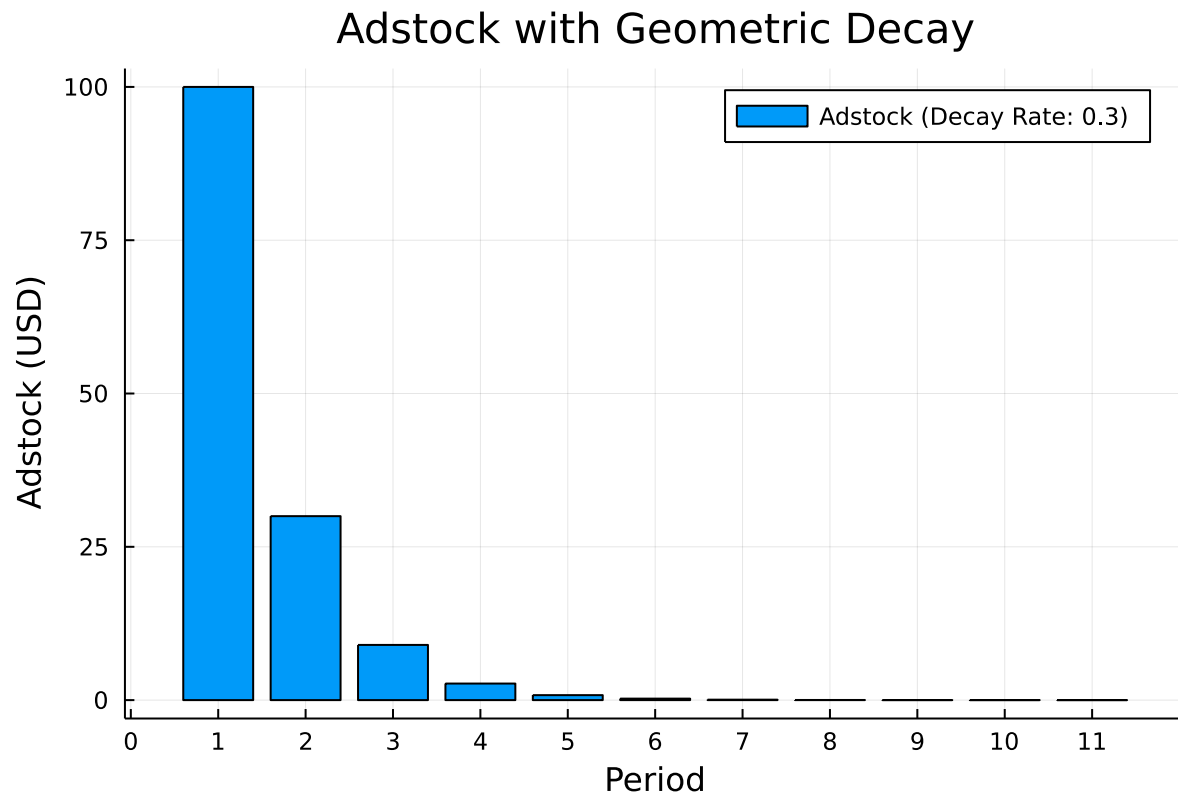
Adstock ("lagged effects")

To capture the lagged effects of marketing spend (ie, seeing conversion revenues in the subsequent periods)

Formula (Geometric Decay): $A_t = X_t + A_{t-1} * \theta$ where X_t is the ad spend in period t and θ is the Decay Rate

Example: Ad Spend of \$100 with $\theta = 0.3$ will lead to an effect equivalent to \$30 in the next period

Decay Rate (θ): 0.3



Saturation ("diminishing returns")

To capture when the effect of Ad spend depends on its magnitude

Formula (Hill Curve): $\beta_{max} * \frac{A^n}{A^n + K_{0.5}^n}$ where β_{max} is the maximum effect, A = Ad spend, n = Hill coefficient ("slope"), and $K_{0.5}$ = Halfmax concentration point is the spend where effect is equal to 0.5

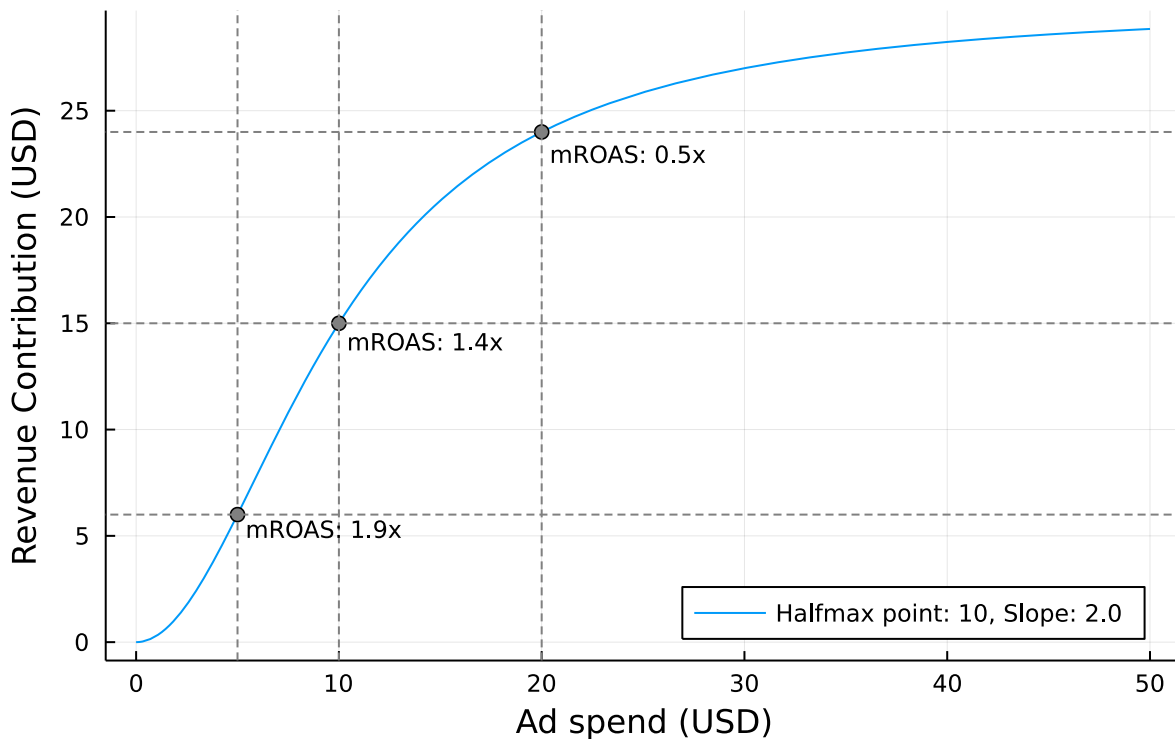
Example: When \$10 -> \$11 creates +\$1.4 of Conversion Revenue (mROAS = 1.4x) versus \$20 -> \$21 creates +\$0.5 of Conversion Revenue (mROAS = 0.5x). This is often referred to as diminishing returns

Maximum effect (β_{max}): 30

Halfmax conc. point ($K_{0.5}$): 10

Slope (n): 2.0

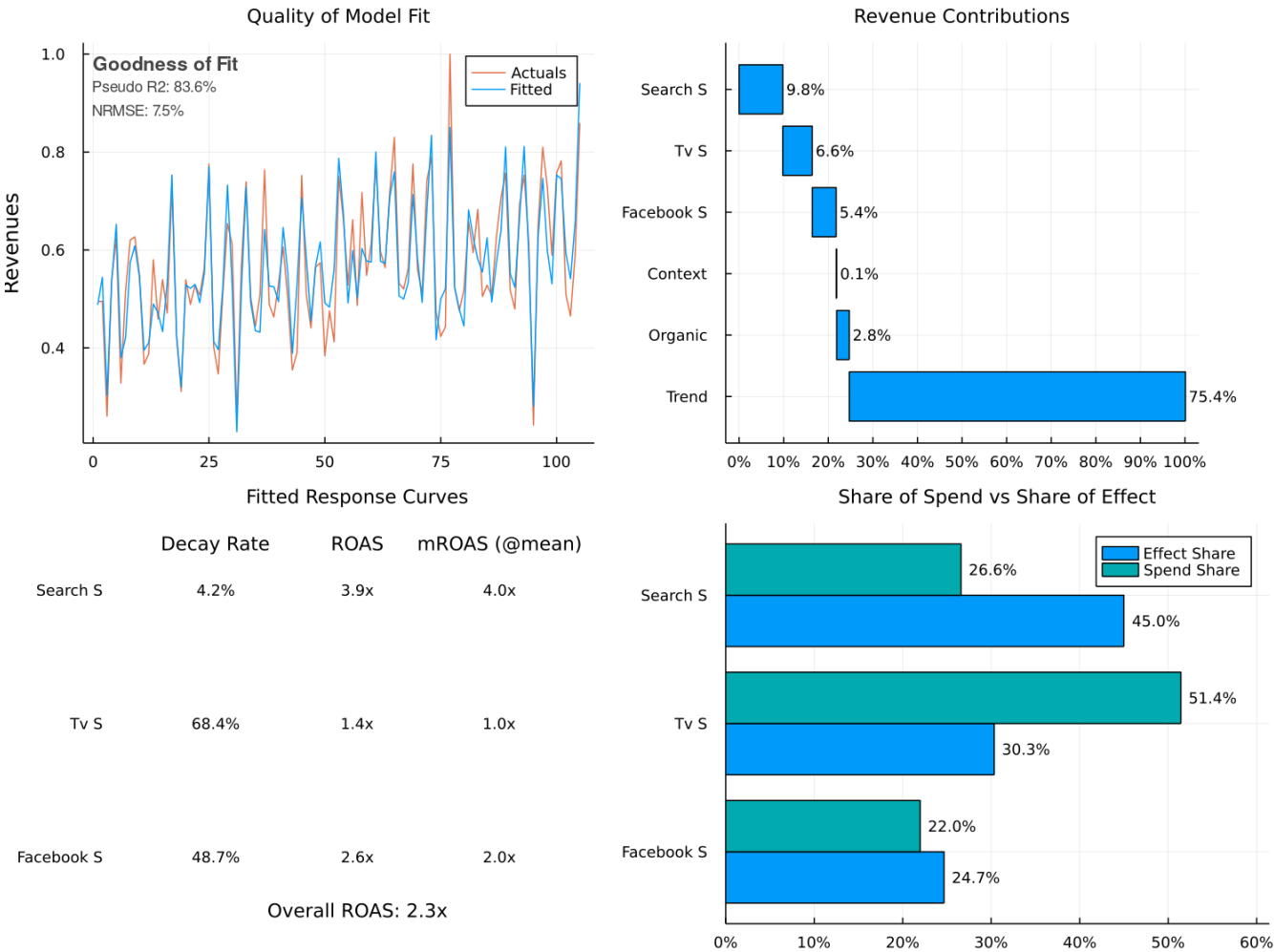
Hill Curve Ad Spend Saturation



Results

Fitted Model 1-Pager Reveal: 5

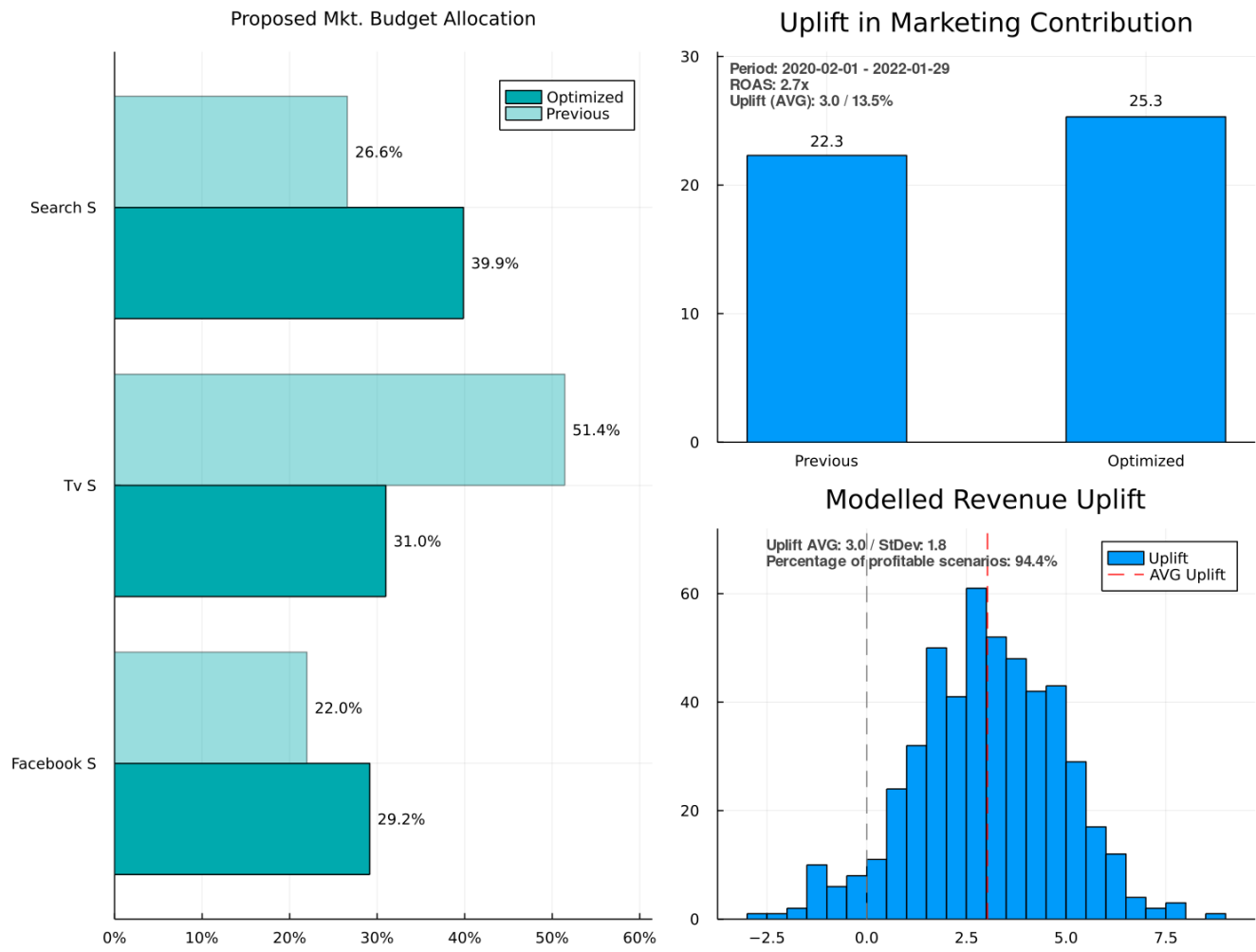
MMM 1-Pager



Optimization

Optimization 1-Pager Reveal: 4

Budget Optimization 1-Pager



Thank you!

Any questions?

This practical example, tips and tricks and additional resources can be found at the associated repository!