Online Advertising Incrementality Testing

Practical Lessons, Paid Search And Emerging Challenges

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Who we are ...



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Tutorial Parts

- **1.** The basics: context and challenges
- 2. Incrementality Testing: concepts, solutions and literature
- **3.** From concept to production: platform building, challenges, case studies
- **4.** Deployment at Scale: test cycle and case studies
- **5.** Emerging trends: identity challenges, industry trends and solutions







Part 4

Deployment at Scale: test cycle and case studies





Experimentation: Incrementality Testing Cycle



Experiment Design and Planning

Minimum Detectable Lift,
Total Audience Available,
Hold-out group size,
Exposed Audience
Needed, Experiment
Duration, etc



Intervention Execution

Start holding out users randomly from showing Ads. Log "Ghost" Ad impressions for these users



User Tracking and Metrics

Track user cohorts.
Compute aggregate
metrics to guide high-level
decisions and campaign
tuning



Incremental Performance and Recommendations

Document incrementality metrics, **CR lift, Cost per Incremental Action**. Provide **recommendations**, side effects and learnings





Testing Cycle

From the beginning to the end, step by step







Experimentation: Design and Planning



Experiment Design and Planning

Minimum Detectable Lift,
Total Audience Available,
Hold-out group size,
Exposed Audience
Needed, Experiment
Duration, etc

Run statistical power analysis to manage expectations, budgets and duration

- In a period of time, given existing advertiser data:
 - Number of users exposed to impressions, conversions, impressions and total spend
 - Metrics derived: converter rate, total audience, eCPM, ad frequency

- Parameters to tune:
 - Holdout size (never 50%!!), minimum detectable lift, test significance





Power Analysis Simulation: 2 months of data

Total Target Population

Aggregate Statistics:

Converter rate: XXX%

Conversions per 100 users: XXX

Conversions per converter: **XXX**

Impressions per user: XXX

Experiment Users: XXXM

Scenario Simulation:

Converter Lift precision: **5%**

Expected Converter Rate:

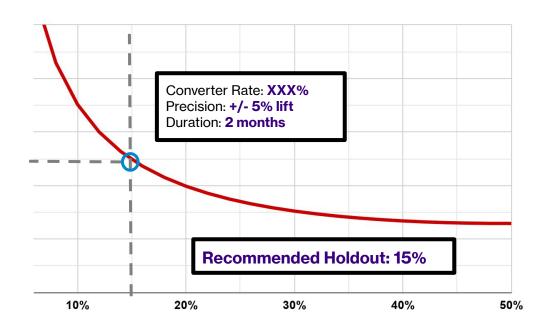
Control: <XXX%

Test: >XXX%

Test Significance: **5%**₁







Hold out Percentage

Response: User signup/acquisition

Experimentation: Intervention Execution



Intervention Execution

Start holding out users randomly from showing Ads. Log "Ghost" Ad impressions for these users

The test actually runs

- Run quick sanity checks:
 - Verify impressions and ghost impressions are generated
 - Is percentage of control users as expected?
 - Define guidelines to hit a button and stop the test if needed





Experimentation: User Tracking and Metrics

Track user groups, metrics and mid-flight recommendations



 Lift, incremental conversions, user ad frequency, CPIA, iROAS

Pay special attention to holidays and fundamental user behaviour changes

- For relevant holidays, cumulative effects are realized in these special periods (ad stock realization)
- For irrelevant holidays sometimes inventory becomes more expensive



User Tracking and Metrics

Track user cohorts.
Compute aggregate
metrics to guide high-level
decisions and campaign
tuning





Experimentation: Performance and Recommendations

Testing is expensive, thus document learnings and insights



Including statistical tests and intervals

These metrics will compared and used to calibrate advertiser's financial models

Incremental Performance and Recommendations

Document incrementality metrics, **CR lift, Cost per Incremental Action**. Provide **recommendations**, side effects and learnings

- Document lessons and make recommendations
 - What can be learned from the test beyond top-level metrics?
 - Why the test was (or was not) successful?
 - Make business recommendations and campaign management practices





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Business Use Cases: Recap Barajas and Bhamidipati (2021)

New advertiser wants to test waters before fully deploying budgets

Existing advertiser wants to scientifically prove incremental value

Incrementality of tactical practices with external validity for future deployment

Quarterly regular strategic adjustments (retrospectively)



Case Studies

New advertiser wants to test waters before fully deploying budgets





Onboarding Funnel: Key Findings

+200%

Statistically Significant Lift Effect on Landing Page Visits (funnel entry)

12.4%

Consistently Positive Average Lift Effect on Purchase Confirmation (successful end of the funnel)

Incremental Funnel
Entries and Successful
Endings





Lift Results: Funnel Entries, Stages and Successful Endings







Lift on **Purchase Confirmation**, end-of-funnel conversions

12.4% +/- 10%

Marketing Effectiveness goes beyond one conversion metric



Case Studies

Existing advertiser wants to scientifically prove incremental value







Insurance quotes: Key Findings

10.5%

Converter rate lift (Statistically significant)

9.5%

Lift in Avg conversions per user (Statistically significant)

\$XX

Cost per Incremental
Conversion over 8 weeks of
experiment

87%

Lower Cost per Incremental Conversion than post-click CPA

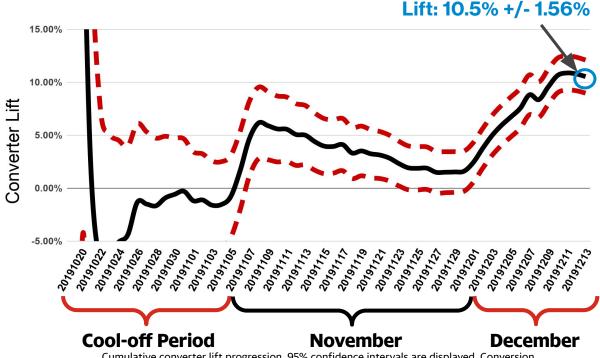






Key Insights: Incrementality Real Study

Industry standard last-touch attribution undervalued native display programmatic advertising by 87%







Cumulative converter lift progression. 95% confidence intervals are displayed. Conversion definition: insurance quote. Duration: 8 weeks, 10/18/2019 - 12/12/2019. Holdout size: 10%.



Case Studies

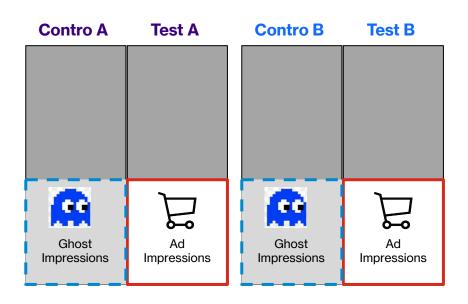
Incrementality of tactical practices







A/B test + Incrementality test: Multi-cell testing



A benefit from double blind designs, an A/B split test fits easily to run a multi-cell test

User split is required to test A and B tactical practices





Tactical Practices on user Signups: Key Findings

4.5%

Tactic A cell reached Statistically Significant **Converter rate lift**

\$XX

Cost per Incremental Converter Tactic A

2.2%

Tactic B cell reached Statistical Significant **Converter rate lift**

vs **\$XX**

Cost per Incremental Converter Tactic B

Tactic A outperforms **Tactic B**







What if we can't randomize user ids?

We'll review identity challenges and trends in the next and last part of the tutorial....





