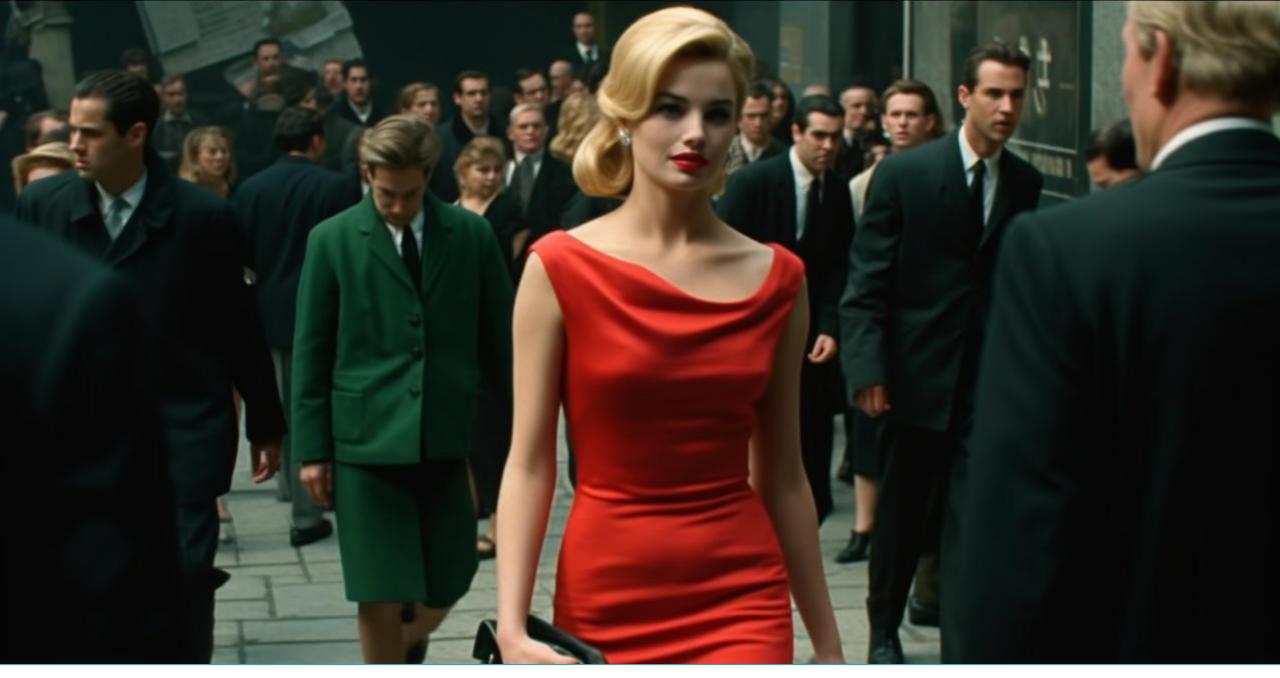


[MILTON DATA]

Synthetic Data: a very real path to cross-media measurement

11 September 2024

IAB Measure Up 2024 New Zealand Synthetic Data & Cross Media Measurement



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Synthetic Data Basics

- Synthetic data most common currently in medical fields
 - Privacy focus
 - Alternative scenarios
 - Capture distribution/structure of real data





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 - Image classification
 - Cyber-security
 - Autonomous system development



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 - Capture distribution/structure of real data
- Military & Security applications
 - Image classification
 - Cyber-security
 - Autonomous system development
- Training AI Systems
 - Alpha Go most famous example
 - Rendering engines as inputs for autonomous vehicles
 - Amazon training Alexa language system on synthetic data
 - American Express & synthetic financial data for fraud detection



Applications in Media and Marketing

- Early days but showing potential
- Projects around the world
 - Iris (Australia/UK)
 - VOZ (Australia)
 - Dovetail (UK)
 - WFA Halo/"North Star" (USA/UK)
- Core elements in each
 - Audience measurement applications
 - Privacy preserving approaches
 - Integration of different data sources
 - Emphasis on cross-media/cross-platform measures









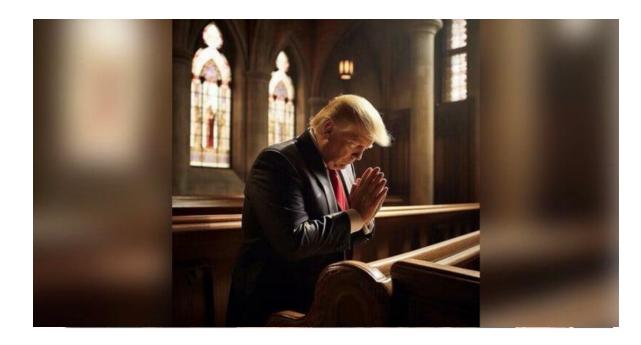
Making Synthetic Data – General Version

- "Classic" approaches
 - Generative Adversarial Networks (GAN)
 - Variational Auto-encoders (VAE)
- Problems for audience measurement applications
 - LLMs revealing training data
 - No control on representativeness



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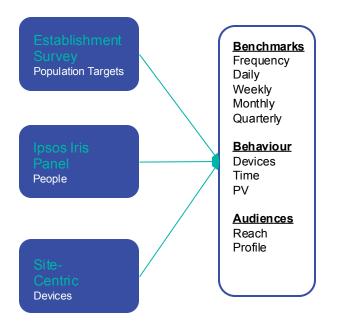


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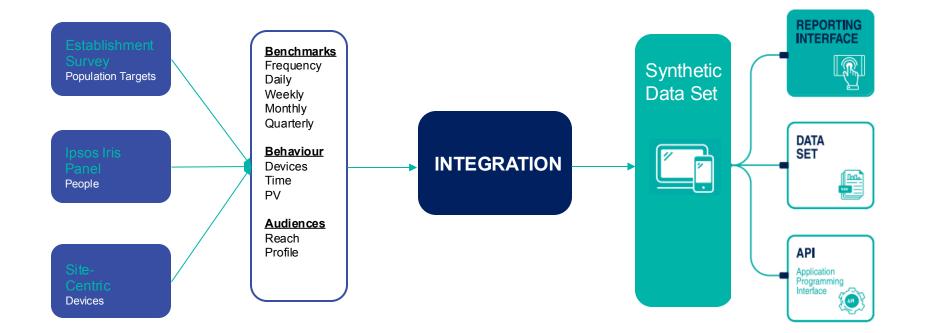
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 - It's not always so obvious



- Synthetic data in media more about controlled integration (Ipsos Iris Australia)
 - Survey data (establishment survey)
 - Panel data (consumer behaviours)
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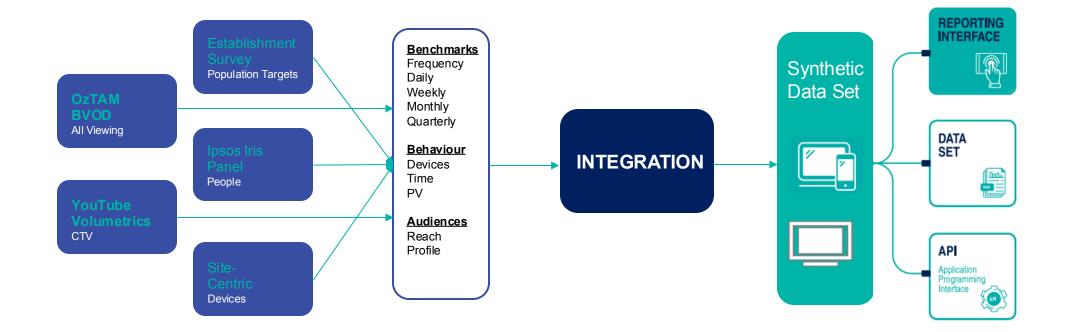




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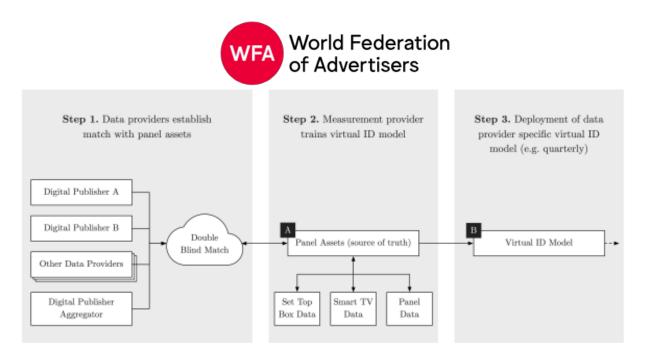
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Can be extended to include additional data sets



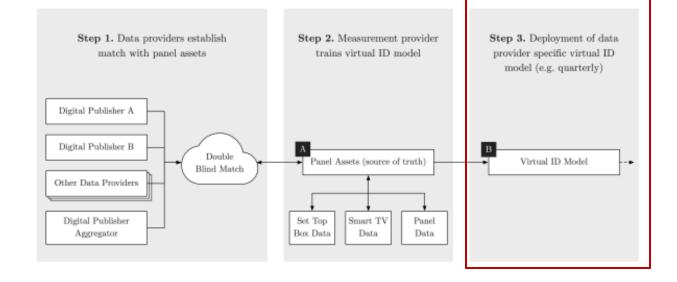


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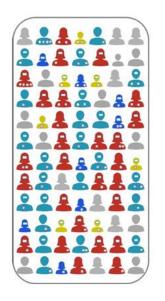




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 - Assignment via demographic models
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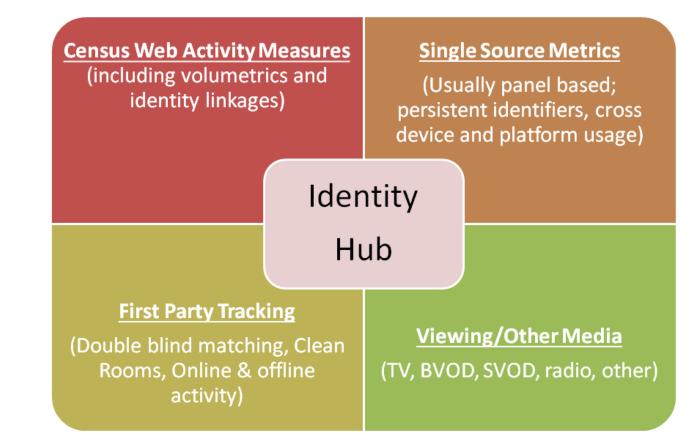


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- Virtual Id Model
 - Create Identity Hub (different to id graph)
 - Train on census/panel data



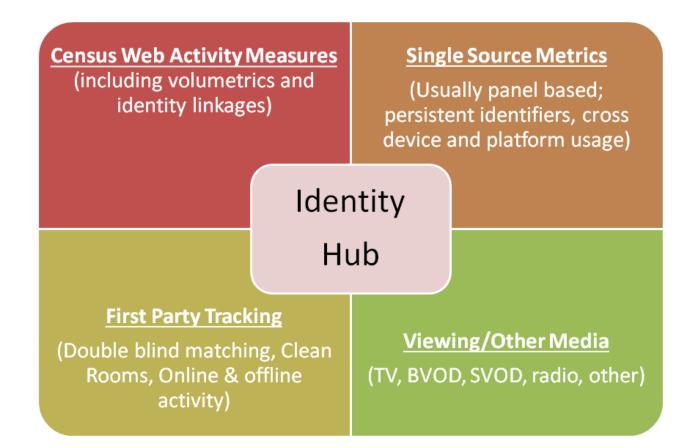
Using Synthetic Data – Data Integration

- Identity Hub approaches
 - Completely anonymous (audience applications)
 - 1:1 mapping (activation applications)



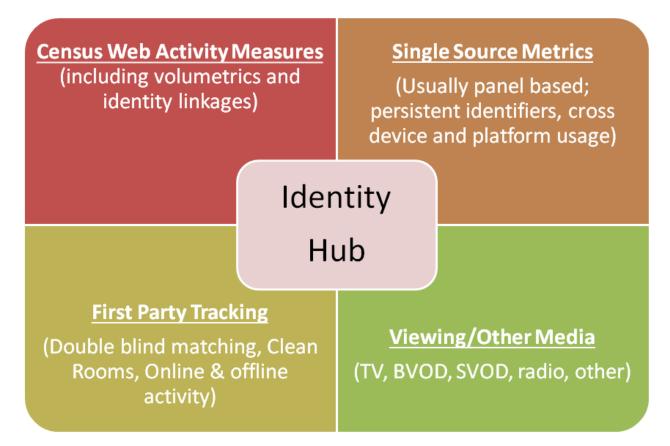
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 - Obtaining cross-platform usage estimates
 - Conforming data sources
 - Quality of inputs



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- Importance of First Party Data
 - Owning your viewer/user/customer
 - Improvements in data matching processes
 - Offline measures



New Zealand Perspective – Social and Policy

- Generating a synthetic version of New Zealand
 - Victoria University
 - University of Auckland
 - Stats New Zealand
- Based on Integrated Data Infrastructure
 - IDI links multiple data sources
 - Restrictive confidentiality requirements
- Synthetic data aim
 - Meets stringent privacy requirements
 - Minimise re-identification risk (differential privacy)
 - No ability to "reverse engineer" source data
 - Cognizant of Māori data sensitivities





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- But what about marketing and media?





• Cookies were never a complete solution





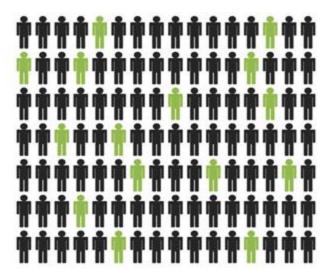
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- Synthetic data supports broader measurement objectives

	IAB Measurement Framework		
	Media Effectiveness	Brand Effectiveness	Sales Effectiveness
	(Improve Delivery)	(Impact Brand)	(Increase Sales
	View ability/attention	Ad recall	Market Mix Modelling
Techniques	Campaign metrics	Brand lift	Multi-touch attirbution
	Cross media R&F	Brand equity	Sales lift
	Valid impressions	Consumer perceptions	Consumer behaviour
Measurement	Viewed by humans	Associations	Long term effects
	Attention	Influences on brand choice	Short term activation
	Impressions	Brand awareness	Customer lifetime value
Metrics	Deduped R&F	Brand favourability	ROI Measures
	Demo profile	Purchase intent	Change in sales
	Attention time	Brand equity	Incrementality

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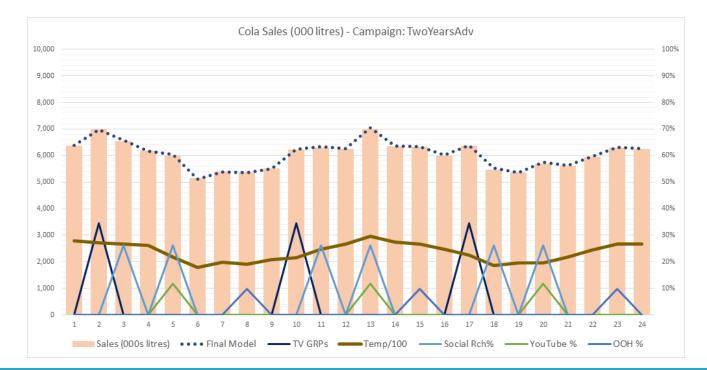
• Start curating your data sets

- Customers (current & potential)
- Find out what you can about them
 - First party data
 - Second party data
 - Third party data





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- Start curating your data sets
 - Customers (current & potential)
 - Find out what you can about them
 - First party data
 - Second party data
 - Third party data
- Enrich your data sets
 - Campaign measures
 - Other media
- Other signals
 - Offline sales/leads
 - Environment (economy/weather)
 - Competitor activity



Conclusions

- Synthetic data applications will grow
 - Complementary approaches
 - Privacy enabled
 - Vs
 - Links to activation
- Core takeaways
 - Start curating data now
 - Don't limit yourself to online datasets only
 - Quality of inputs essential
 - Importance of matching algorithm(s)
- How to improve
 - Richer data sets
 - Link to real-world
- Keep learning





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