

Recommender Systems

Deploying recommender systems into production, and measuring the results



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Recommender Systems

Digital Content



Recommender Systems

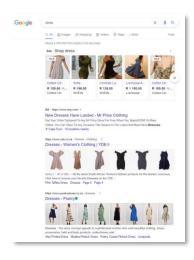
E-Commerce (On/Offline)



Social Media

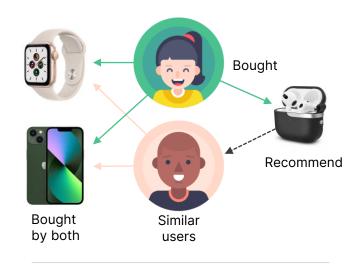


Online Advertising



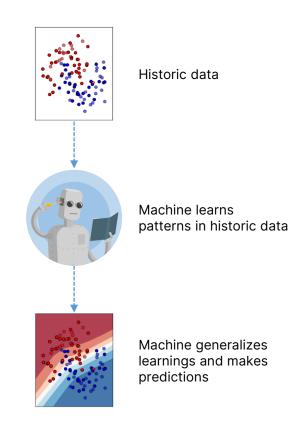
Typical Methods Behind Recommender Systems

Neighborhood-Based





Model-Based

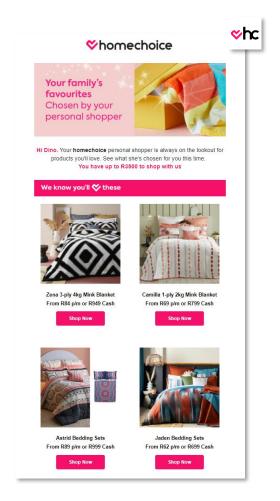


Recommender Systems

Hybrid Examples







Powering The Change

Programming Paradigm



Al is software that writes software

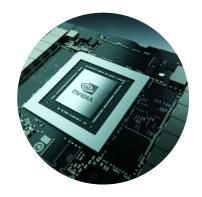
Recommender Systems

Data **Modernization**



Big data collection, organization and operationalization

Accelerated Computing



80x GPU improvement over CPU in a few years

Best Practice Al Production

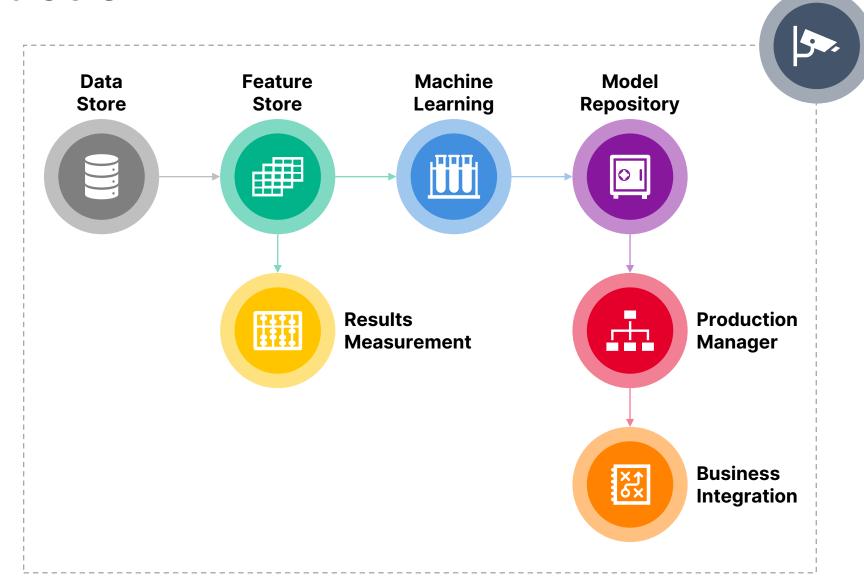


From concept to integrated and scaled product

Slide 5

Al Production

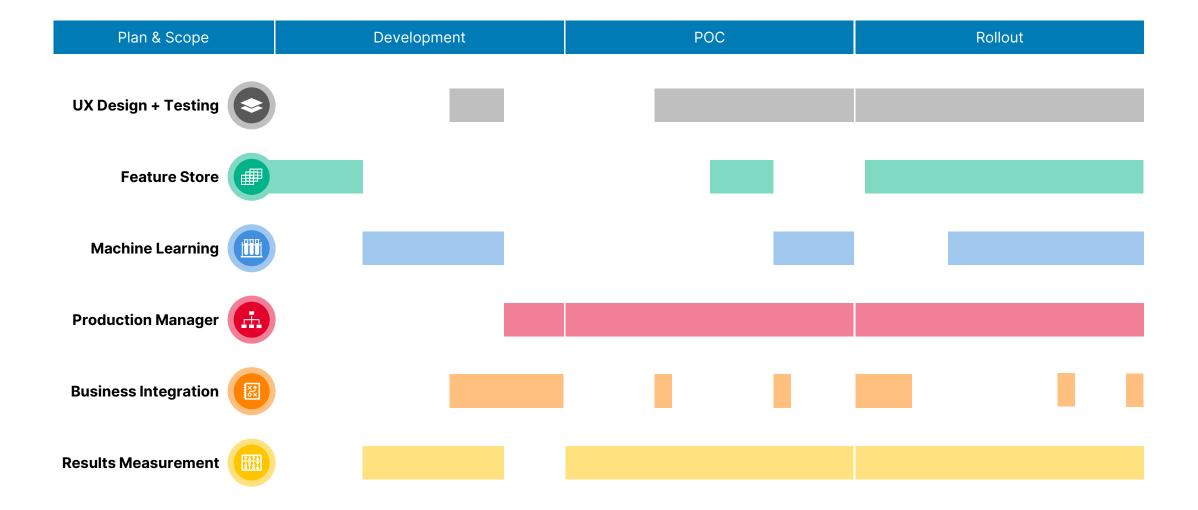
Recommender Systems



Process

Monitoring

Development and Deployment Plan



Tracking Performance

Model Development

Model **Training**



Training data

Model **Testing**



Data not used in training

Model **Validation**



Unseen data

Retro **Testing**



Train on one period, validate on a future period

Shadow Scoring



As if the model was live

Model Production

POC



Live - small sample or single channel

Production



Live – full rollout

Structured Approach To Results Measurement

Build Test Framework



Flexible Scalable Reusable

Recommender Systems

Select Test Groups



A/B selection Test grain Reselection horizon

Estimate Uplift Drivers



Track key metrics Est. behavior change Explain calculations

Calculate **Profit & ROI**



Convert uplift to profit Work with Finance Business buy-in

Typical Production Problems

- Over engineering
- Loss of critical data feeds
- Manual execution
- End-user internet / data constraints
- Not measuring results
- Slow rollout lack of momentum 6
- Underestimating the costs

Common Recommender System Flaws

Significant Data Requirements



Recommender Systems

Sparse User-Item Data



Missing **Product Data**



User/Item Cold Start Problems



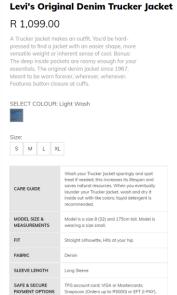
Long Payback Period



Perception-Based Recommenders



Natural Language Processing systems understand product name, brand, description, pricing, sizing, fit, & other text-based data



Additional product shots augment image data

Recommender Systems







Computer Vision systems understand image composition, color, gender, style, design, shape, size, material, branding, & other image-based data







GAN inpainting improves image occlusions

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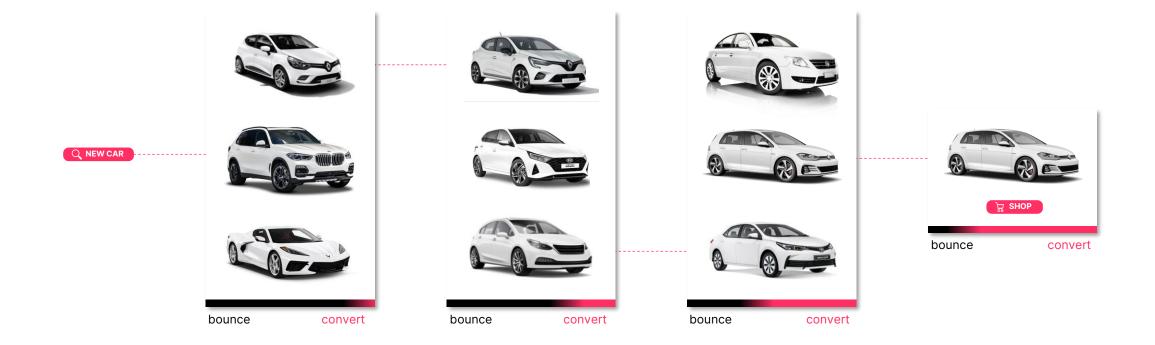
Unstructed data embeddings allow real-time processing







Session-Based Recommenders





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