Large Scale Machine Learning in Digital Advertising

Seyed Abbas Hosseini Cofounder, Pegah Inc. Ph.D. 2018, Sharif abbas@tapsell.ir







Outline

- Digital Advertising
 - **o** Sponsored Search
 - Display Advertising
- RTB Mechanism
- Bid Estimation
 - CVR Estimation
- Other Interesting Issues
- Who We Are?!

Conveying advertisers' message to target audience in online media



Sponsored Search

Google	iphone 6s case	🔱 🔍 We	inan III 🕻
	Web Shopping News Images Videos More ▼ Search tools About 16,900,000 results (0.33 seconds)		2 0
iPhone 6s Cases - case-mate.com M www.case-mate.com/iPhone-6s-Cases ▼ 4.6 ★★★★★ rating for case-mate.com Shop The iPhone 6s Case Collection. Free Standard Shipping! Refined Protection - Slim & Tough - Genuinely Crafted - Premium Designs iPhone 6s M www.apple.com/ ▼ The only thing that's changed is everything. Learn more. A9 chip - Two sizes - Now in rose gold Pre-order 9.12 - iPhone Upgrade Program - 3D Touch - Cameras		Shop for iphone 6s case on Google Case-mate - Karat Case Fo S49.99 Best Buy (163) Shop for iphone 6s case on Google Moshi - Iglaze Armour Case S39.99 Best Buy (161)	Sponsored ① Moshi - Overture Wall \$49.99 Best Buy (18)
Ļ	In the news Speck's iPhone 6s CandyShell + MightyShell cases bring best-of-breed protection to Apple's latest iPhones 9 to 5 Mac - 1 day ago With the iPhone 6s and iPhone 6s Plus debuting next week, it's important to start thinking	Case-mate - Brilliance Cas \$44.99 Best Buy ***** (294) Case-mate - Wallet Folio C \$54.99 Best Buy ***** (173)	Case-mate - Karat Hard Sh \$49.99 Best Buy
	Moonin's immone as and as Plus cases offer premium protection More - 23 hours ago Top 5 Best Leather iPhone 6s Cases Heavy.com - 12 hours ago More news for iphone 6s case	Search: iphone 6	is cas

Search Engine

! Q		سوپرمارکت
نصب	دیجیکالا ۲۰۰ ★ اکمی	digikala
نصب	سوپرمارکت - بازی برای بچهها ۴.۳ *	
نصب	اسنپ مارکت (سوپرمارکت آنلاین) ★ ۴.۲	Market!
نصب +خرید درون برنامهای	سوپرمارکت - بازی فروشگاه کودکانه ۸ ۵.۳	
لا همه		ويديوها
		THE AISLES
ک بازار من	ويديو دستهها ج	ہ صفحہ اصلی

www.apple.com > iPhone > iPhone 6s ▼ Apple Inc. * The essential Apple-designed cases, accessories and all-new aluminum docks for iPhone 6s and iPhone 6s Plus.

Get protection that inspires confidence with iPhone 6s cases and covers from OtterBox.

iPhone 6s Cases & Covers from OtterBox www.otterbox.com/en-us/iphone-6s-cases * OtterBox *

iPhone 6s - Accessories - Apple

Demandware SiteGenesis.

App Market



- Advertiser sets a bid price on Keywords
- User searches the keyword
- Search engine or market owner ranks ads and selected the best match

Display Advertising

اِیگانپیش بینیکن! شروعکن ہ

<mark>قرعهکشی</mark> جوایز



	» مورینیو: نمیدانم چه در سر اریکسن میگذر د	شیخ دیاباته (استقلال)	آخرين اخبار ساير ورزشها
	» جای بوسکتس باشم از بار سا میروم	مېدى ترابى (پرسپوليس)	
	» دو رکور د تاریخی برای مباجم آلمانی لایپزیش	عيسى آل كثير (صنعت نفت آبادان)	» لشگری رییس کمیته پزشکی فدر اسیون جهانی ووشو شد
		محمدر ضا سليمانی (سايپا)	» لوبه با غفور همچنان بدون شکست در سریآ
ىد	کسی های جد	کہروش استنلی سوارز (سپاھان)	» میزبانی جامجهانی به ایر ان بر گردانده می شود؟
ارجى	\delta 👗 🔪 باشگاه های خ	حسين مالکی (شاھين	🕷 خلاصه بسکتیال فینیکس سانز – دنور ناگتس
	Iminto	مېکل شېردارىيوشېر)	» وزنه برداری فقط مردانه نیست!
	لیگ های خارجی	ساسان انصاری (تراکتور)	» حذف کمانداران ریگرو از قبرمانی آسیا
			🖬 خلاصه بسکتبال دالاس ماوریکس – هیوستون راکتس
•	لیک برتر انگلیس	مطالب پیشنہادی 🗢	» لیگ NBA: شکست هیوستون در خانه
	انتخاب هفته :		» شب فراموش نشدنی آدم فضایی و شرکا در مادرید
÷	هفته ۱۳		» عبادیپور: هنوز ۱۰۰ درصد آماده نیستیم
	لیگ برتر انگلیس		🕷 حرکت های بر تر لس آنجلس کلیپرز در فصل ۲۰۱۹-۲۰۱۹
	شنیه ۲ آذر ۱۳۹۸		» عطایی: ار دوی ۲۰ روزه تیم ملی والیبال منطقی نیست
00	وستبام ۲۰۳ تاتنبام		» شاهین طبع : پیگیر ار دوی تدار کاتی بر ای تیم ملی هستیم
0	واتفورد ه – ۳ برنلی		» جام جہانی کشتی فرنگی در ایران لغو شد
00	آرسنال ۲ - ۲ ساوتیمپتون		🕷 اسپانیا قہر مان رقابتھای دیویس کاپ شد
0	بورنموث ۱ - ۲ ولورهمپتون	خرید پیراهنهای مردانه رسمی و شیک از ۱۲۹ هز ار تومان!	» شېر دارې ورامين همچنان بدون شکست
	برایتون ۲۰۰ لسترسیتی		» پیروزی نانجینگ مقابل فوجیان با در خشش خدادی
n	دریستان پانس ۲۰۱۱ نیورپون	بېترين پاسورها	📧 خلاصه والیبال راه یاب ملل مریوان ه – شهر داری ور امین ۳
00	منچسترسیتی ۲-۱ چلسی	انتخاب لیگ :	🖿 خلاصه والبیال خاتم از دگان – بیام مشید
	یکشنبه۳آذر ۱۳۹۸	لیگ برتر ایران	» دادگر : البام هاشمی گزینه اول و آخر ماست
10	شفیلدیونایتد ۳-۳ منچستریونایتد		» دوست ندار م به من بگویند مر د عنکیوتی
	دوشنبه ۴ آذر ۱۳۹۸	剂 ۴ پاس معین عباسیان (سایپا)	» معدد ۲ کماندا، ایران به دور دوم مسابقات حذف
P 10-10	استون ويلا – نيوكسل	وریا غفوری (استقلال)	» شواندان سور بری با در در برای در ایران ایران شد
	مشاهده جدول کامل لیگ بر تر انگلیس	امید نورافکن (سپاهان)	»» شغبانیان شریح می میرونیدن بنون بیزان سد. ۲۰۰۰ با ۱۰۰۰ بین ای گارههای شدکاگه
~	a 51, 5 Å	مسعود شجاعی (تراکتور)	»» برد استنتایی برای دوسی سیدنو به د مماد بادین شکست. میمان چینفان
2	مطالب پیسبادی	محمد محبی (سپاهان)	» نیمهای بدون سخست، میهمان حریقان سایا داد ۲۰ در مدر معنان جامحیات سایر بازماند
		۳ پاس حسین ابراهیمی (نفت	» ایران از معود به نیمه بایی جام جهانی سابر بار ماند. استان از معود به نیمه بایی جام جهانی ایران از ماند.
-	The second second	مسجدسلیمان)	🖬 خلاصه بسکتبال میامی هیت – فیلادلفیا سیکسرز
		میلاد جہائی استعت نفت آباداں	» سیدبندی مسابقات بسکتبال انتخابی المپیک مشخص شد
1		مېدى بر دەن بېر س جىوبى جىر	» یک پیروزی و یک شکست در انفرادی کامپوند
1/2		مېدى بر،بى پر سپويىس،	» برتری قاطع پر سپولیس و سایپا مقابل حریفان -
-11		لیگ قبر مانان ار ویا	
ں کلیک	خوش ساختترین خانههای گلستان را با یک		ليک برتر واليبال
	ببينيدا	گروه A	انتخاب هفته :
			A 5166



- Advertiser targets a segment of users
 - No matter what the user is searching or reading
- Ad Network selects the best ad to show to the user

Digital Advertising Ecosystem



Display Advertising Ecosystem



6. User Feedback (click, conversion)

- Buying ads via RTB, 10 billion per day
- A real big data battlefield

	Query per Second
Turn DSP	1.6 million
Google	40,000 search

Auction Mechanism



First Price Auction





Second Price Auction



- Each Advertiser has many campaigns
- With different Pricing Schemas
 - CPM: cost per mille impression [favored by publisher]
 - CPC: cost per click
 - CPA: cost per action [favored by advertiser]
- Goal: Maximize Revenue



- Simple Solution:
 - Select ad based on Expected Revenue per Impression
 - suppose: ad a, goal cpc





CVR Estimation: Problem Definition

• Problem Definition

• Available Data about

- User
- Context
- **Ad**

One instance data

- Date: 20160320
- Hour: 14
- Weekday: 7
- IP: 119.163.222.*
- Region: England
- City: London
- Country: UK
- Search Query: "iphone 6s case"
- OS: Windows
- Browser: Chrome
- Ad title: "iphone 6s case on sale!"
- Ad content: "Customize your case design"
- Bid keywords: "iphone case"
- User occupation: Student
- User tags: Sports



⇒

Click (1) or not (0)?

Predicted CTR (0.15)

• One-Hot Binary Encoding

Sparse representation: x=[5:1 9:1 12:1]

- **Prediction Challenges:**
 - High Dimensional Data
 - **o** Too Sparse Feature Vectors
 - Very Unbalanced Classification [The convert events are too rare]
 - Real-time response [<100ms]

CVR Estimation: Predictive Models

- Generalized Linear Models
 - Logistic Regression
 - Bayesian Probit Regression
- Factorization Machines
 - Sparse Factorization Machines
 - Field-Aware Factorization Machines
 - Field-Weighted Factorization Machines
- Deep models
 - Deep CTR Predictor
 - Deep Factorization Machines
 - Wide and Deep Recommender Systems

Generalized Linear Models

General Form

$$p(y|x,w) = f(w^T x)$$

- Logistic Regression
 - Likelihood is convex and hence Parameters can be learnt using ML ٠
 - Learning can be done in an online fashion using stochastic Gradient Descent ٠

$$p(y = 1 | x, w) = \sigma(w^{t}x)$$
$$E(w) = -\ln p(Y | X, w) = \sum_{n=1}^{N} y_{n} \ln \sigma(w^{T}x) + (1 - y_{n}) (1 - \ln \sigma(w^{T}x))$$

- Bayesian Probit Regression
 - A fully Bayesian method based on a Gaussian prior over latent weights ٠
 - Posterior can be found online using stochastic variational inference ٠
 - **Bing's Sponsored Search CTR Prediction algorithm** ٠

$$W \sim \prod_{i=1}^{N} \prod_{j=1}^{M_i} N(w_{ij}; \mu_{ij}, \sigma_{ij}^2)$$
$$y = sgn(w^T x + \epsilon) \quad where \quad \epsilon \sim N(0, \beta^2)$$
$$\Rightarrow p(y|x, w) = \Phi(\frac{y \cdot w^T x}{\beta})$$



0.5

Generalized Linear Models

- Pros
- Fast Prediction
 - Only one inner Product should be calculated
- Fast Learning Methods
 - Efficient online algorithms exist for both proposed methods
- Interpretable
- Cons
 - Linear models don't consider correlation among features
 - Linear models can only memorize feature combinations which users have already performed actions on

Factorization Machines

• One way to consider inter-feature correlations is using polynomial kernels

$$p(y|x,w) = f(\phi(x,w))$$

$$\phi(x,w) = \sum_{i,j\in F} w_{ij}x_ix_j$$

- Challenge: the model has $O(N^2)$ parameters where N is the number of features
 - A very common idea in machine learning in this scenario is using factorized models

$$\phi(x,w) = \sum_{i,j\in F} v_i^T v_j x_i x_j$$



Field-Aware Factorization Machines

- In FMs, every feature has only one latent vector to learn the latent effect with any other feature
- In FFMs, each feature has several latent vectors. Depending on the field of the other features, one of them is used to do the inner product.

Clicked	Publisher (P)	Advertiser (A)	Gender (G)
Yes	Tabnak	Digikala	Male

$$\phi_{FM}(x, w) = v_{Tabnak}^T \cdot v_{DigiKala} + v_{Tabnak}^T \cdot v_{Male} + v_{DigiKala}^T \cdot v_{Male}$$

Factorization Machines

- Pros
- Fast Prediction
 - Only one inner Product should be calculated
- Considers Correlation Among Features
 - FFM won many Kaggle challenges due to its superior performance

- Cons
 - Learning FM models is more computational expensive than linear models
 - Learning the parameters can't be done online
 - FMs can't consider correlations among more than two features
 - Over-generalization

Wide & Deep Model

- Memorization of feature interactions through a wide set of cross-product feature transformations are effective and interpretable
- Generalization requires more feature engineering effort.
- Deep neural networks can generalize better to unseen feature combinations through low dimensional dense embeddings learned for the sparse features.
- Deep neural networks with embeddings can over-generalize and recommend less relevant items when the user-item interactions are sparse and high-rank

$$P(Y = 1 | \mathbf{x}) = \sigma(\mathbf{w}_{wide}^{T}[\mathbf{x}, \phi(\mathbf{x})] + \mathbf{w}_{deep}^{T}a^{(l_f)} + b)$$



Wide & Deep Model

- Pros
- Good generalization and memorization
- Cons
 - Learning deep models is computationally expensive
 - Time consuming prediction method
 - Deep features need to be calculated in prediction time
 - Can't be scaled to RTB size but can be used in sponsored search

Other Interesting Issues





Fraud Detection

Frequency Capping



Attribution

Who we are

- Sponsored Search Advertising
 - Bazaar Search Advertising
- Display Advertising
 - Websites
 - Mobile Applications
- Social Media Advertising
 - Micro Influencer Advertising



media.ad







Tapsell 1st Generation

Business state:

- 500K daily impression
- Video advertising SDK with 50 Publishers
- CPM and CPC campaigns

- Centralized system to answer the requests
- Estimating CTRs using a simple Bayesian Bernoulli Model
- Visualizing the historical data and improve algorithm incrementally
- Cons:
 - Not scalable
 - Large error in CTR estimation
- Pros:
 - Best Performance based advertising platform in its own time



Tapsell 2nd Generation

- Business state:
 - 1M+ daily impression
 - 150+ Publishers
 - CPI Campaign

- Adding multi-level cache to response more requests (still centralized)
- Estimating CVRs in lower granulity
- Adding time effect to the CVR estimation model
- Using feedback data to improve CVR estimations
- Cons:
 - Not scalable
 - Large error in CVR estimation for post-click actions
- Pros:
 - The Only CPI based advertising platform in its own time



Tapsell 3rd Generation

- Business state:
 - 100M+ daily impression
 - 500+ Publishers
 - CPI, CPA Campaign

- Making the model horizontally scalable in all levels
 - Changing the servers' OS to DCOS
 - Switching to distributed programming platforms (Apache Spark)
 - Switching to distributed Databases (Cassandra, ...)
 - Dockerizing all modules
- Making the CVR estimation model much more efficient by considering all users' history
- Pros:
 - The system is completely scalable and there exist no technical limitation to get the market
 - Best Performance based advertising platform in Iran



Tapsell 4th Generation

Business state:

- 200M+ daily impression
- 3500+ Direct Publishers
- About 2x traffic in comparison to 3rd generation

- Decreasing response time to global standards
- Connecting to different ad exchanges through RTB
- Estimating Bid using CVR and other DSPs values
- Pros:
 - Be able to easily increase traffic by connecting to ad exchanges



Current Challenges

Improving CVR estimation method

- We still have a far way to be optimized in CVR estimation
- Improving bid estimation algorithm
 - Bid estimation in competition to other DSPs is still a new challenge for us
- Making the system more scalable and efficient
 - Responding to millions of requests per second with our limited resource is still a dream for us

- Co-op Program for B.Sc. students
 - Learn cutting edge technologies by working in a professional atmosphere
 - Designing, Evaluating and Deploying Large Scale ML Algorithms
 - Distributed Databases and Programming Platforms
 - Cloud Computing technologies
- Research Topic for M.Sc. and Ph.D. students
 - Computational Advertising is a hot topic in top conferences such as KDD, WSDM, WWW, ...
 - Real world problems
 - Real Datasets
 - Baseline Methods that can be used to develop more advanced ones
- Apply for full time or part time job by
 - Send your resume to jobs@tapsell.ir
 - Fill the form at jobs.tapsell.ir



