

## Notes for LIVE session: Ad Click Prediction

1. For cracking programming, DS and Algo rounds in top product based companies:  
InterviewPrep Course: <https://interviewprep.appliedcourse.com/>
2. Real-world problems solving(end to end) in interviews.
3. My experiences at Yahoo! Labs & Amazon.
4. How do we do this session: 2 hour mock interview on Ad-Click prediction for Search.
  - a. I will introduce and explain the business context.
  - b. I will ask questions and wait for your thoughts/responses on Slack.
  - c. 1:n interview/classroom-discussion in which everyone can participate
  - d. I will try to simulate an interview as closely as possible.
  - e. Let's make it as interactive as possible.
  - f. We have seen a ton of code in the course and hence will not prioritize that.
5. Business problem:
  - a. E.g: Google Search
  - b. Pay per click ads.
  - c. Bidding by advertisers: PPC, Search keywords, Ad-creative.
  - d. Search engine ranking: Maximize profit (objective) while keeping the results relevant.
  - e. Real world constraints: ??
  - f. ML problem: predict the probability of click:  $P(\text{Click} | \text{ad}, \text{user}, \text{context})$
  - g. Expected Revenue =  $p\text{Click} * \text{bid}$
  - h. Position bias.
  - i. Typical CTR: ~4%.
6. Mapping to an ML problem:
  - a. What type of problem: classification with probabilistic outputs.
  - b. Objective: Min Log-loss
  - c. ML constraints: low-latency, interpretability, parallelizable training,
  - d. KPI/metric: log-loss, AUC (if you only care about ranking and not pClick values)
7. Dataset:
  - a. Search Logs
  - b. Format: each row corresponds to an ad impression:  $\langle x_i = \text{features}, y_i = \text{isClick?} \rangle$
  - c. How to correct for Position bias?
  - d. Features: user, ad, context, historical rates, geo,
  - e. Encoding of each feature: categorical features, Numeric features, ....
  - f. Dataset properties: imbalanced data: undersampling, reweighting-schemes.
  - g. Time-based-splitting, Demographic split,

## 8. Modeling?

- a. Choice of models: NB, Logistic regression,
- b. NEXT WEEK  
??
- c. Measurement of model performance: ??
- d. Modeling at Scale: ??

## 9. Productionisation & monitoring

- a. Choice of architecture: ??
- b. Challenges: ??
- c. Monitoring: ??
- d. Retraining :??
- e. Real-time/Online training: ??
- f. Cold-start:??
- g. State of the art @ Google: <https://ai.google/research/pubs/pub41159>

## 10. Real world data and Code:

- a. <https://www.kaggle.com/c/kddcup2012-track2>
- b. <https://drive.google.com/drive/folders/1yRpfqSxojPwmPHVJXSMf-hC8EiOs8hTn?usp=sharing>

## 11. External References:

- a. Google: "ad click prediction"
- b. Google Scholar: "ad click Prediction"  
[https://scholar.google.co.in/scholar?hl=en&as\\_sdt=0%2C5&q=ad+click+prediction&btnG=](https://scholar.google.co.in/scholar?hl=en&as_sdt=0%2C5&q=ad+click+prediction&btnG=)
- c. <https://towardsdatascience.com/mobile-ads-click-through-rate-ctr-prediction-44fdac40c6ff>
- d. [https://turi.com/learn/gallery/notebooks/click\\_through\\_rate\\_prediction\\_intro.html](https://turi.com/learn/gallery/notebooks/click_through_rate_prediction_intro.html)