# AusNOG 2016

1 & 2 September 2016, Swissotel, Sydney

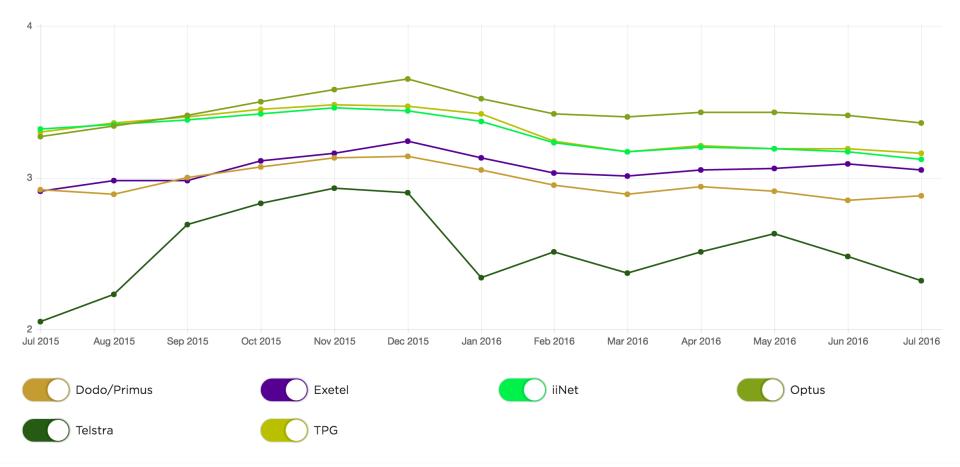


Rethinking Broadband Performance using Big

**Data from M-Lab** 

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# How do you think of your ISP?



## This Talk

- Overview: Data, Speed, user test pattern
- Problems:
  - Various Network Variables effect on Speeds; different distribution for different ISP
  - Sampling bias
- Solution:
  - Consider all affecting variables:
    - Casual inference model
  - De-Biasing
- Conclusion





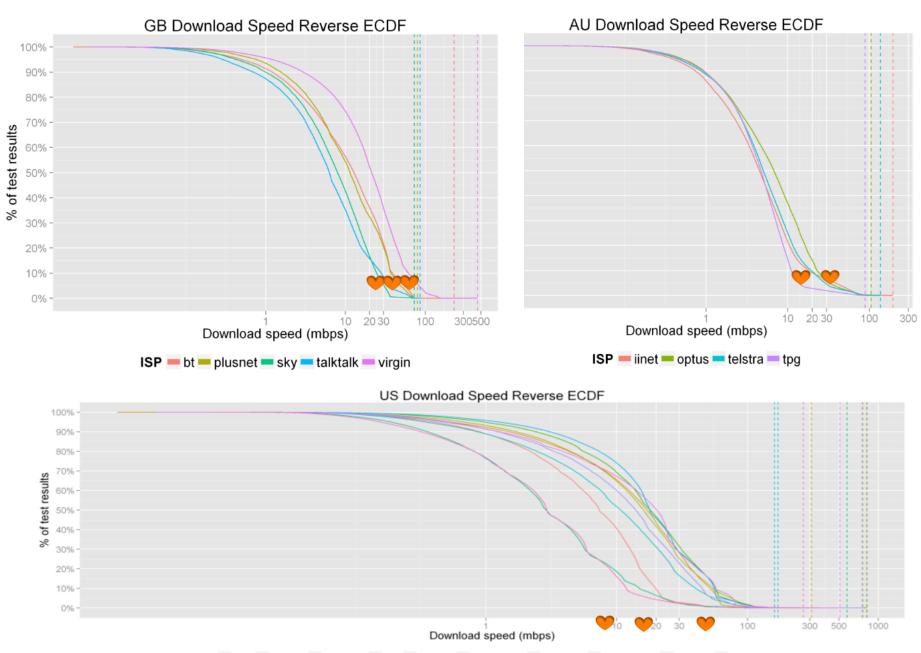
# NDT does:

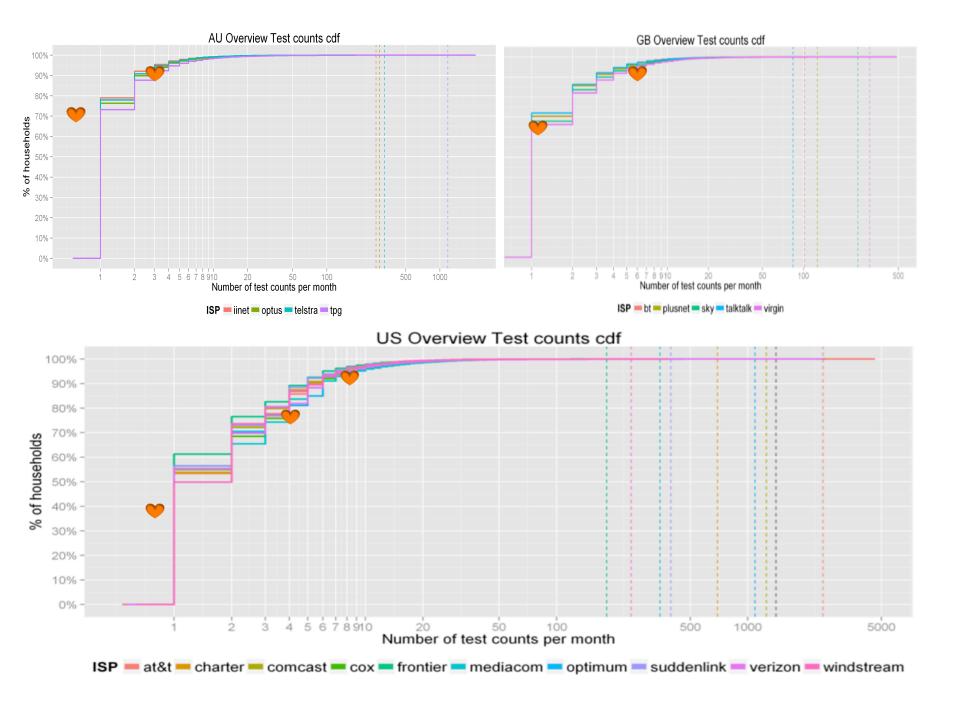
- TCP performance / speed test:
- Record TCP web100 variables during one test session: RTT, loss, MSS, Congestion signal counts, ECN..

Country	Amount of test results (2015)	Number of households
Australia	313090 0.3M	* Estimated by distinctive IP addresses 163854 0.16 M
UK	1012925 <u>1</u> M	457486 0.4 M
USA	3625154 3.6M	967141 0.9 M

Amount of NDT Data in 2015







#### Frequent households' influence



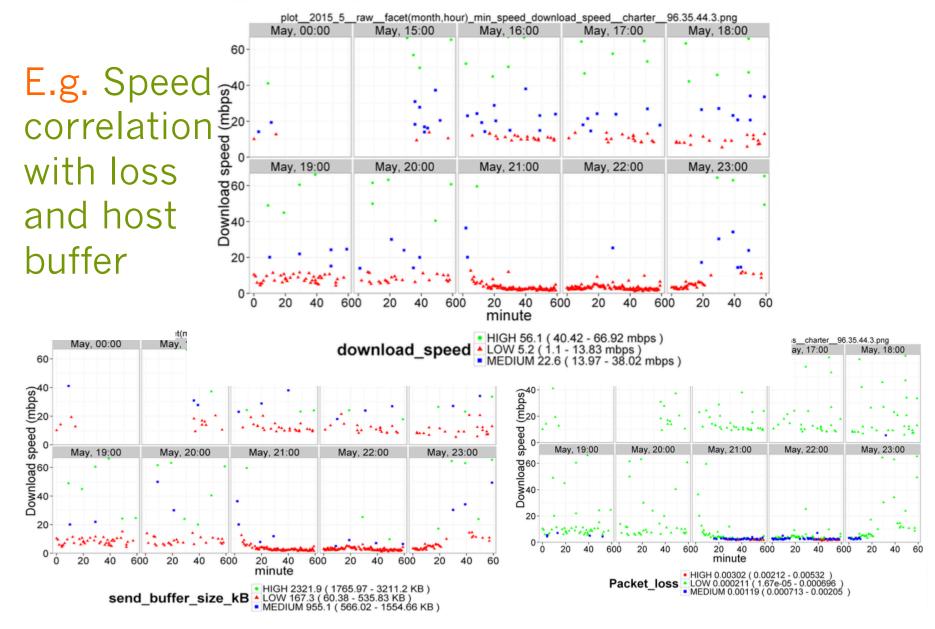
Various Network
 Variables effect on
 Speeds



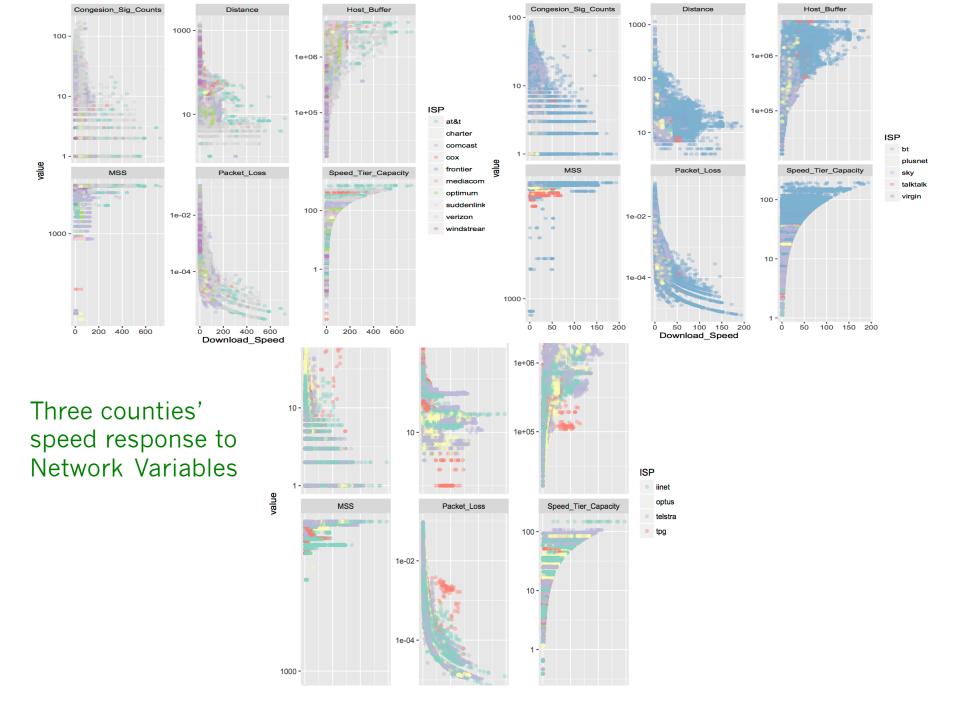
Network Variables(configurations) affects Speed

- User subscribed speed tier \*estimated
- User host's configuration : TCP send buffer
- TCP MSS
- Client Server Distance
- IP address Family (IPv4/IPv6):
  - Data not currently available with NDT
- Time of a day





http://104.154.87.31/static/

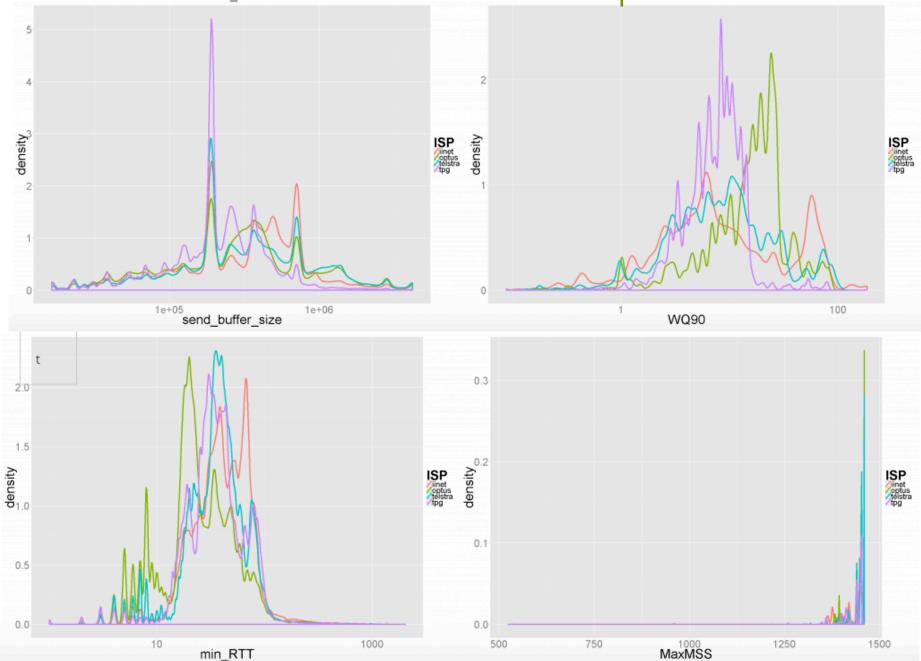


OPERATION ROBLEMS FIN MOTH HELP

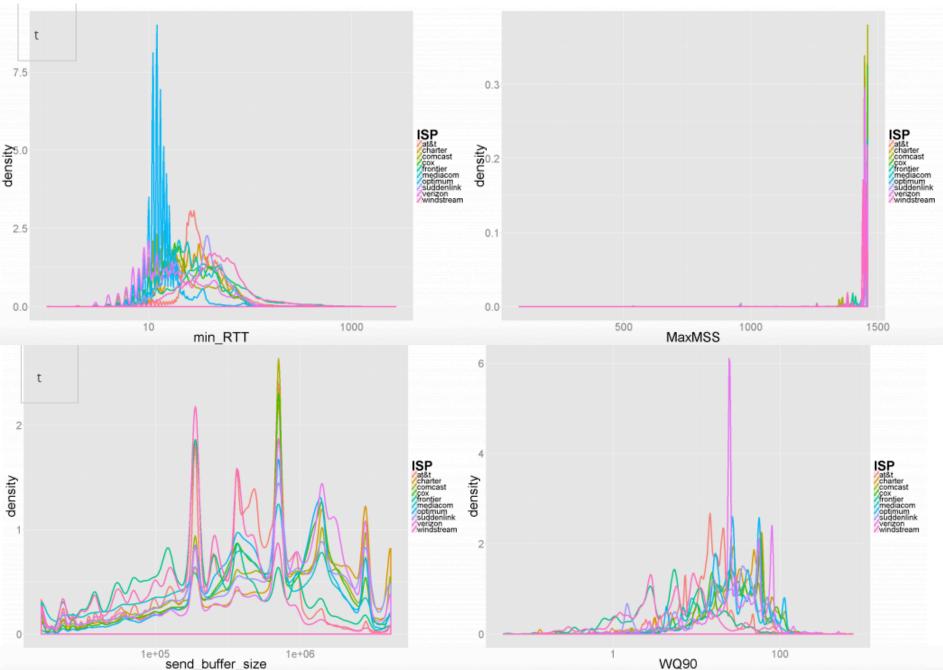
1) Mis-matched Network Variable distribution

2) Sampling Bias

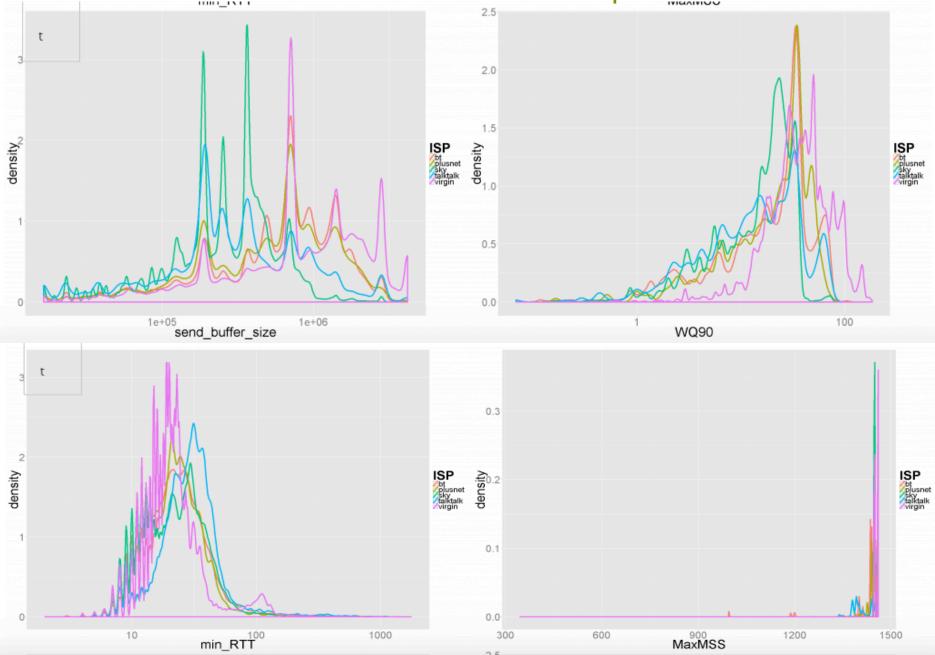
#### AU ISPs' Network Variables Comparison



#### US ISPs' Network Variables Comparison



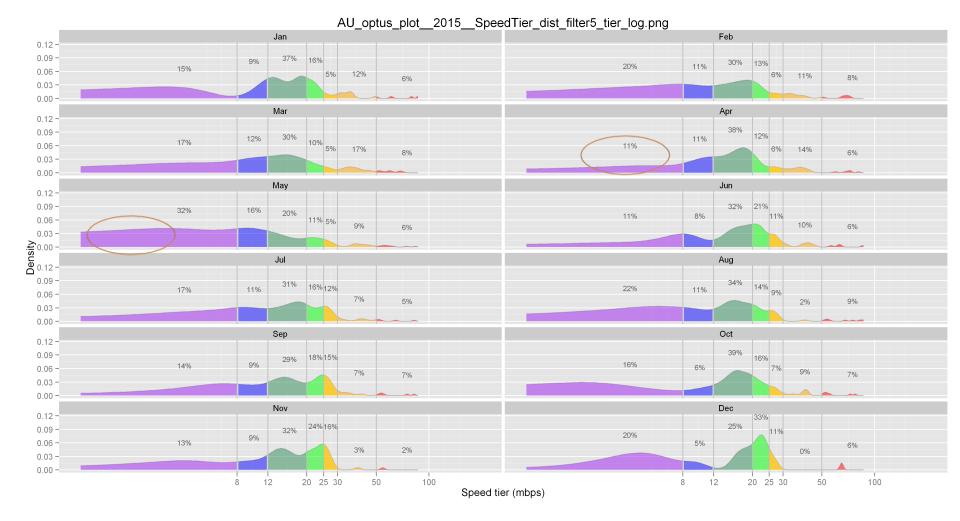
# UK ISPs' Network Variables Comparison

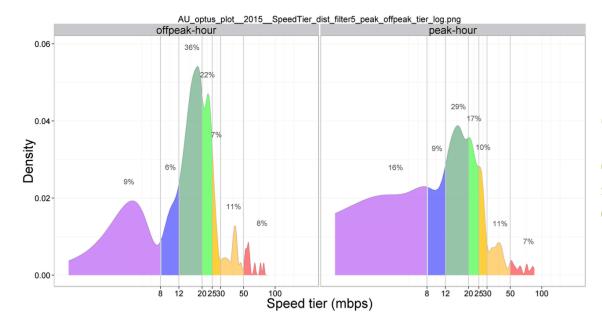




# SamplingBias

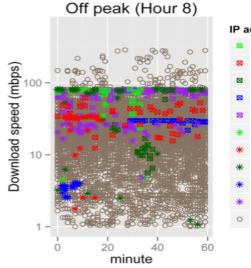
## Monthly fluctuation caused by sampling bias





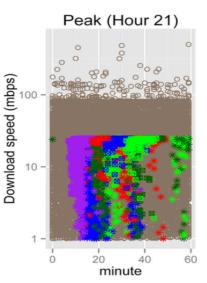
#### Peak hour v.s None peak hour sampling bias – User/ client behavior pattern

Observed low speed tier users from some ISPs performing More tests during peak hours



#### IP address

- 108.41.184.161 (20 tests)
- 108.51.193.79 (37 tests)
- 108.53.36.26 (26 tests)
- 173.62.186.69 (33 tests)
- 173.64.216.148 (79 tests)
- \* 71.102.60.230 ( 24 tests )
- \* 71.173.136.27 (35 tests)
  \* 71.187.61.61 (46 tests)
- \* 71.240.222.193 (28 tests)
- \* 72.69.218.34 ( 42 tests )
- o thers (3682)

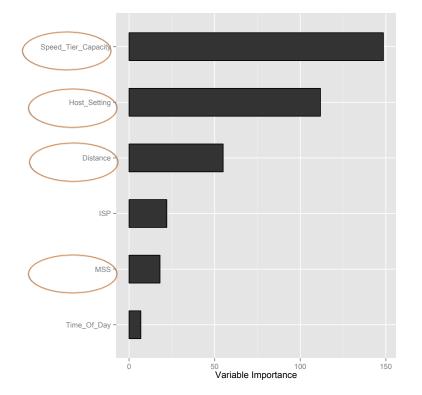


#### IP address

- 100.2.204.154 (583 tests)
- 100.7.6.222 (942 tests)
- 108.29.174.169 (687 tests)
- 108.35.83.174 (441 tests)
- 108.36.239.164 (1292 tests)
- \* 108.5.97.27 ( 925 tests )
- \* 173.61.92.67 (1082 tests)
- \* 173.68.137.30 (1273 tests)
- \* 96.226.159.106 (648 tests)
- \* 96.245.129.56 (1394 tests)
- o thers (9344)

#### Verizon year 2015

### ISPs' Network Variables Importance\*



#### \*Variable importance (VIMP)

is the difference between OOB prediction error before and after permutation,

a large VIMP value indicates that misspecification detracts from the variable predictive accuracy.

#### Random Forests for Regression

1. % variance explained: 80.16, A good modle fit

# 2. Time of Day has little affect or Speeds

$$VI^{(t)}(\mathbf{x}_{j}) = \frac{\sum_{i \in \overline{\mathfrak{B}}^{(t)}} I\left(y_{i} = \hat{y}_{i}^{(t)}\right)}{\left|\overline{\mathfrak{B}}^{(t)}\right|} - \frac{\sum_{i \in \overline{\mathfrak{B}}^{(t)}} I\left(y_{i} = \hat{y}_{i,\pi_{j}}^{(t)}\right)}{\left|\overline{\mathfrak{B}}^{(t)}\right|}$$

 $\hat{y}_{i}^{(t)} = f^{(t)}(\mathbf{x}_{i})$  = predicted class before permuting

 $\hat{y}_{i,\pi_j}^{(t)} = f^{(t)}(\mathbf{x}_{i,\pi_j}) = ext{predicted class after permuting } X_j$ 

 $\mathbf{x}_{i,\pi_j} = (x_{i,1}, \dots, x_{i,j-1}, x_{\pi_j(i),j}, x_{i,j+1}, \dots, x_{i,p})$ 

Note:  $VI^{(t)}(\mathbf{x}_j) = 0$  by definition, if  $X_j$  is not in tree t

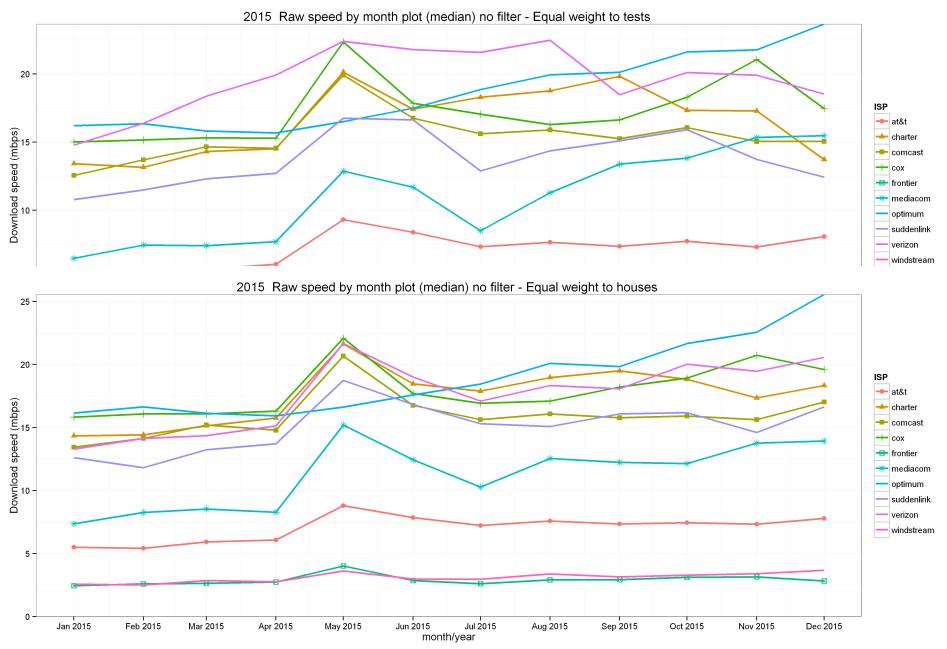
The permutation importance



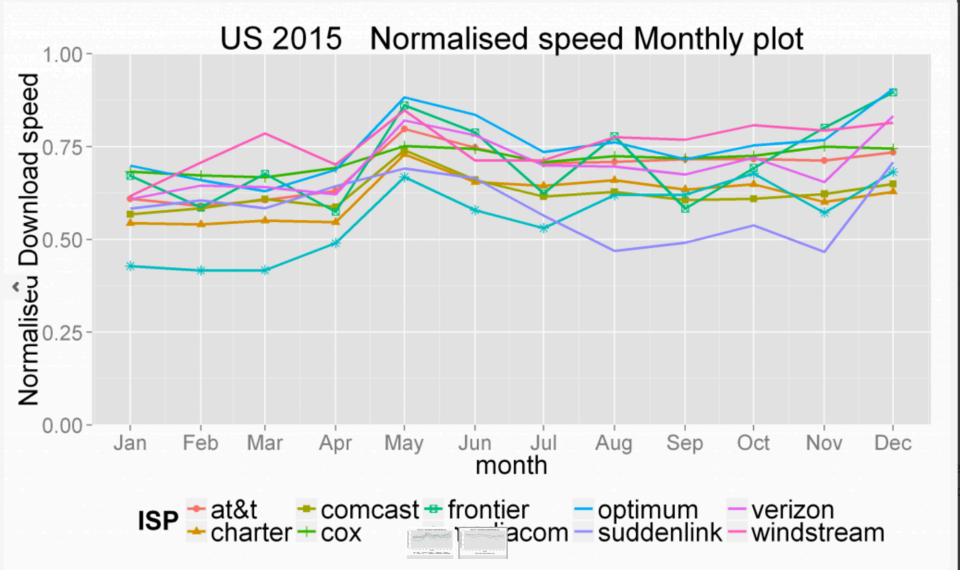
### 1.De-Biasing: Equal weight to household

2. Casual Inference model: Matching tests with similar network variables between two ISPs

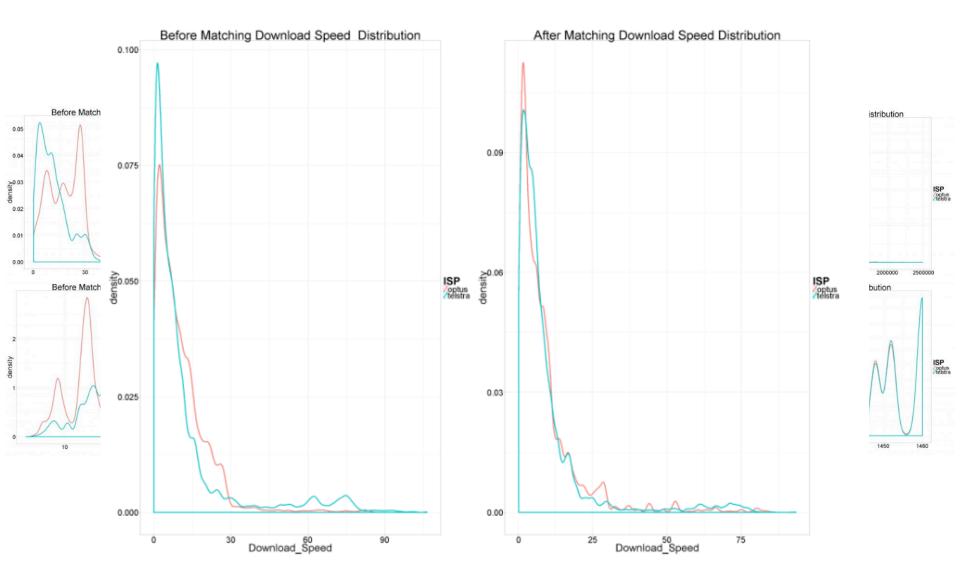
## Equal weight on household applied on US data



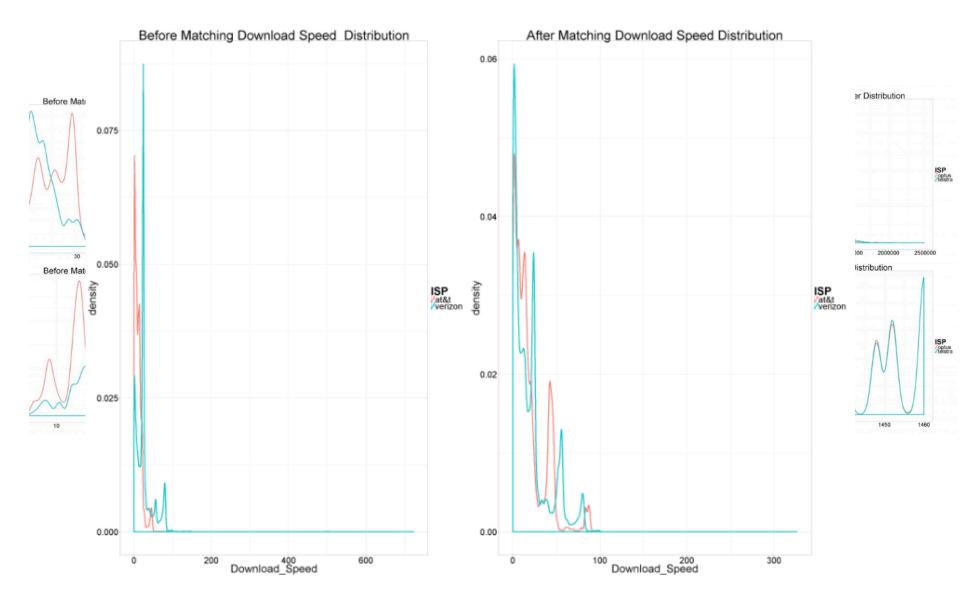
# Normalisation: another aggregated performance indicator



#### Before and After Matching: A pair of AU ISP



#### Before and After Matching: A pair of US ISP



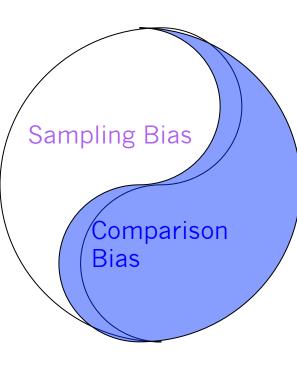
## Before and After Matching: A pair of UK ISP





ISP	Raw difference	Data10 ATE (95%CI)	Data20 ATE (95%CI)
ISP	Average	Matched Data10	Matched Data20
Telstra vs. Optus	1.5 Slower	0.6 Faster	No Difference
Telstra vs. linet	0.9 Faster	No Difference	No Difference
Telstra vs. Tpg	0.2 Faster	No Difference	No Difference
Optus vs. linet	2.4 Faster	3.7 Slower	2.9 Slower
Optus vs. Tpg	1.7 Faster	0.9 Slower	1.4 Slower
Tpg vs. linet	0.7 Faster	1.1 Slower	1.2 Slower
Tpg vs. linet	0.71	-1.14 ( -1.71 , -0.57 )	-1.21 ( -1.99 , -0.43 )





#### Insight gained: **Comparison Bias**

1) Simple Averaging Lead to misleading ISP rankings to Internet users

2) More sophiscated methods such as Statistical tools & Machine Learning shall be used, for a fairer ISP performance comparison.

**RESULTS:** A fairer comparison often shows a different ISP ranking – sometimes even opposite to ranking based on average

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Insight gained:

- 1) Aggregated average number is impacted by sampling bias that caused by user behaviors.
- 2) Sampling Bias can cause ISP monthly fluctuation.
- 3) Sampling Bias opens doors to gaming the system. Warning

#### SOLUTIONS

- a) Equal weight to household than test
- b) Casual Inference Model, Bayes addictive Tree Model (future work)





static/



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