

A: Datasheet

Algorithm: tevia<sub>n</sub>\_006

Developer: Tevian

Submission Date: 2021\_04\_16

Template size: 1032 bytes

Template time (2.5 percentile): 596 msec

Template time (median): 597 msec

Template time (97.5 percentile): 630 msec

Investigation:

Frontal mugshot ranking 84 (out of 329) — FNIR(1600000, 0, 1) = 0.0024 vs. lowest 0.0009 from sensetime\_006

Mugshot webcam ranking 51 (out of 291) — FNIR(1600000, 0, 1) = 0.0111 vs. lowest 0.0057 from sensetime\_006

Mugshot profile ranking 29 (out of 260) — FNIR(1600000, 0, 1) = 0.1230 vs. lowest 0.0550 from sensetime\_006

Immigration visa–border ranking 38 (out of 218) — FNIR(1600000, 0, 1) = 0.0026 vs. lowest 0.0009 from sensetime\_006

Immigration visa–kiosk ranking 18 (out of 215) — FNIR(1600000, 0, 1) = 0.0707 vs. lowest 0.0487 from cubox\_000

Identification:

Frontal mugshot ranking 38 (out of 329) — FNIR(1600000, T, L+1) = 0.0098, FPIR=0.001000 vs. lowest 0.0017 from nec\_005

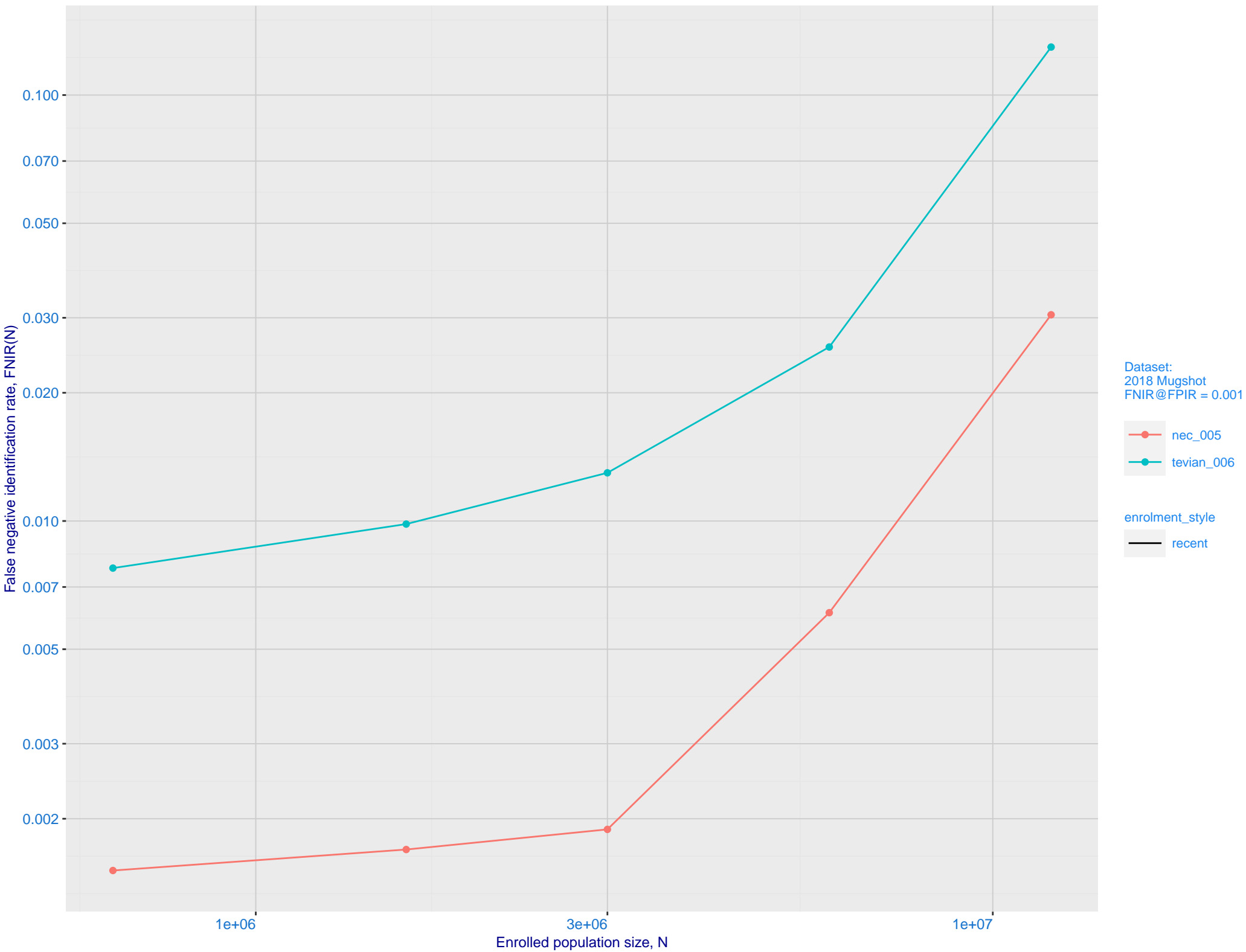
Mugshot webcam ranking 32 (out of 289) — FNIR(1600000, T, L+1) = 0.0318, FPIR=0.001000 vs. lowest 0.0120 from nec\_005

Mugshot profile ranking 13 (out of 259) — FNIR(1600000, T, L+1) = 0.4266, FPIR=0.001000 vs. lowest 0.1331 from cloudwalk\_hr\_000

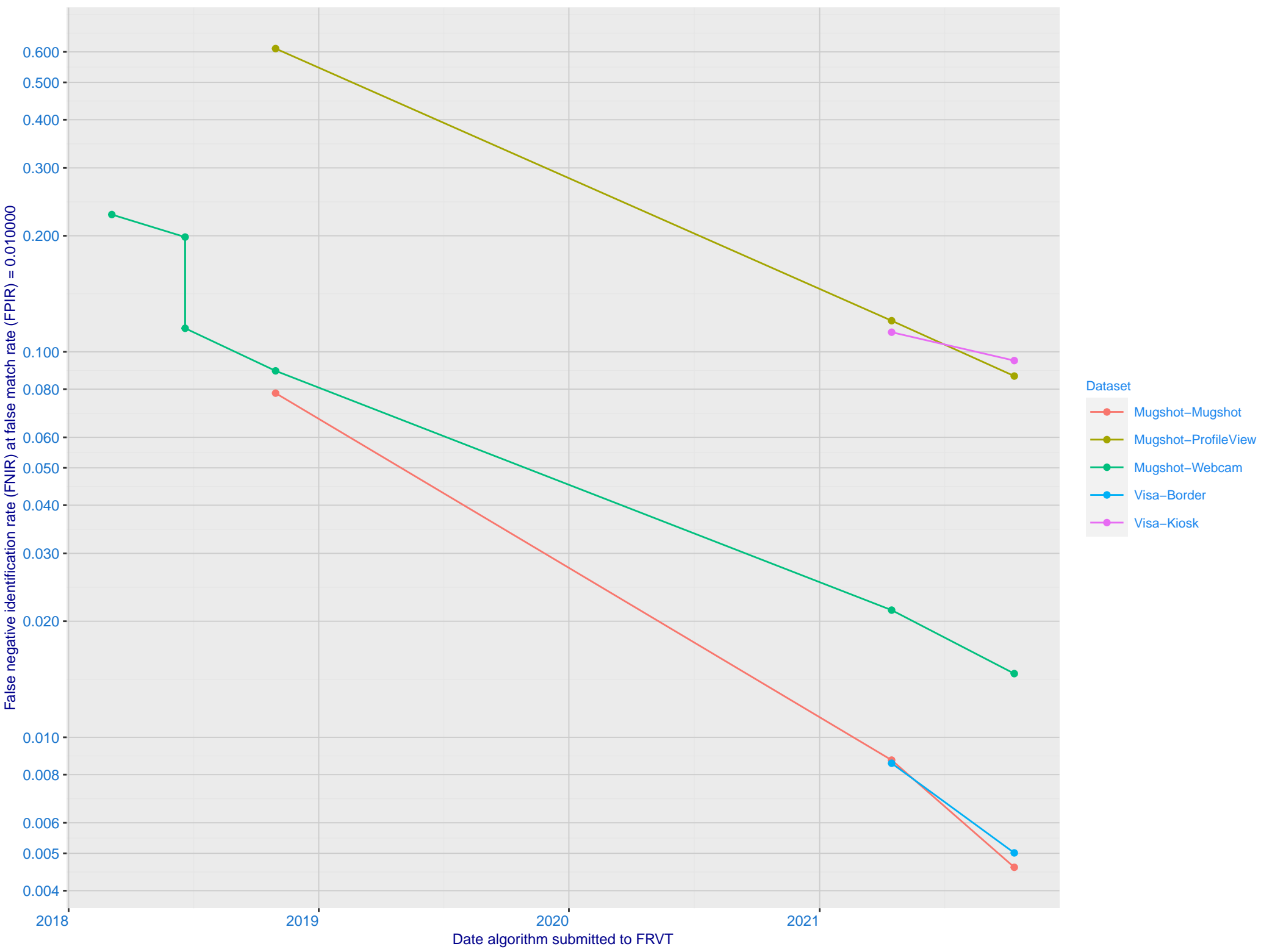
Immigration visa–border ranking 37 (out of 217) — FNIR(1600000, T, L+1) = 0.0164, FPIR=0.001000 vs. lowest 0.0032 from paravision\_009

Immigration visa–kiosk ranking 172 (out of 212) — FNIR(1600000, T, L+1) = 0.9675, FPIR=0.001000 vs. lowest 0.0728 from paravision\_009

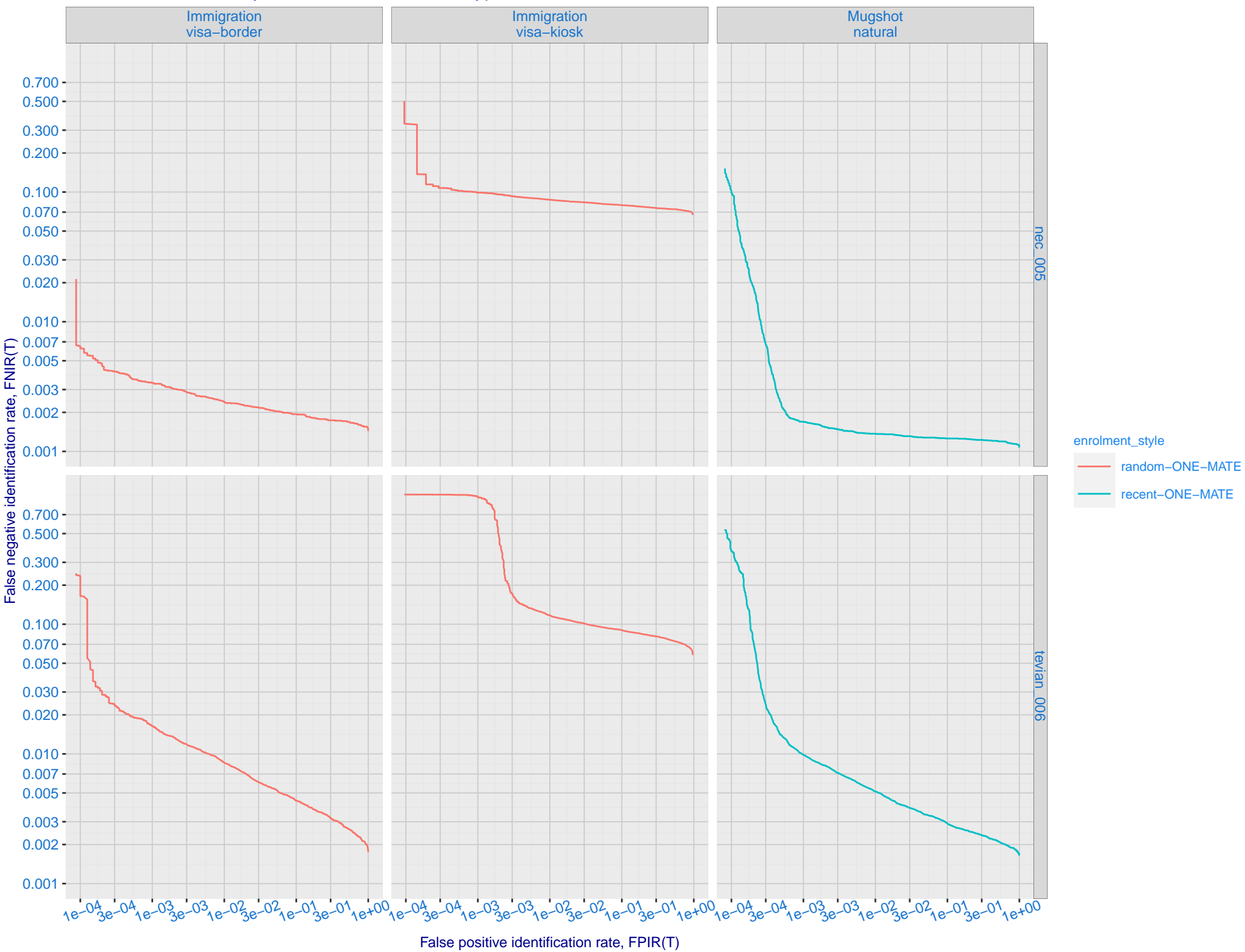
B: Mugshot natural images, identification mode: FNIR(N, L+1, T) vs. most accurate (nec\_005)



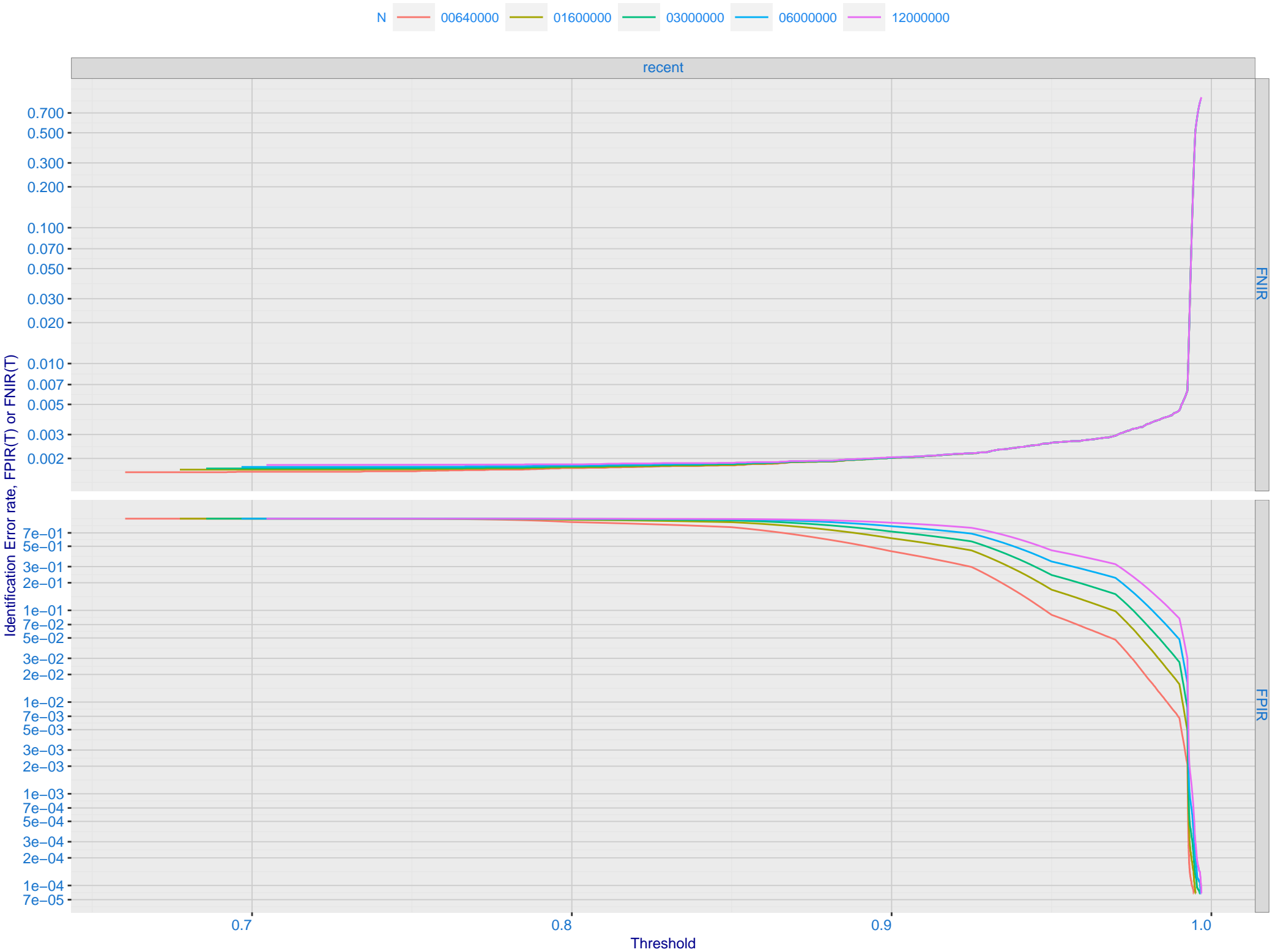
C: Evolution of accuracy for TEVIAN algorithms on three datasets 2018 – present



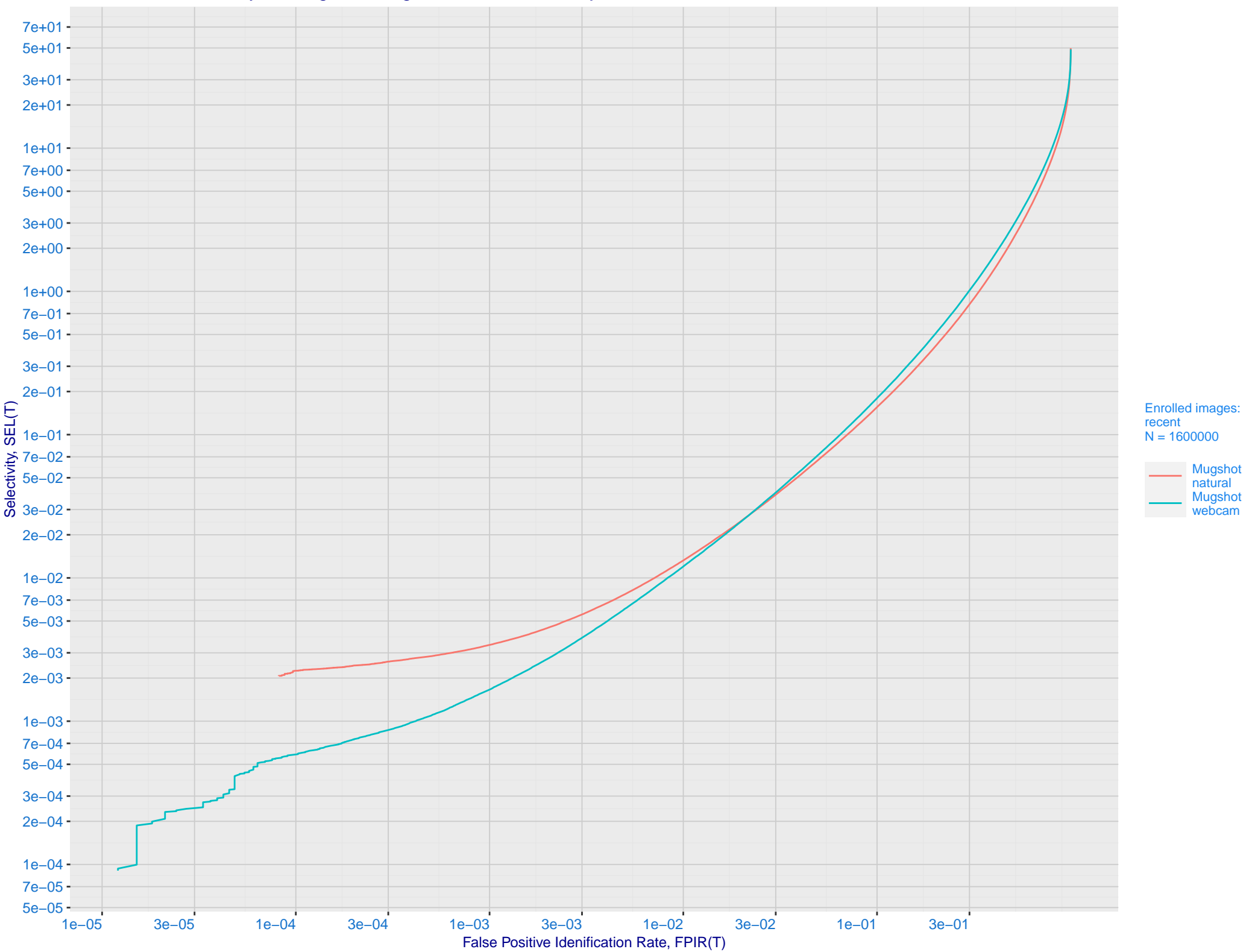
D: 1:N error tradeoff by dataset and enrollment type. N = 1600000 individuals



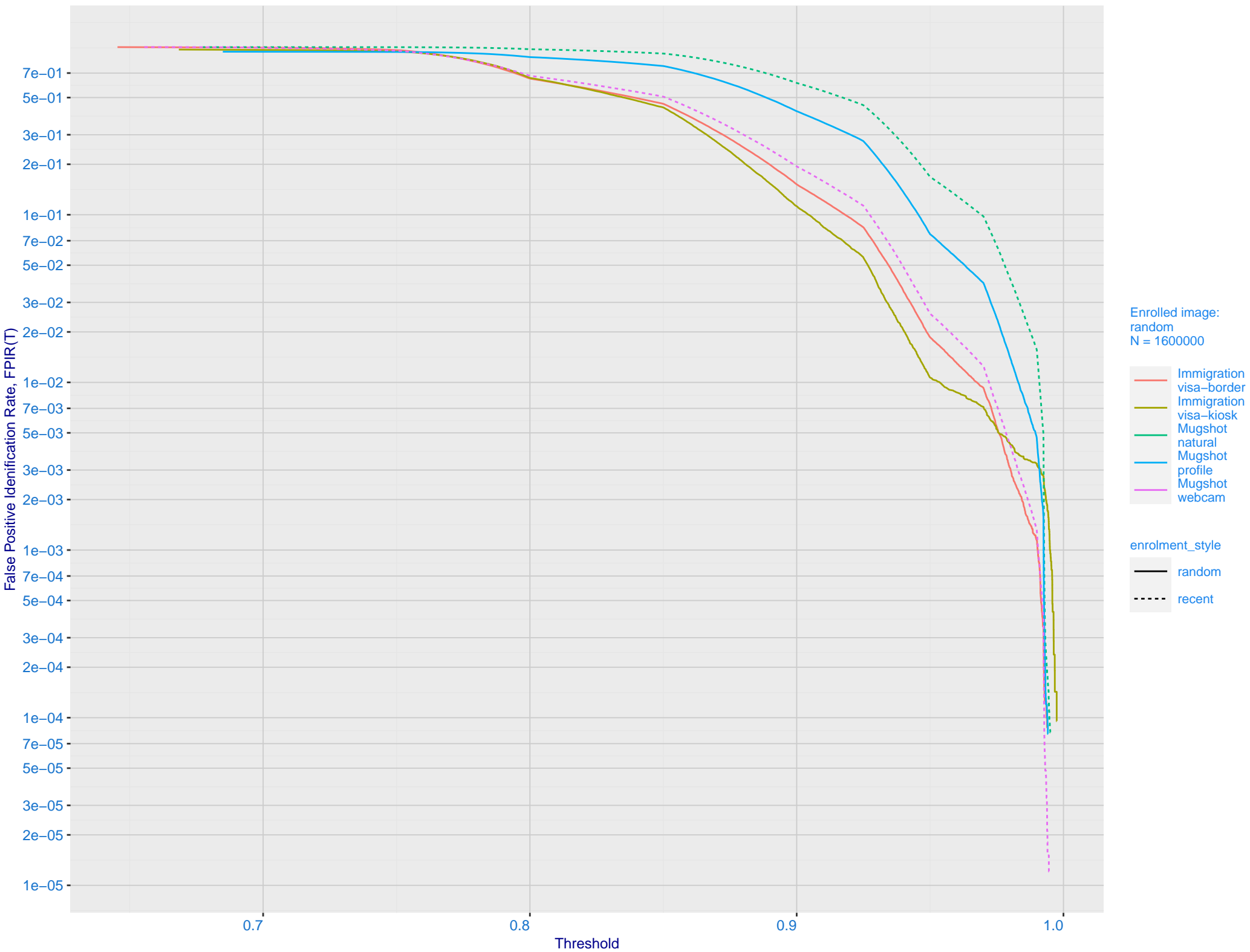
E: Dependence of error rates on T by number enrolled identities, N, for Mugshot natural images



F: FPIR vs. Selectivity for mugshot images, N = 1600000 subjects enrolled with one recent mate

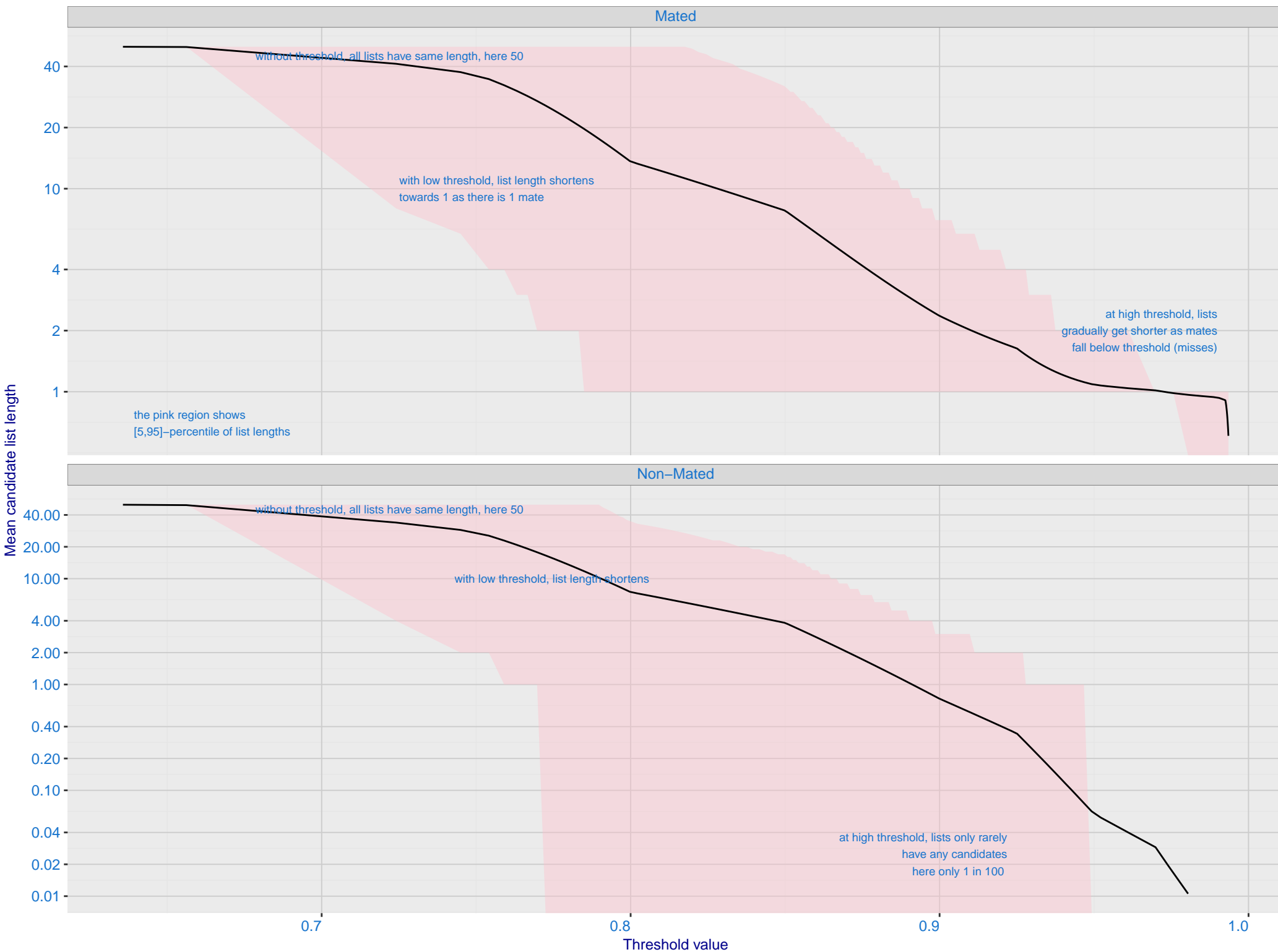


G: FPIR dependence on T by probe type for N = 1600000 subjects



# H: Reduced length candidate lists for human review

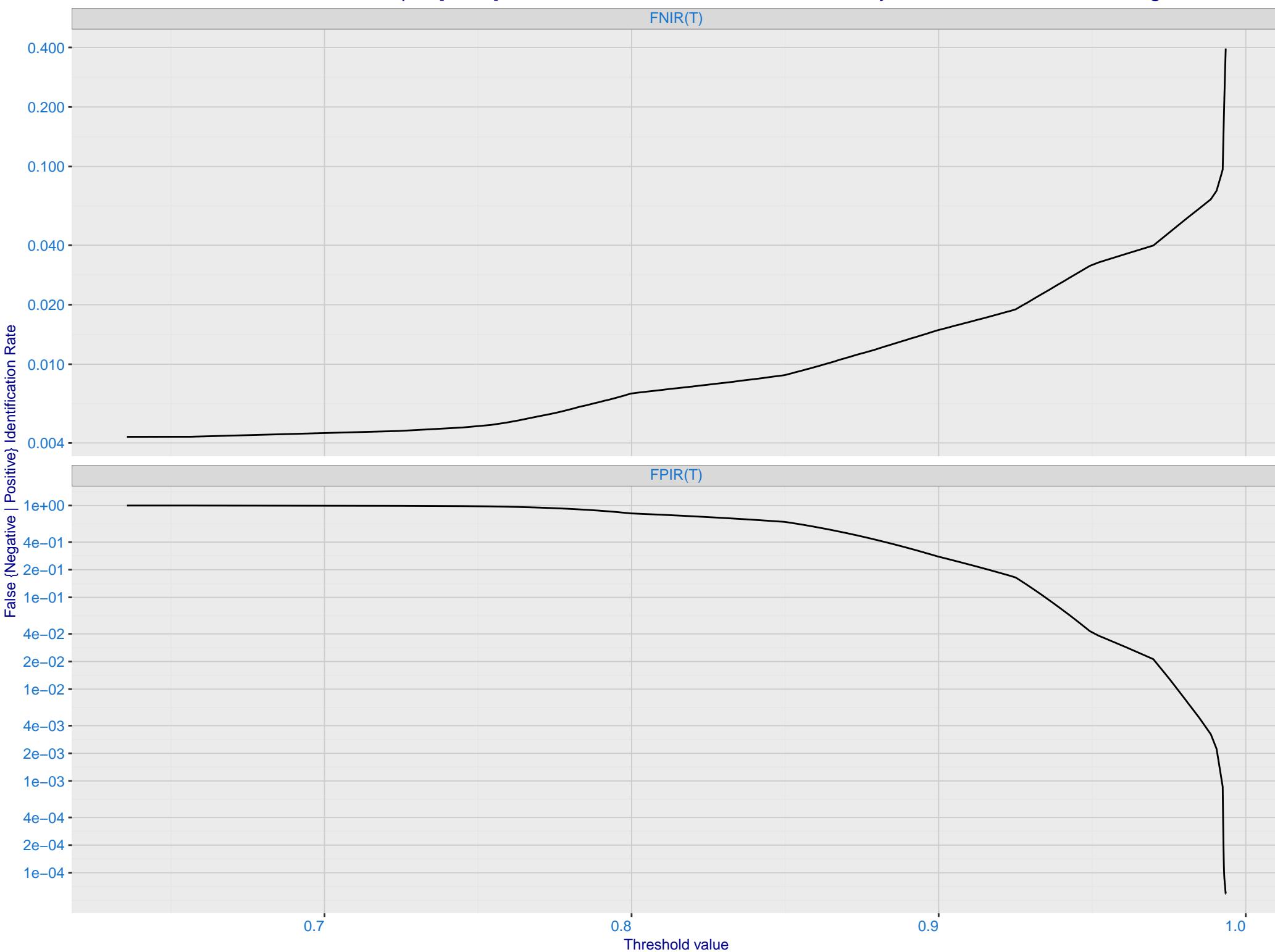
Dataset is border–border with time–lapse [10,15] YRS with N = 1600000. Probes are 10–15 years later than enrollment image



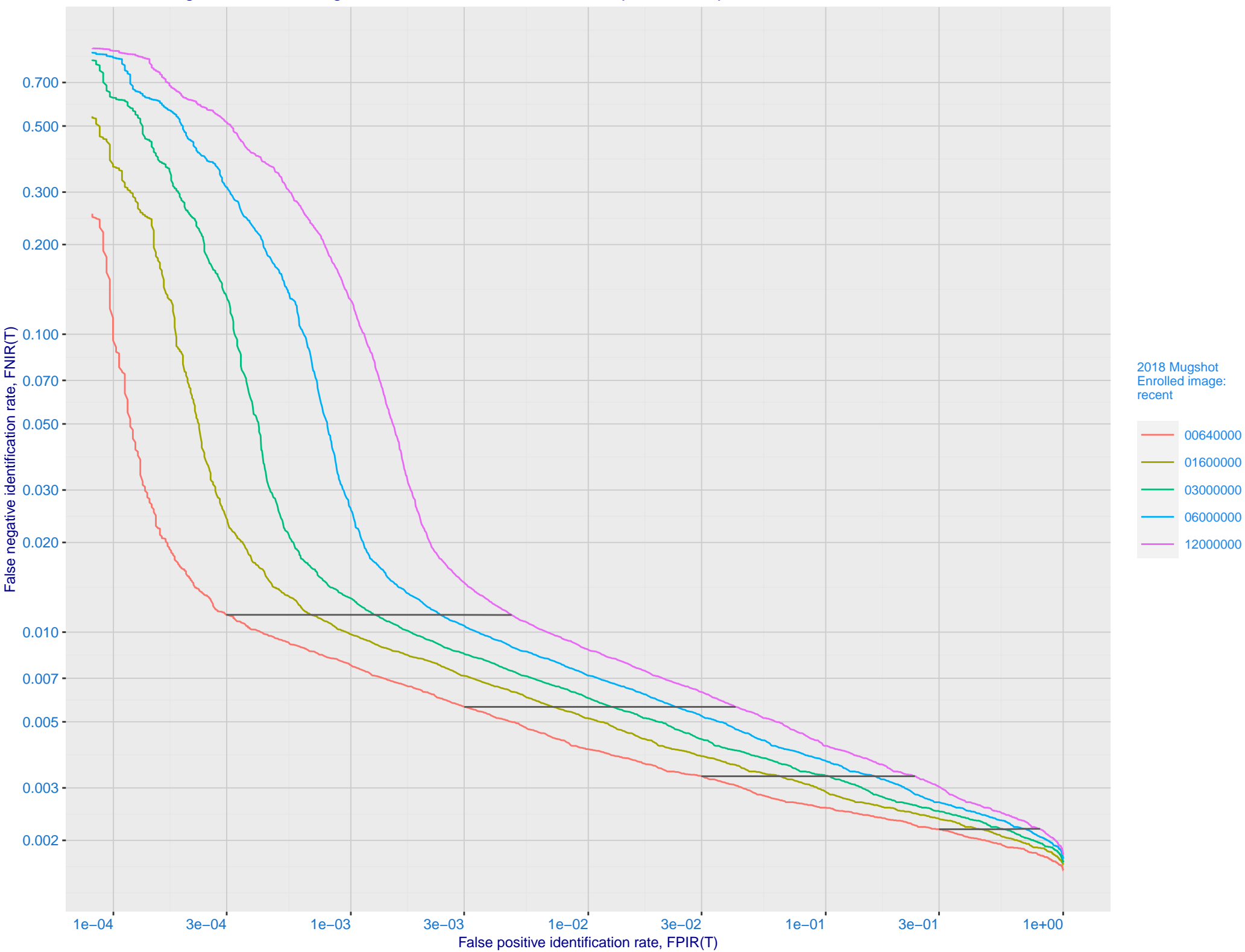


# I: FNIR and FPIR dependence on threshold

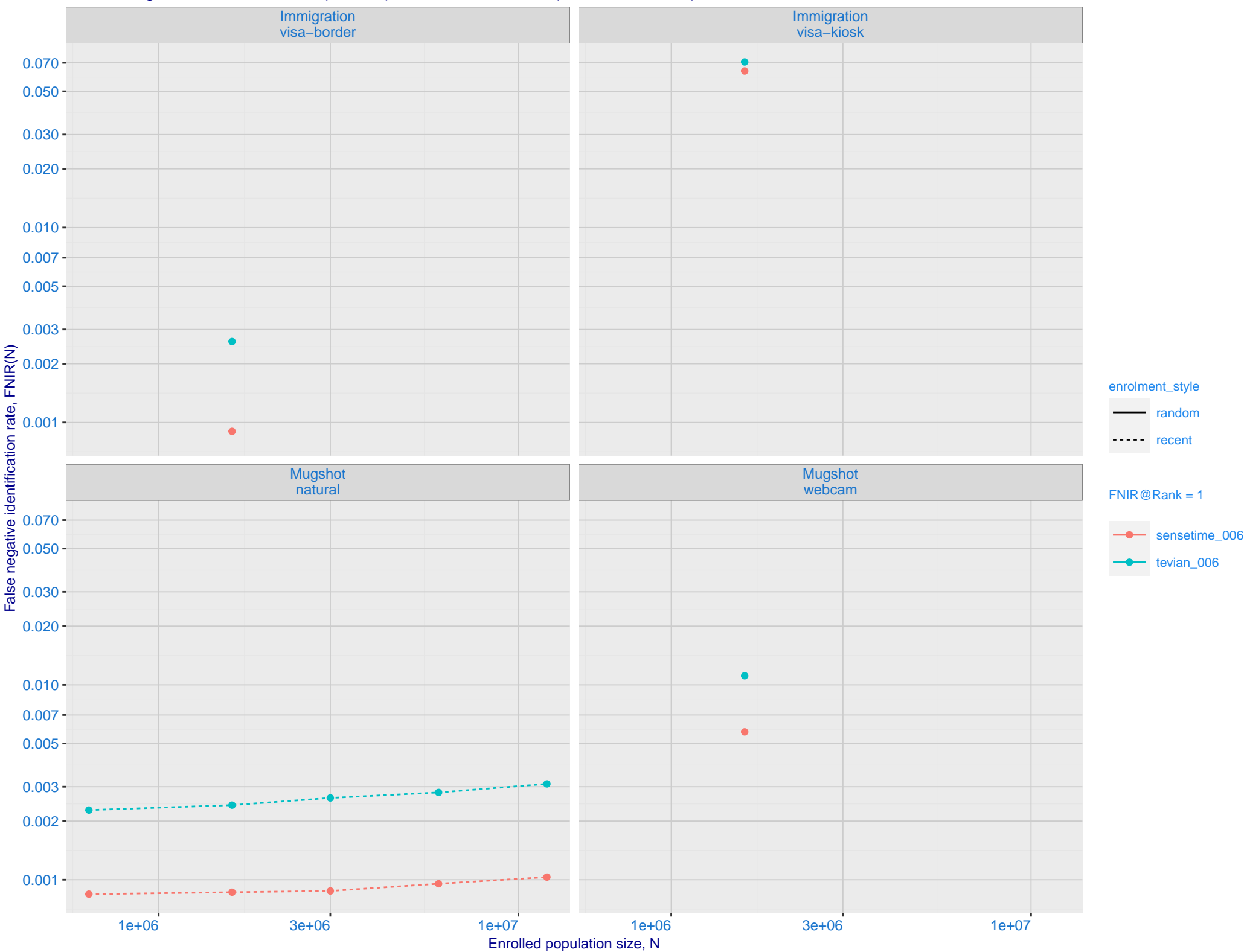
Dataset is border-border with time-lapse [10,15] YRS with N = 1600000. Probes are 10-15 years later than enrollment image



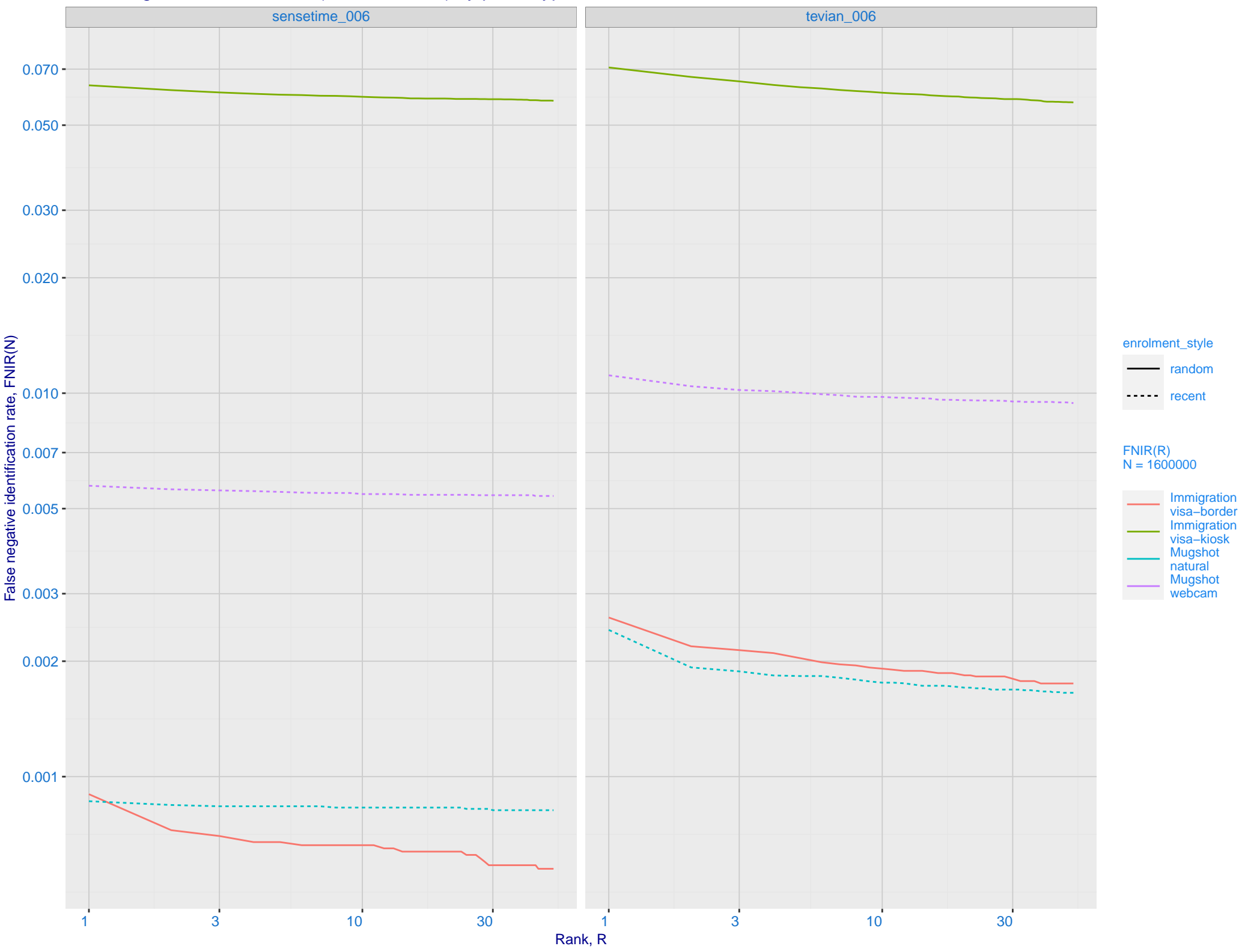
J: DET for Mugshot natural images and various N. Links connect points of equal threshold.



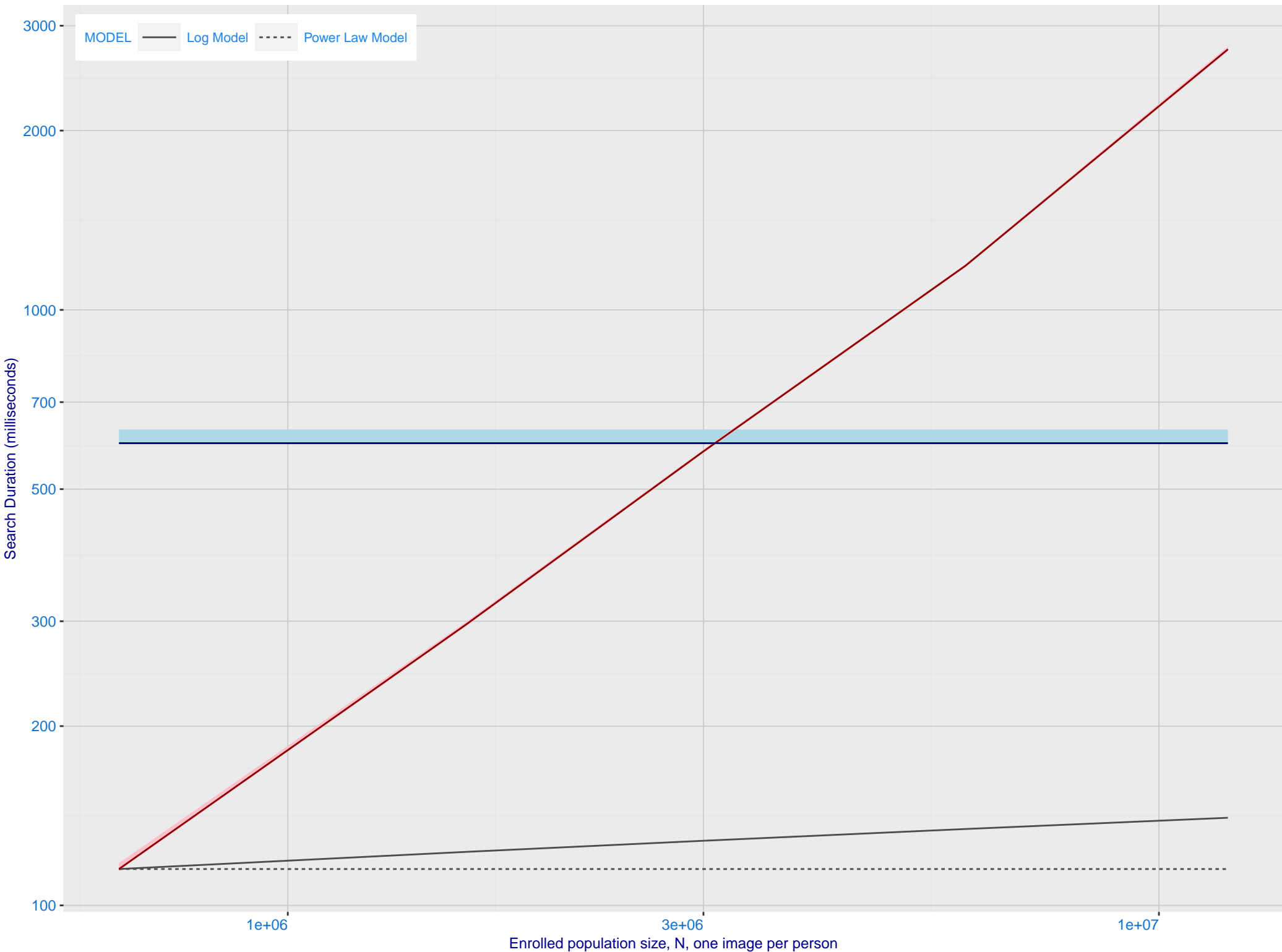
K: Investigational mode: FNIR(N, 1, 0) vs. most accurate (sensetime\_006)



L: Investigational mode: FNIR(1600000, R, 0) by probe type

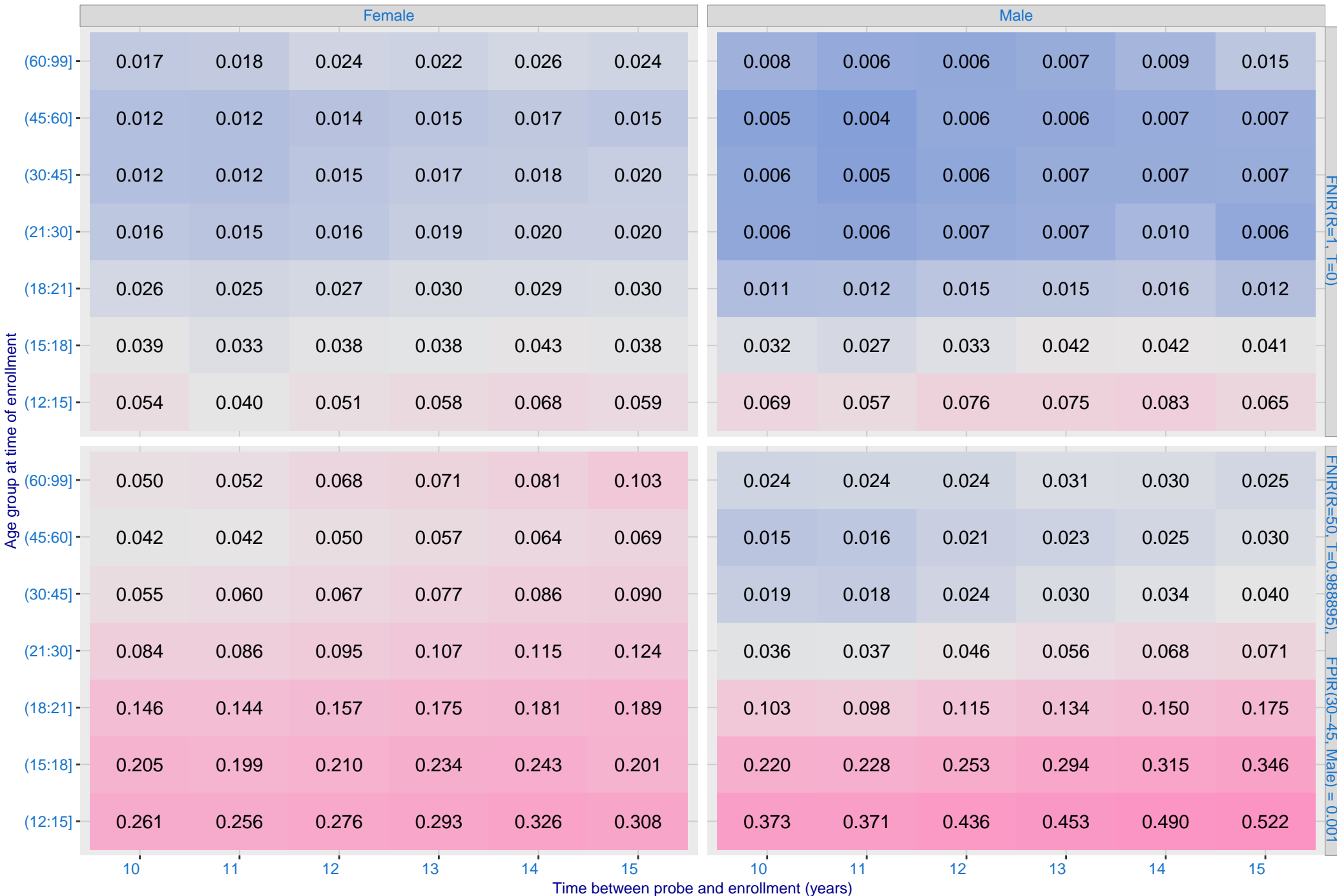


M: Template duration; search duration vs. N. The blue and pink ribbon covers 95 percent of observed measurements. The template generation time is independent of N. The log and power-law models are fit to the first two (N,T) observations



O: FNIR(T, N = 1.6 million) by sex, age and time-lapse. The top row gives investigational rank-1 miss rates. The bottom panels give high threshold for more lights-out identification with low FPIR.

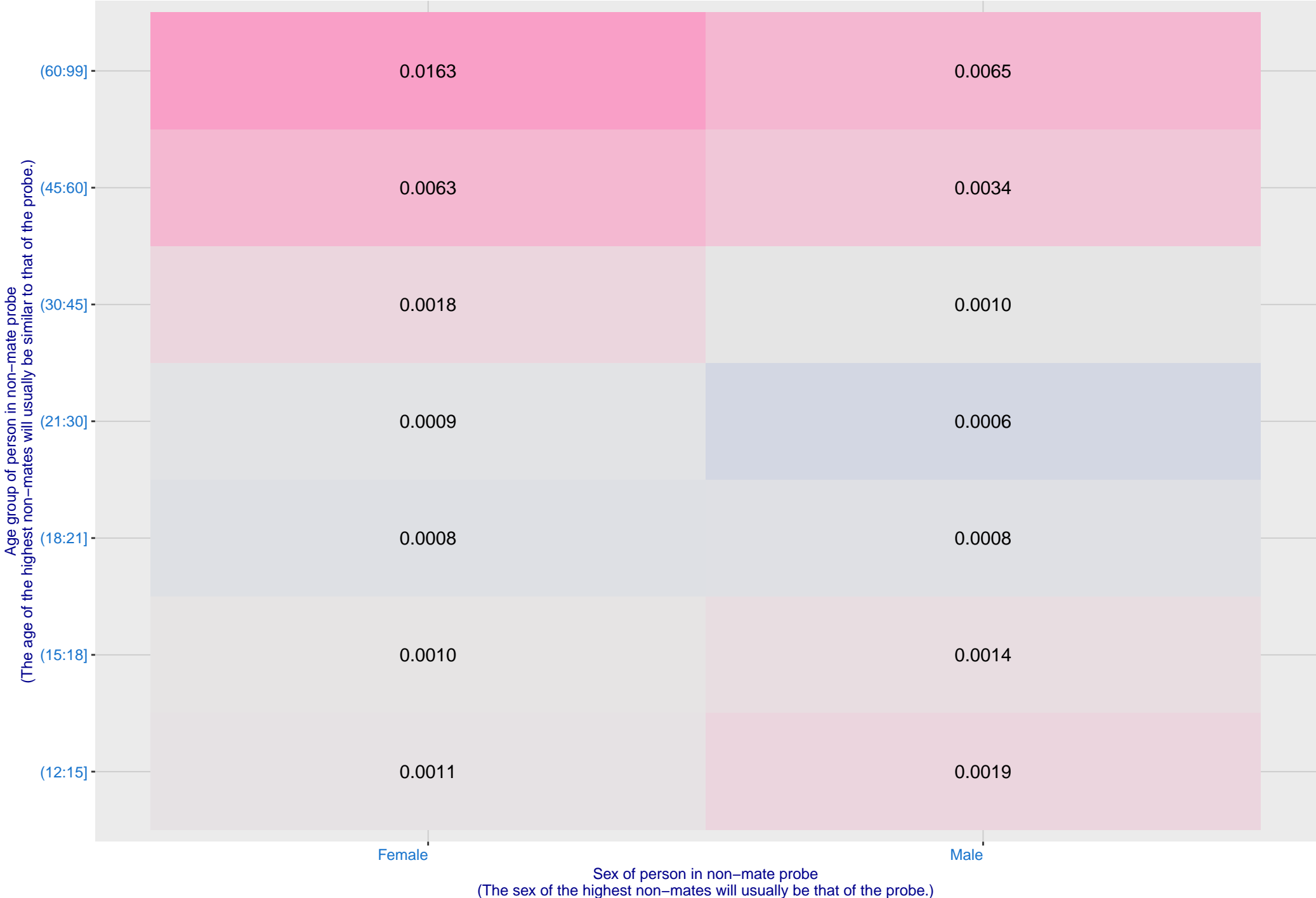
Algorithm: tevia\_n\_006, Dataset: Border-Crossing Ageing N = 1600000  
Text encodes FNIR, Color encodes log(FNIR)



P: FPIR(N = 1.6 million) by sex and age. It is typical for false positive identification rates to be higher in women except in their teens.

Algorithm: tevian\_006, Dataset: Border-Crossing Ageing  
Threshold: 0.988895 set to achieve FPIR(30–45, Male) = 0.001

Color encodes log(FPIR)



Q: Identification FNIR(N, T, L+1) and Investigational FNIR(N, 0, R) under ageing

Dataset: 2018 Mugshot N = 3068801

