A run a day won't keep the hacker away: Inference Attacks on Endpoint Privacy Zones in Fitness Tracking Social Networks

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Running is enjoying a boom because of the coronavirus pandemic



SPORT SEPTEMBER 23, 2020 / 1:03 AM / UPDATED 10 MONTHS AGO

Exclusive: Brits on bikes as fitness app data shows pandemic boom

By Allen Kim, CNN

Updated 0953 GMT (1753 HKT) April 25, 2020

Bloomberg

The Pandemic Bike Boom Hits in Some Unexpected American Cities

Los Angeles and Houston are hardly cycling capitals. But both saw surges in biking after Covid-19 began, according to new data from the fitness app Strava.

By Laura Bliss

September 23, 2020, 3:00 PM GMT+2

By Kate Kelland

2 MIN READ



Fitness apps grew by nearly 50% during the first half of 2020, study finds

15 Sep 2020

Carmen Ang Reporter, Visual Capitalist

BBC

Fitness app Strava lights up staff at military bases

() 29 January 2018

CNN

Garmin is slowly coming back online after a massive ransomware hack

By Oliver Effron, CNN Business

Updated 1937 GMT (0337 HKT) July 27, 2020

The Washington Post

Democracy Dies in Darkness

Fitness app Polar revealed not only where U.S. military personnel worked, but where they lived

By Rebecca Tan

July 18, 2018 at 10:00 a.m. UTC

Cycling

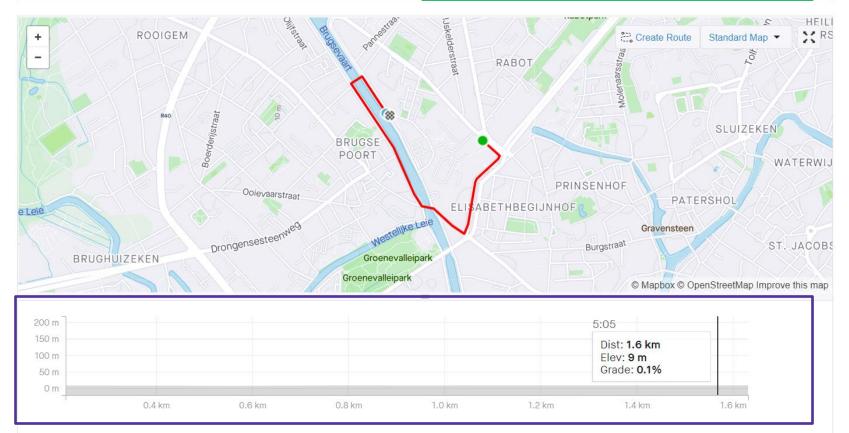
Strava removes automatic flybys after safety concerns

The ride-tracking app has now made the comparison feature opt-in

BY ALEX BALLINGER OCTOBER 15, 2020

Fitness Tracking Social Networks: Activities

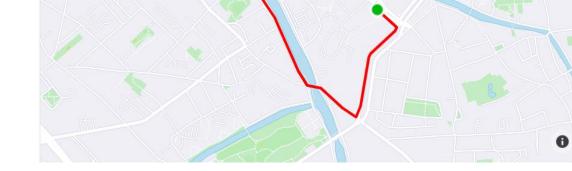
Strava User – Ride	Give Kudos 🖉 0 🗐 0	*
Thursday, May 20, 2021 · Ghent, Flanders Evening Ride	1.87 km 5:55 Om Distance Moving Time Elevation	
	AvgMaxSpeed19.0km/h20.9km/hElapsed Time5:55	



4

Endpoint Privacy Zones





+

View of owner of activity

View of user that doesn't own activity

[1] Hassan et al. Analysis of Privacy Protections in Fitness Tracking Social Networks -or- You can run, but can you hide? In USENIX (2018)

[2] GRUTESER et al. An onymous usage of location-based services through spatial and temporal cloaking. In Proceedings of the 1st international conference on Mobile systems, a pplications and services (2003)



> Threat model

- \rightarrow capabilities of *regular* user
- → only based on *public* (meta)data

- > Two subproblems:
 - 1. Discovering EPZs
 - 2. Finding protected location inside EPZ

Attack: Discovering EPZs



repeat

assign each endpoint to closest fitted circle of cluster

Isq fit new circle for cluster

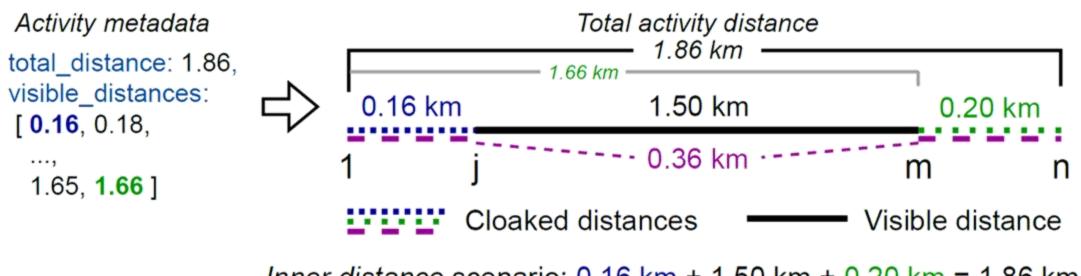
until convergence criterium is met



Attack: Protected Location Inside EPZ

> Two scenarios:

1. Inner Distance 2. Total Distance



Available distances: Inner distance scenario: 0.16 km + 1.50 km + 0.20 km = 1.86 km Total distance scenario: 0.36 km + 1.50 km = 1.86 km

Inner Distance Scenario

Strava User – Ride		Give Kudos 🖉 0 🗐 0 🔺	
Thursday, May 20, 2021 - Ghent, Flanders Evening Ride	1.87 km 5:55 Om Distance Moving Time Elevation		
	Avg Speed 19.0km/h Elapsed Time 5:55	Max 20.9km/h	
pageViewstreams	.streamData.data		
<pre>▶ grade_smooth: ▶ latlng: (31)</pre>	(31) [0, 12 -0.3, -1.4, -0.5, Array(2), Ar ay(2), Array(2), I, 56, 57, 58, 61, 64, 81, 84, ORI ELI:ABETHBEGIJNHOF ELI:ABETHBEGIJNHOF Burgstras	-0.4, -0.4, -0.3, -0.1, 0, Array(2), Array(2), Array(, 496.2, 551.9, 607.3, 645.8, 699.2, 737.4, 780.4, 797.7, 844.3, 851.1, 901.8, 982.4, 100 0.3, 0.5, 0.6, 0.7, 0.5, 0.4, -0.3, -0.3, 0, 0, 0.2, 0.2, 0, 0, 0, 0, 0.1, 0.1, 0.1, 0, 2), Array(2), Array(2), Array(2), Array(2), Array(2), Array(2), Array(2), Array 0, 148, 151, 160, 162, 171, 186, 202, 208, 226, 244, 263, 272, 280, 292, 305, 307, 309]
		e mapped a perior cerial inprove and map	
	n 1.0 km 1.2 km 1.4	9 m	
m m m 0.4 km 0.6 km 0.8 h	Dist: 1. Elev: 9 Grade:	9 m 0.1%	Inner distance scenario
	Dist: 1. Elev: 9 Grade:	0.1% 4 km 1.6 km	Inner distance scenario Distance covered inside EPZ leaked

Total Distance Scenario

> distance covered inside EPZ = total distance – track distance





> Two scenarios: 1. Inner Distance 2. Total Distance

	Total Distance Attack	Inner Distance Attack
Strava	\checkmark	\checkmark
Garmin Connect	\checkmark	
Komoot	\checkmark	
Map My tracks	\checkmark	\checkmark
Map My Run	\checkmark	
Ride With GPS	\checkmark	\checkmark

Attack: Finding Protected Locations Inside EPZ Intuition of attack

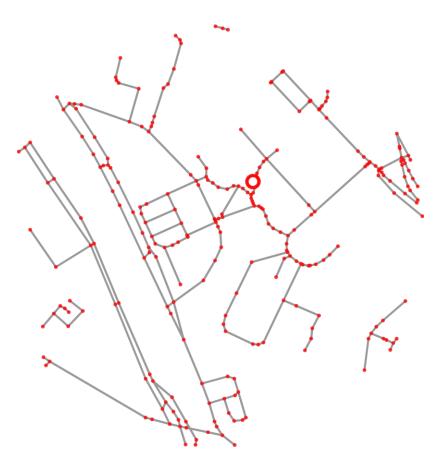


Attack: Finding Protected Locations Inside EPZ

Intuition of attack



Attack: Finding Protected Locations Inside EPZ Preprocessing

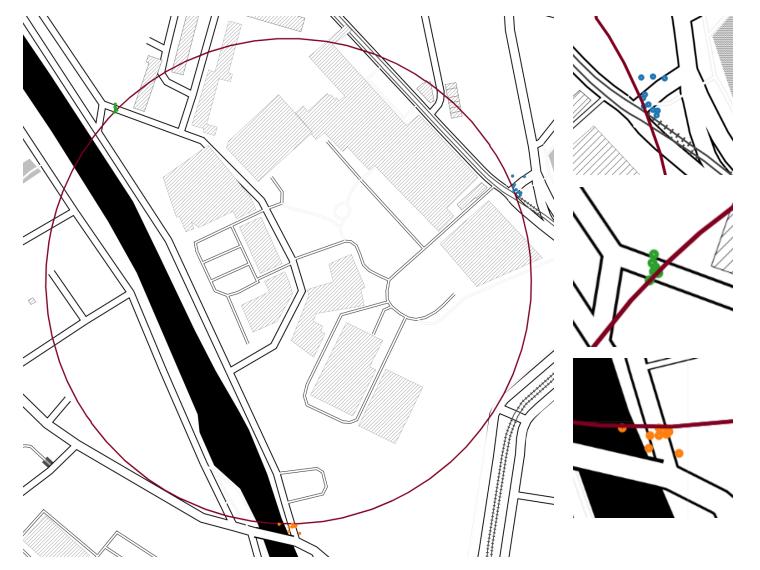


Downloaded road graph

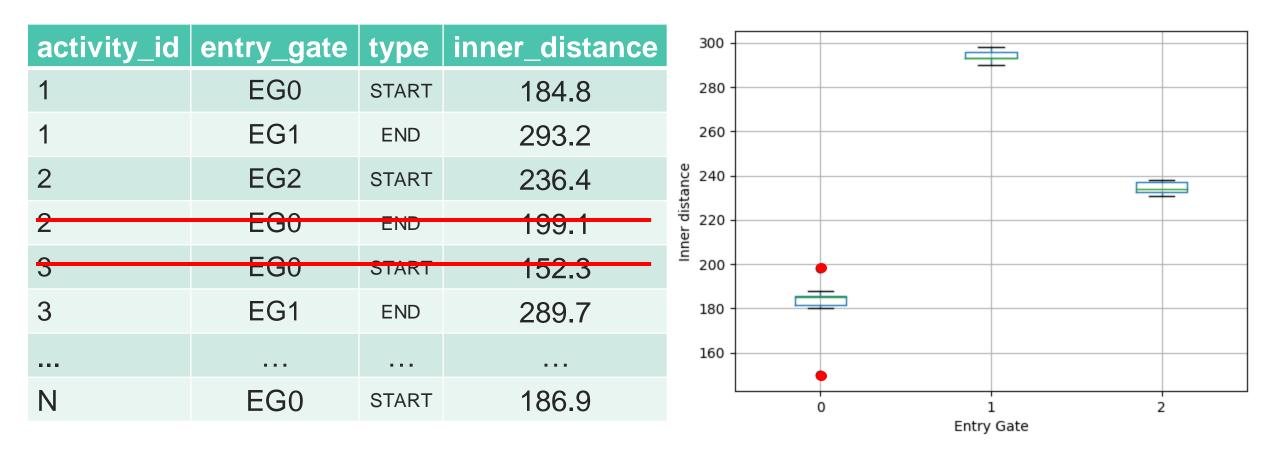


Node resolution increased through chaining

Attack: Finding Protected Locations Inside EPZ Identifying Entry Gates



Attack: Finding Protected Locations Inside EPZ Filtering outliers



Attack: Finding Protected Locations Inside EPZ Predicting Location

- > For each node of interpolated road graph:
 - LAD fit of *N* observed distances and *M* theoretical distances

activity_id	entry_gate	type	EPZ_distance
1	EG0	START	184.8
1	EG1	END	293.2
2	EG2	START	236.4
3	EG1	END	289.7
Ν	EG0	START	186.9

Observed Activity Distances

node_id	EG_0	EG_1	EG_2
0	$d_{0,0}$	$d_{0,1}$	<i>d</i> _{0,2}
1	$d_{1,0}$	$d_{1,1}$	<i>d</i> _{1,2}
2	<i>d</i> _{2,0}	<i>d</i> _{2,1}	<i>d</i> _{2,2}
3	$d_{3,0}$	<i>d</i> _{3,1}	<i>d</i> _{3,2}
Μ	$d_{M,0}$	$d_{M,1}$	$d_{M,2}$

Theoretical Distances

Attack: Finding Protected Locations Inside EPZ

Predicting Location

Constructing Confidence Intervals

activity id

activity_id	entry_gate	type	inner_distance
1	EG0	START	184.8
1	EG1	END	293.2
2	EG2	START	236.4
2	EG2	START	236.4
Ν	EG0	START	186.9

entry gate

EG0

EG1

EG1

activity_id	entry_gate	type	inner_distance	
1	EG0	START	184.8	
1	EG1	END	293.2	
2	EG2	START	236.4	-
3	EG1	END	289.7	
Ν	EG0	START	186.9	

 EG0
 START
 184.8

 ...
 ...
 ...

 EG0
 START
 186.9

START

END

END

inner distance

184.8

293.2

293.2

Observed Activities

activity_id	entry_gate	type	inner_distance
1	EG0	START	184.8
2	EG2	START	236.4
2	EG2	START	236.4
3	EG1	END	289.7
N-1	EG0	START	185.3

. . .



Confidence Interval

Resamples

Privacy Metrics



Success: prediction within threshold of GT

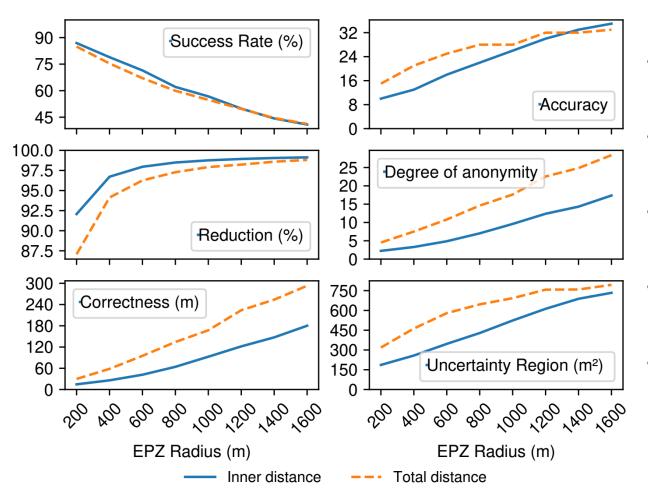
Accuracy: # unique predicted locations

Reduction: Accuracy / # locations inside EPZ

Correctness: avg distance between predictions and GT

> Uncertainty region: joint area around predictions

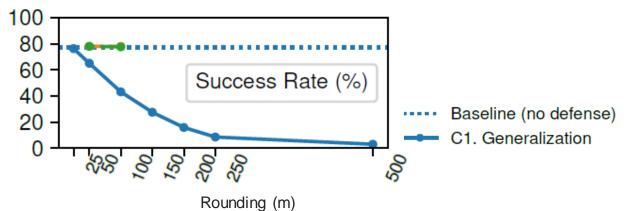
Results



- **Success:** prediction within threshold of GT
- Accuracy: # unique predicted locations
- **Reduction:** Accuracy / # locations inside EPZ
- **Correctness:** avg distance between predictions and GT
- **Uncertainty region:** joint area around predictions

Data minimization

- » "What you don't have, you can't leak"
- » (On-device) Generalization

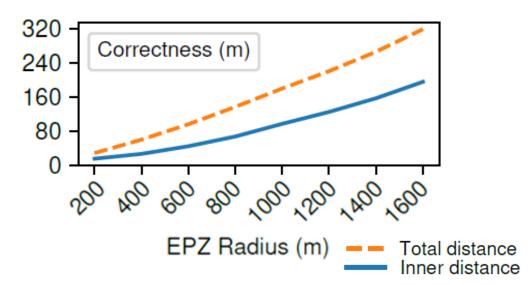


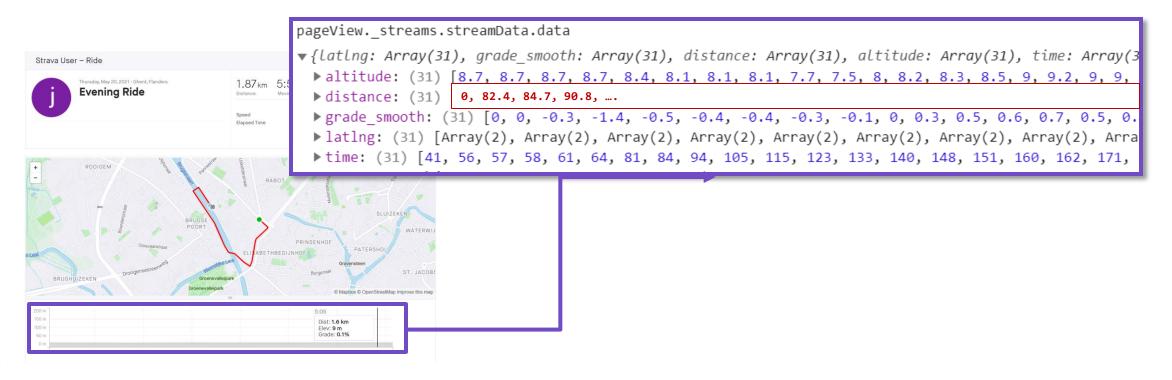
- » Truncation
 - Trade-off with usability: activity gets shorter

Reflect on data minimization at design time

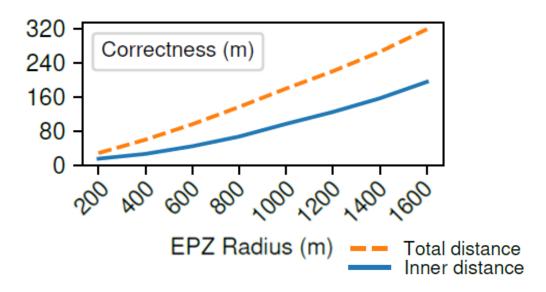
> Data leak prevention

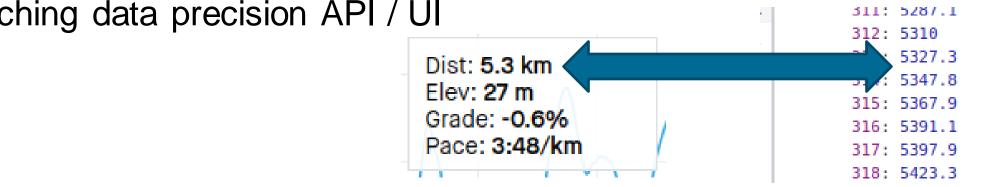
» Avoid inner distance scenario





- Data leak prevention
 - » Avoid inner distance scenario
 - » Fixing API leaks
 - » Matching data precision API / UI







> Reduce the possibility of inferences



- Reduce the possibility of inferences
 - » Metadata leaks may enable inferences!
 - » Model and mitigate possible inferences during design

>>> May require some out-of-the-box thinking





Consider inferences during algorithm design

- > Noisy distances?
 - » Random noise distributions average out!
- > Shifting distances?
 - » No influence on total distance scenario!

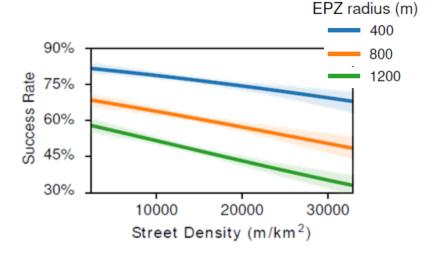


- > Regenerating EPZs yields more diverse data
- > Smoothing tracks makes regression more accurate

Apparent solutions might not work!

- > Nudge and support users towards privacy-friendly options
 - » Enable privacy zones by default
 - » Suggest EPZ radius based on street density
 - » Requires effective solutions

that do not violate user privacy perception





Proof-of-concept Service

- > 'Sanitize' sports activities
 - » Create privacy zone based on street density
 - » Avoiding the "inner distance" scenario
 - » Applying generalization
 - » Upload sanitized activity to service

Privacy Zones - Activities ① Info **Privacy Zones** Click the ⊕ button on the map to add a PZ Marker can be dragged to alter the + Enter address location. UWPUT Name LOLAN KU Leuven Campus Ð Leuven BELLE-VUI Radius 1200m Based on the location we recommend a 1200m N3 Bertem Heverlee CADOL Regenerate Privacy Zon Save ← Back Oud-Hey Leaflet | © Stadia Maps, © OpenMapTiles © O

https://priva.distrinet-research.be/

> All affected networks were contacted

> 3 out of 6 acknowledged our report

> Strava has engaged in a substantial discussion



- > We develop a novel **inference attack** on privacy zones
- > Intuition: distance metadata + street grid = protected location

Black Hat Sound Bytes

1. Thoroughly test API implementations for leaks

2. Consider inferences during algorithm design

3. Provide users with clear privacy options







DStrinet Thank you!

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