

Precise Positioning for the Mass Market

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Mass-Market GNSS Positioning Today



Where It's Going









Virtual Reality

Augmented Reality

Large-scale Mapping Autonomous Driving

Disruption of the RTK-Capable RX Market











Handheld RTK result with some signals passing through Ken's body







Work Proceeding on Two Fronts

Reference Network

Q: If we shift focus from network cost to end-user performance, i.e., time to reliable fix despite significant blockage and multipath at mobile device, how does this alter network layout?



Existing permanent reference network: 70-100 km interstation spacing

Mobile Device

Q: How must we modify existing low-cost mobile devices – handsets, VR headsets, intelligent vehicle sensors – to enable these to achieve cm-accurate positioning?



Samsung Galaxy phone with internal <\$1 Broadcom GNSS chip

Mass market (MM) use of cm-accurate positioning

upends the conventional wisdom (CW) on network design



<u>CW</u>: Permanent reference stations cost \$50k for initial equipage+install and \$1k/yr to maintain, so network must be <u>sparse</u>.

<u>MM</u>: Cost of network is negligible when amortized across millions of users; performance is key, so network must be <u>dense</u>. Besides, adequate-quality reference stations can cost less than \$1k for equipage+install, \$100/yr maintenance.



University of Texas & Samsung *Centimeter-Accurate Positioning System* (CAMPS) Dual-Frequency GPS/Galileo Reference Station Parts and assembly: ~\$1k <u>CW</u>: Minimize number of stations while ensuring every user is within R km of a station; results in *uniformly-distributed network*.

<u>MM</u>: Deploy stations <u>however necessary</u> to ensure P% of population experiences RMS corrections errors less than 2 mm; must take into account <u>population density</u> and <u>atmospheric spatial statistics.</u>



$$\nu = c_x dx + c_y dy + c_z dz + c_0$$

Standard deviation of network corrections for a linear DD iono + tropo model and iid multipath + thermal noise errors at each of 8 reference stations whose locations are known Red = high, blue = low



Dense network benefits: (1) iono+tropo DD errors can be made negligible, and (2) master station multipath attenuated



Reality check for single-frequency RTK with network-provided corrections: For low-cost antennas in realistic multipath environments, a dense network can't ensure reliable single-epoch fixing.



Adding a second frequency helps tremendously: for RMS corrections errors of 2 mm (attainable with < 20 km interstation spacing), reliable single-epoch fixing is possible with only m = 9 satellites. Q: What is functional relationship between network density and the size of RMS errors in the DD iono + tropo corrections?

(In other words, how dense must our reference network be to achieve 2 mm RMS corrections errors?)





Uncertainty in DD corrections at any point depends on (1) interpolation model, (2) accuracy of interpolation model, and (3) location of point relative to master The theory of stochastic geometry shows the way to a closed-form relationship between network density and average uncertainty of corrections



Average correction uncertainty when stations are distributed according to a Poisson point process



Average correction uncertainty when stations are distributed according to a repelling point process (e.g., determinantal)



Empirical data drawn from 23 stations in existing high-density public reference network in California



Corrections uncertainty is a highly nonlinear function of density. Floor due to multipath at mobile device reached only with < 20km interstation distance.



Empirical data drawn from 23 stations in existing high-density public reference network in California.



For the 30-km radius dense network studied, the difference between a linear and a quadratic interpolation model is near the multipath error level. Hence, the linear model is valid.





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Samsung Galaxy phone with internal <\$1 Broadcom GNSS chip Q: We already know that (fragile) fixedinteger RTK is possible using a smartphone's internal antenna. But what about using its internal GNSS chip (and clock) for this purpose?





produces pseudorange and carrier phase observables for all signals tracked (multiconstellation), but Android provides no access to these.



This Galaxy S5 produces RINEX files, thanks to Frank van Diggelen and Sergei Podshivalov (Broadcom).







The 5th anomaly is a nonzero and drifting bias in the carrier phase measurements that prevents both float and fixed solutions





Phone produced cycle-slip-free carrier measurements for 9 GPS L1 C/A signals over a half hour interval.

Primary Challenge: Large Time-Correlated Multipath Errors



Q: Is the phone antenna's irregular gain pattern to blame for the outlier DD phase measurements?



Experimental setup for Galaxy S3 gain pattern analysis







But our smartphone also has a camera ...



... what could we do with the camera?

Over the past 15 years, the computer vision community has made stunning advances in what is variously known as structure from motion (SFM), photogrammetry, and Visionbased Simultaneous Localization and mapping (VSLAM)







Extending the SFM technique, we optimally fuse camera images and GNSS phase measurements to shorten time to fix and create geo-referenced 3D maps









Sparse 3D Reconstruction



Dense 3D Reconstruction







Point #392758 P#392758: (-24.280500; -3.722230; 7.438940) Normal: (0.460726; 0.513380; 0.723998) Output Color: (94; 117; 136) Output	
	Position East North Up
Result: A globally-referenced 3D point	CDGNSS -24.278 -3.721 -7.440
cloud accurate to better than 1 cm	Difference -0.002 -0.001 0.002 Expected Std 0.004 0.008 0.005
	1.000 - 0.00

