

# Optimizing Resilient Time Delivery Through Estimation Calculations



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# Introduction

## The Importance of Precision Time

- Increasingly, precision time has become the critical parameter in synchronizing networks with migration first to LTE-TDD and now to 5G
- The principal source of UTC time is GNSS, which, while exceptionally robust, is vulnerable to interference, jamming, spoofing and solar flares, and in rare cases system errors
- Technologies such as the enhanced Primary Reference Time Clock (ePRTC), which couples a UTC source with an autonomous atomic clock, have been developed to address this
- Whether the backup clock is cesium as it generally would be for the ePRTC, or quartz or rubidium, as it generally would be for the PRTC, understanding the time error when UTC traceability is lost, is important

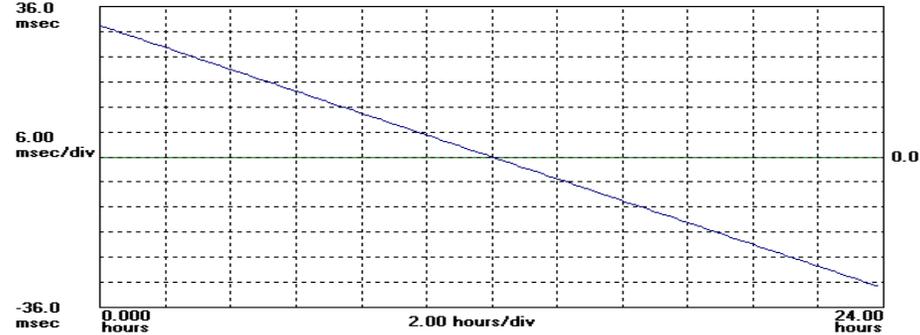
# Introduction

- **Time error estimator calculations**
  - **Random walk** (from frequency jump statistics)
  - **Holdover estimator** (from how well an oscillator conforms to predicted offset and drift)
- **How these time error estimator calculations can be used**
  - 1) **Optimize:** Optimize precision time delivery when UTC time not available
  - 2) **Select:** Measure and characterize stand-alone oscillators/clocks in order to select the best frequency sources in a population for deployment in timing systems
  - 3) **Predict:** Monitor an active system to indicate how long that system can hold time within required bounds when operating autonomously



# Oscillators: Offset and Drift Dominate

Microchip TimeMonitor Analyzer  
Phase deviation in units of time: Fs=296.3 mHz; Fo=10.000000 MHz; 03/12/97:02:37:24  
No filter selected; Test #1423; set 97.75; #23

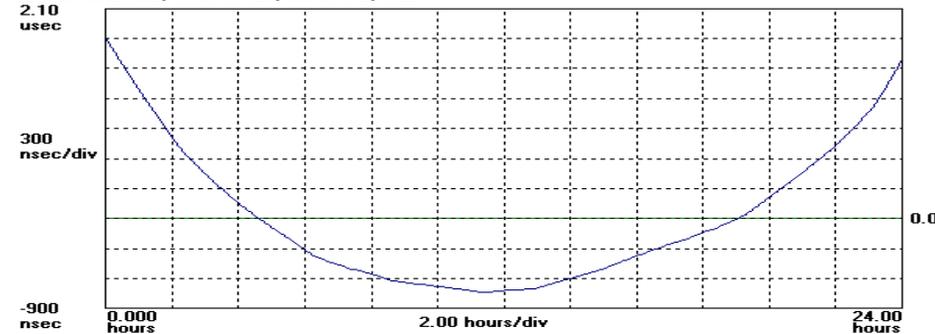


70 ms  
p-p

Relationship between phase and frequency:  $\omega = \frac{d\phi}{dt}$

Original oscillator phase measurement (0.7ppm frequency offset, constant phase slope)

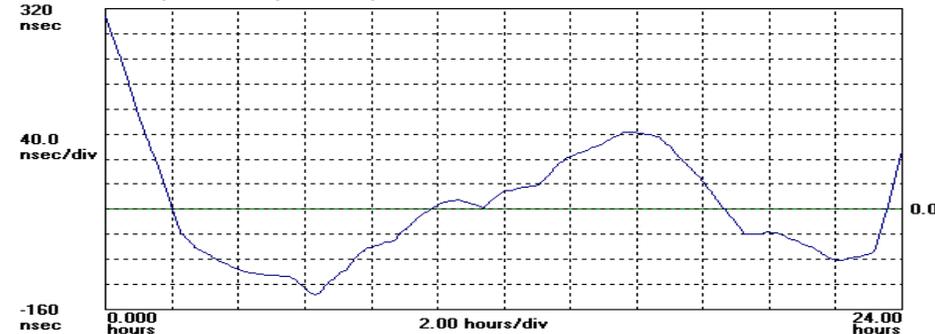
Microchip TimeMonitor Analyzer  
Phase deviation in units of time: Fs=296.3 mHz; Fo=9.9999927 MHz; 03/12/97:02:37:24  
No filter selected; Test #1423; set 97.75; #23



2.5  $\mu$ s  
p-p

Frequency offset removed (quadratic shape shows linear frequency drift of 0.2 ppb/day; given phase/frequency relationship, quadratic in phase means linear in frequency)

Microchip TimeMonitor Analyzer  
Phase deviation in units of time: Fs=296.3 mHz; Fo=9.9999927 MHz; 03/12/97:02:37:24  
No filter selected; Test #1423; set 97.75; #23



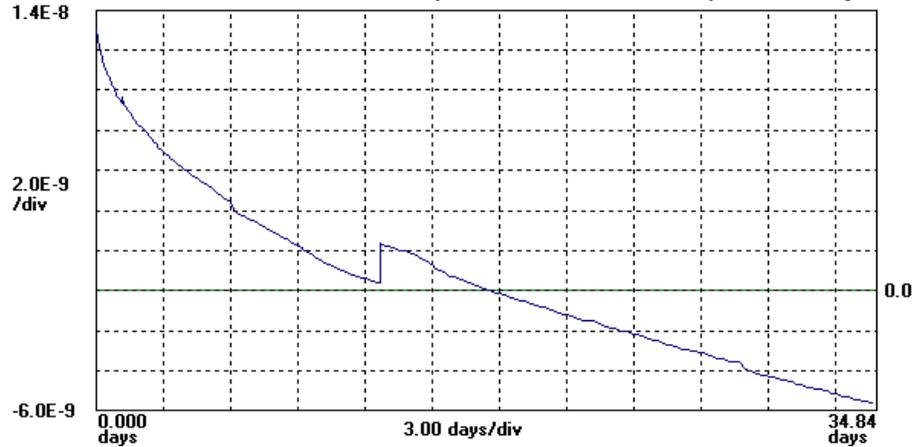
440 ns  
p-p

Frequency drift removed (shows residual phase movement)

# Oscillator Frequency Jump: Effect on Holdover

Microchip TimeMonitor Analyzer

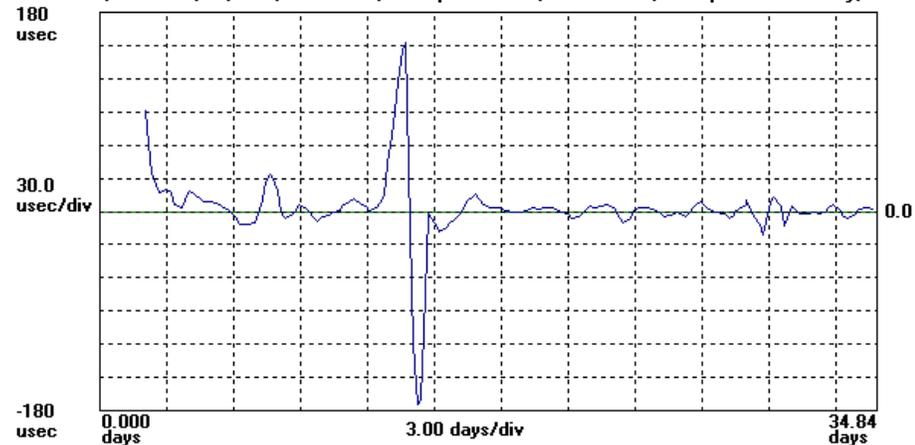
Fractional frequency offset; Fs=11.38 mHz; Fo=10.00 MHz; \*3/21/97 1:43:35 PM\*; \*4/25/97 9:50:08 AM\*;  
HP 53132A; Test: 87; Quartz; oscillator; Samples: 34259; Gate: 10 s; Freq/Time Data Only;



Oscillator measurement with frequency jump at 12 days

Microchip TimeMonitor Analyzer

Holdover vs. time; N=200; Start/Learn/Holdover(h): 0.000,24.00,24.00; Offset/Drift: \*3/21/97 1:43:35 PM\*;  
HP 53132A; Test: 87; Quartz; oscillator; Samples: 34259; Gate: 10 s; Freq/Time Data Only;



> 150  $\mu$ s rather than 1 to 10  $\mu$ s

The stochastic part is a very important component for determining performance!

The “random walk” and “holdover estimation” focus attention on this

# Random Walk

The difference between two adjacent frequency measurements is considered a “jump.” All jumps are stored in a histogram in accordance with their magnitudes. “Random Walk” is computed based on the statistics of jumps as the probable rms phase-walk. Data during warmup may be excluded.

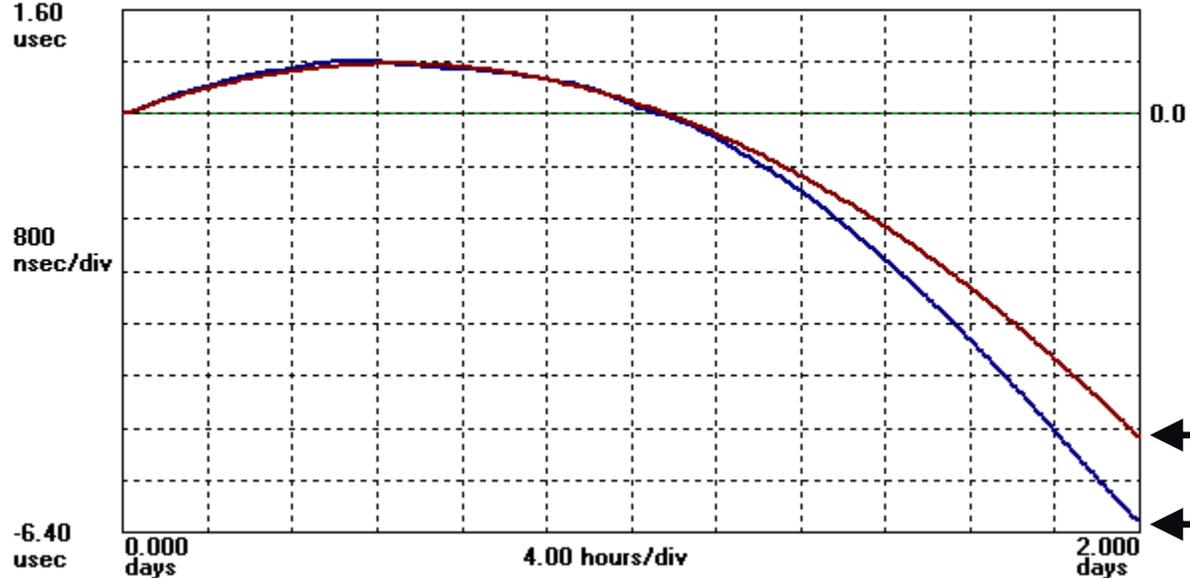
$$\sigma_{rwZ}^2 = \frac{Z^2}{3} \cdot \sigma_{jZ}^2 = \frac{Z^2 \cdot R}{6 \cdot \sum_{b=1}^M N_b} \sum_{b=1}^M N_b \left( \frac{h_b^2 + h_b l_b + l_b^2}{3} \right) \text{sec}$$

Random Walk calculation where Z is the random walk tau in seconds, T is tau in seconds, R is the ratio Z/T, M is the number of histogram bins,  $N_b$  is the number of frequency jumps in a particular bin b, l is the lower bin boundary, and h is the upper bin boundary.

# Holdover Estimator

## How to compute each individual “holdover estimator” point

Microchip TimeMonitor Analyzer  
Phase deviation in units of time; Fs=8.879 MHz; Fo=1.000000 Hz; 1999/01/18; 15:04:18  
1 (blue): Time Phase; Samples: 1535; Stop: 1535; HP 53132A; Test: 437; NGOP; D10565; Samples: 23;



Through curve fit, determine these three coefficients from the clock equation:

$$\phi_0 + \omega_0 t + \frac{1}{2} A t^2$$

Predicted time after 24 hours

Actual time after 24 hours

24-hour learning period, fit curve (in red) so it matches data (in blue)

24-hour holdover period

Holdover time error is difference between “predicted” and “actual”: 1.3  $\mu$ s

# Holdover Estimator

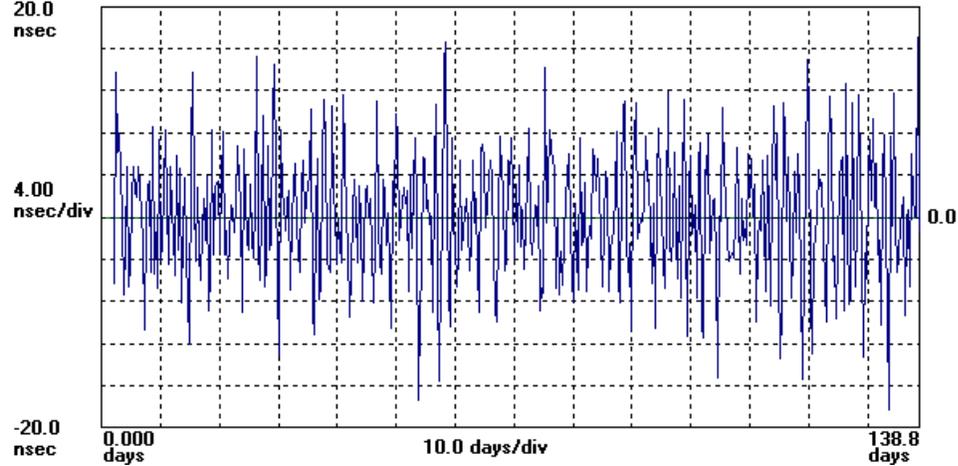
Quartz, Rubidium:  $\phi_0 + \omega_0 t + \frac{1}{2} A t^2$   
Fit to parabola

Cesium:  $\phi_0 + \omega_0 t$   
Fit to line

- “Holdover” computes frequency offset and drift during a specified learning period by fitting a 2<sup>nd</sup> order equation, and then uses that prediction of future drift for the specified holdover period. The predicted value is then compared to the actual value to compute a single holdover point; if the prediction is perfect then the holdover value is zero. Note for a primary reference such as a cesium clock, a 1<sup>st</sup> order equation fit is used as offset dominates.
- Then the starting point is moved forward in time and the whole procedure is repeated to compute another point. At the end, the statistics are calculated on the magnitude of the set of points, and a result determined, say the one-sigma or two-sigma point on the distribution.
- For a concrete example, consider an oscillator measured for three days with a 24-hour learning period and a 24-hour holdover period. If the starting point is advanced one hour starting with time zero, there will be 25 holdover points calculated to make use of the entire 3-day data set.

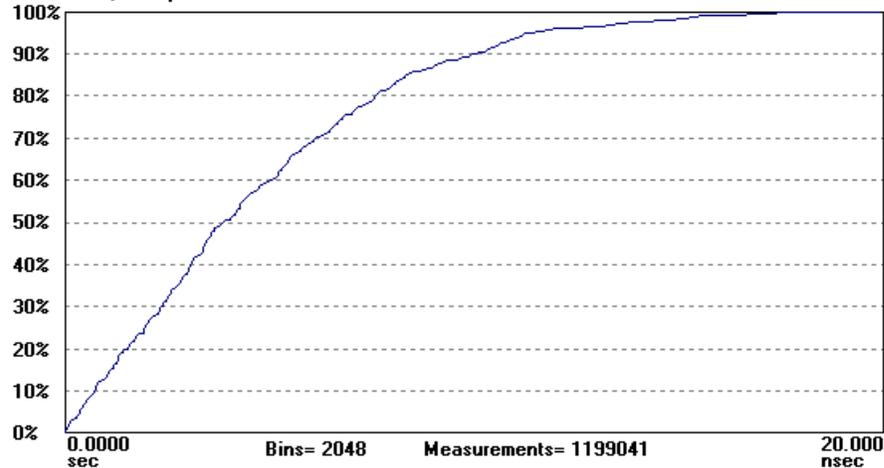
# Holdover Estimator Statistics

Microchip TimeMonitor Analyzer  
Holdover vs. time; N=500; Start/Learn/Holdover(h): 0.000,24.00,24.00; Offset/Drift; 2019/11/13 00:00:00  
Stable32 Phase; Samples: 1199041



Plot of holdover estimator results

Microchip TimeMonitor Analyzer  
Holdover CDF; Fs=100.0 MHz; Fo=10.00 MHz; 2019/11/13 00:00:00  
Stable32 Phase; Samples: 1199041



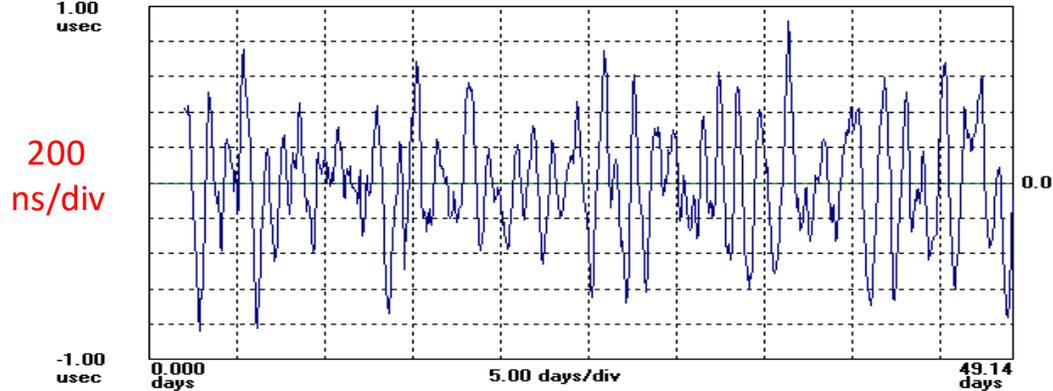
Cumulative distribution function of magnitude of the results

1 sigma: 4.41 ns

2 sigma: 10.35 ns

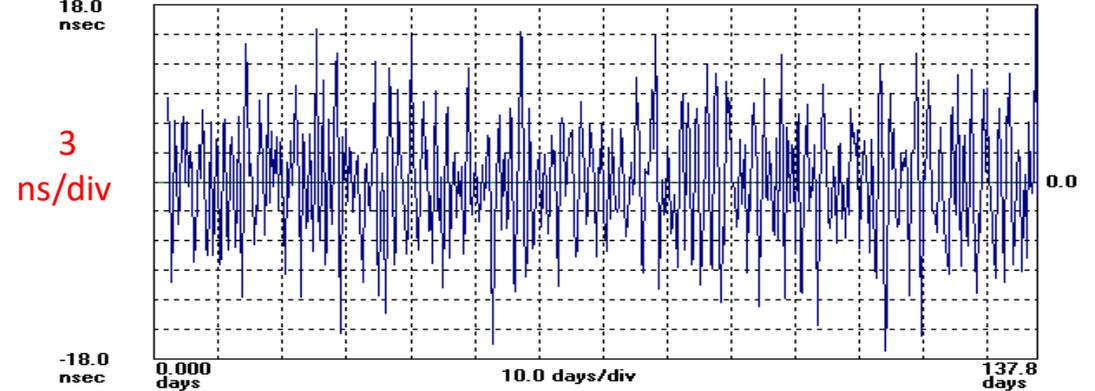
# Time Error Estimator Examples

Microchip TimeMonitor Analyzer  
 Holdover vs. time; N=1000; Start/Learn/Holdover(h): 0.000,24.00,24.00; Offset/Drift: 2018/11/15 11:22:50  
 Squid Phase; Chan 5; Samples: 529471; Start: 2018/12/03 11:22:54; Total Points: 723879;



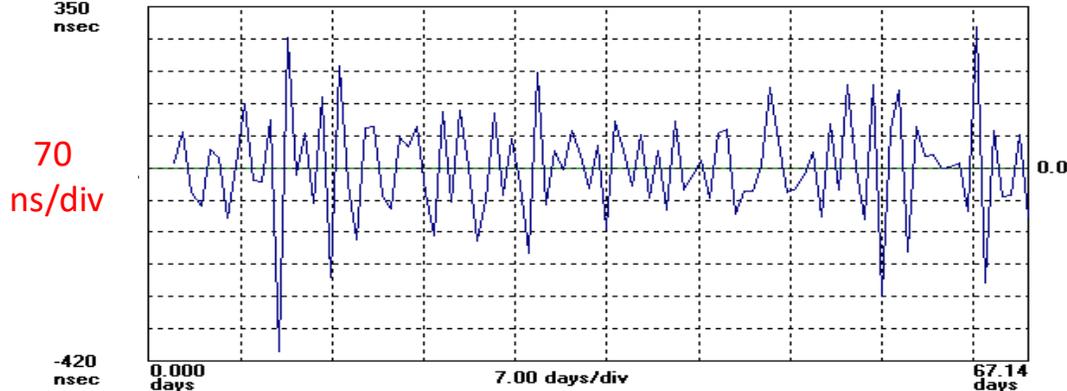
Quartz random walk: 638 ns  
 Quartz holdover estimator: 620 ns

Microchip TimeMonitor Analyzer  
 Holdover vs. time; N=500; Start/Learn/Holdover(h): 0.000,24.00,24.00; Offset/Drift: 2019/11/13 00:00:00  
 Stable32 Phase; Samples: 1190402; Start: 8640



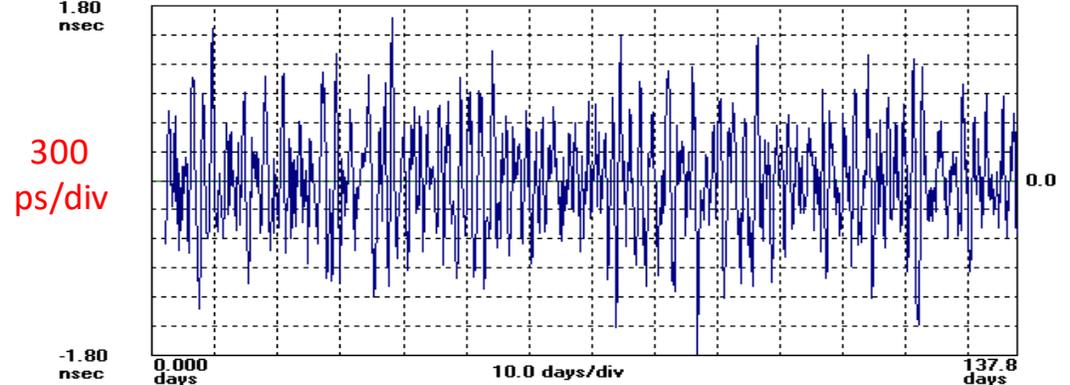
Cesium random walk: 14 ns  
 Cesium holdover estimator: 14 ns

Microchip TimeMonitor Analyzer  
 Holdover vs. time; N=100; Start/Learn/Holdover(h): 0.000,24.00,24.00; Offset/Drift: 2018/11/15 11:22:50  
 Squid Phase; Chan 6; Samples: 723844; Total Points: 723879; Rubidium



Rubidium random walk: 201 ns  
 Rubidium holdover estimator: 188 ns

Microchip TimeMonitor Analyzer  
 Holdover vs. time; N=1000; Start/Learn/Holdover(h): 0.000,24.00,24.00; Offset/Drift: 2019/11/13 00:00:00  
 Stable32 Phase; Samples: 1190408; Start: 8641



H-Maser random walk: 1.07 ns  
 H-Maser holdover estimator: 818 ps

# Time Error Estimator Examples

## Quartz

File count: 10		
Start: 8640		
ADEV(1000s); MDEV(7200s); TDEV(10000s)		
TUNC Start/Learn/Holdover(h): 0;24;24		
Random Walk FTau(seconds)/RWTau(days): 1000;1		
	TUNC(ns)	RW(ns)
Mean	1375	1291
Median	1414	1198
Stddev	242	223
Min	1003	1060
Max	1800	1677
1	1568	1216
2	1582	1114
3	1265	1069
4	1399	1677
5	1048	1179
6	1428	1070
7	1003	1563
8	1502	1447
9	1157	1515
10	1800	1060

TUNC (“time uncertainty”) is the same thing as the holdover estimator.

Note consistency between oscillators. More oscillators would be good to improve the statistics.

For these quartz oscillators, TUNC and random walk match reasonably well, so either or both could reasonably be employed as a predictor of holding time error.

## Cesium/H-Maser

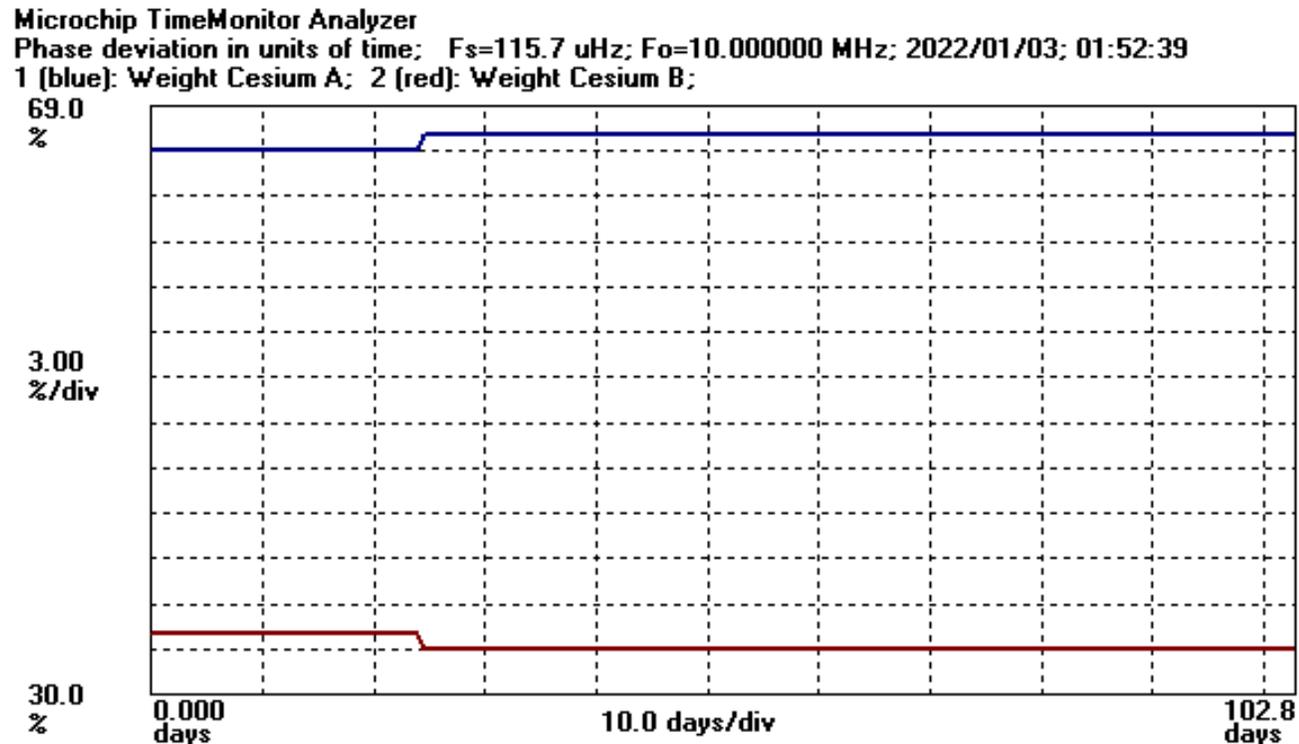
Random Walk FTau: 86400s			
Random Walk RWTau: 5 days			
TUNC in ns (0/120h/120h)			
	RW (ns)	TUNC(ns)	Days
Cesium1	23	40	182
Cesium2	13	24	138
Maser	2	4	138

Cesium learning period fit to first order equation as offset dominates.

Random walk tau and learning/holdover period extended for these atomic clocks.

# Time Error Estimation Calculators: Optimize

Use time error estimators calculated during normal system operation (with UTC available) to hold time optimally during stand-alone operation. For example, a time scale operating with connection to multiple atomic clocks could adjust relative weight based to some degree on these time error estimation calculators which applies for stand-alone operation.



# Time Error Estimation Calculators: Select

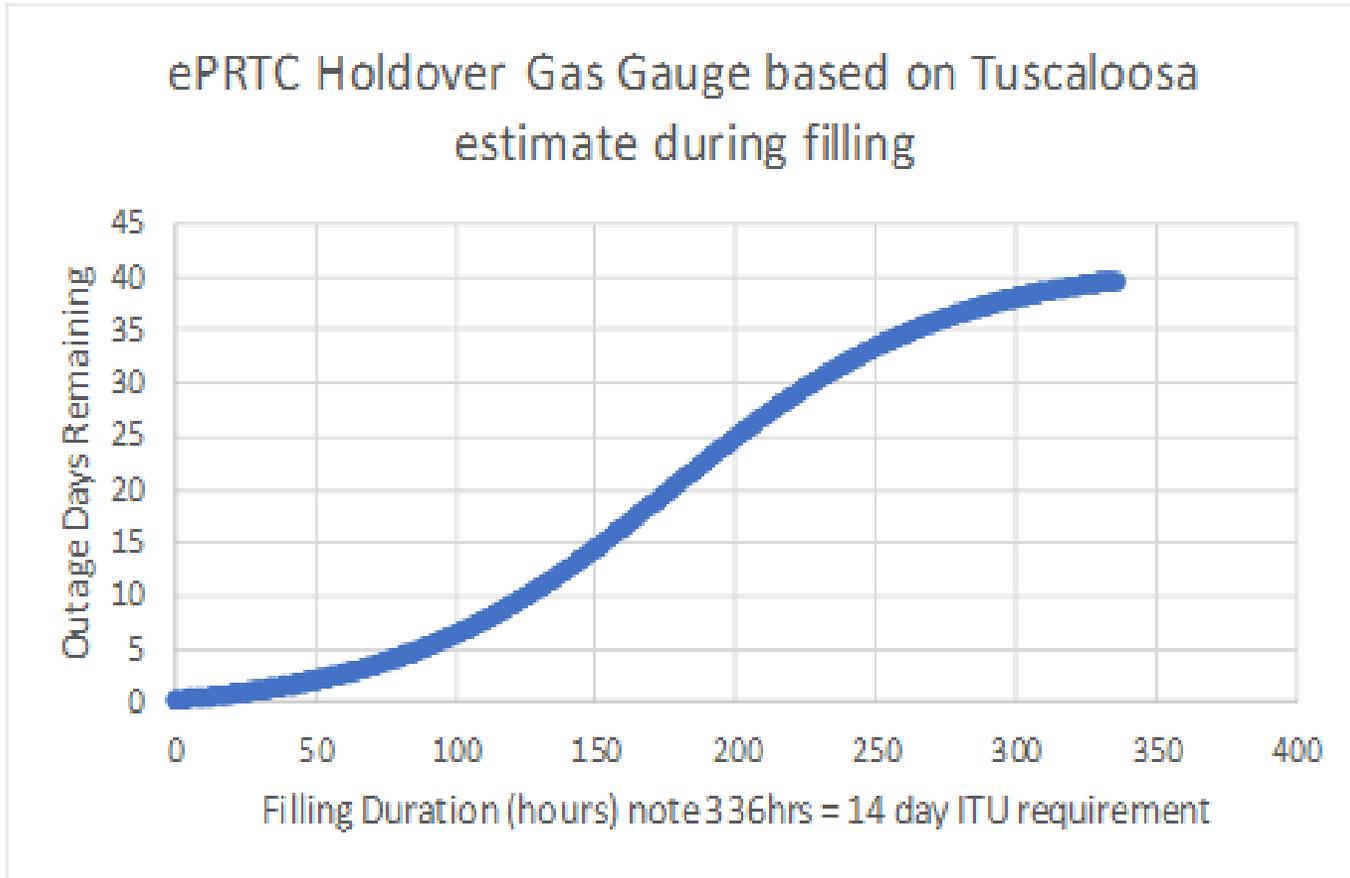
- **Select oscillators/clocks for a timing system**
  - **Qualify:** Is a particular oscillator model suitable from a performance standpoint for the particular system given its time accuracy/time holdover/time bridging requirements?
  - **Choose:** If several oscillator models could be suitable, choose which is preferable from a performance standpoint

File Count: 16		
Start/Duration Days: 3/16		
TUNC Start/Learn/Holdover(h): 0;24;24		
Random Walk FTau(seconds)/RWTau(days): 800;1		
	TUNC(ns)	RW(ns)
Mean	2688	1162
Median	1969	1094
Stddev	1401	221
Min	1141	873
Max	5932	1789

File count: 200		
Start: 8640		
TUNC Start/Learn/Holdover(h): 0;24;24		
Random Walk FTau(seconds)/RWTau(days): 800;1		
	TUNC(ns)	RW(ns)
Mean	4044	1609
Median	3793	1545
Stddev	1802	376
Min	228	963
Max	13533	2862

- **Bin:** For a selected oscillator/clock model, particularly for stringent system timing performance requirements, use time error estimators applied to measurement data to select which individual samples are suitable for a particular application

# Time Error Estimation Calculators: Predict



“Filling” refers to the “learning period”

From this graph, for example, it can be seen that after 14 days of learning,  $\pm 100$  ns can be held for an estimated 40 days, based on studying the autonomous cesium clock while UTC time via GNSS is available

# Summary

- **Holding time in a network is increasingly important given such technologies as TDD and 5G**
- **For determining predictive oscillator metrics, we start with the clock equation, which applies to all types of oscillators, from quartz to rubidium to cesium to hydrogen maser**
- **The stochastic term in the clock equation is critical: this is the part we can't predict or completely eliminate by design**
- **Two-time error prediction calculations, "random walk" and "holdover estimation," are two different approaches for understanding the stochastic term**
- **When a system has access to traceable UTC time, the oscillator can be studied, helping to predict how well time can be held if UTC traceability goes away**
- **These time error calculations can be used in three ways: (1) optimize system performance during time holdover, (2) select the best performing oscillator/clock samples from a population for use in timing systems, (3) predict and indicate during system operation how long required timing performance can be maintained if UTC connectivity is lost**

# Thank you

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