

Planning for Priority with Pings

Using High Resolution Transit Geolocation Data to Analyze Performance
and Identify Intersections that Contribute to Transit Delay

Tal Green, P.E.

*SFMTA (Muni), Transit Planner
San Francisco, CA*

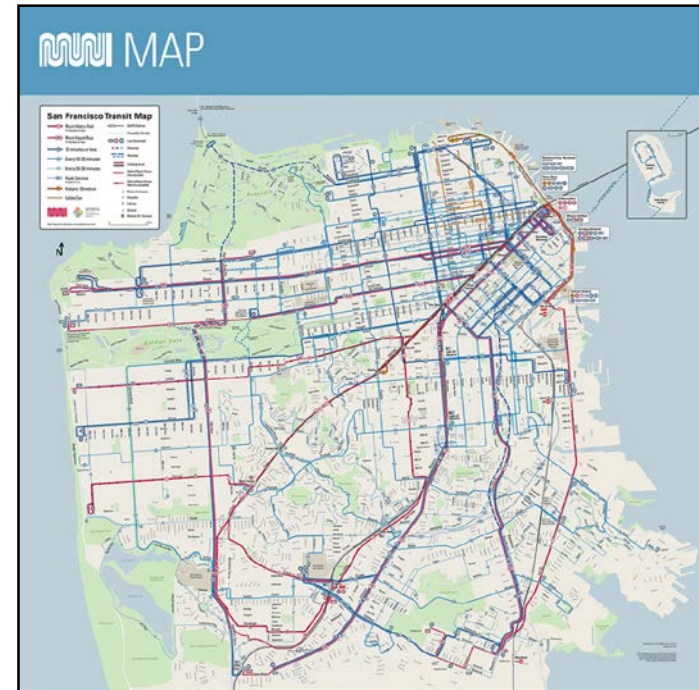


Sustainability & Multimodal Planning Workshop //



Presentation Outline

- Title breakdown!
- Current transit metrics
- Need for high res data analysis
- Data collection and methodology overview
- Metric #1 – transit signal delay
- Metric #2 - % of buses stopped at intersection
- Next steps



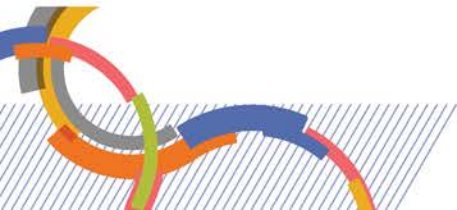
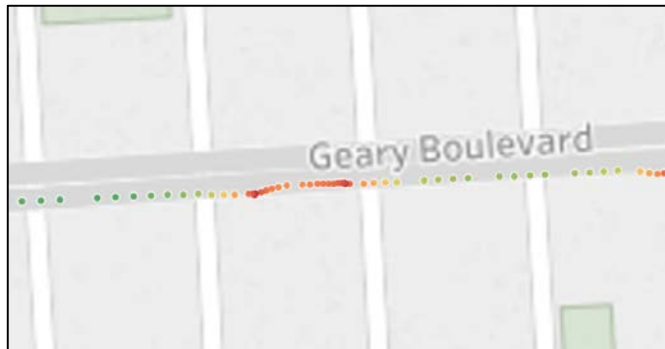
Title Breakdown

Using High Resolution Transit Geolocation Data
to Analyze Performance and Identify Intersections
that Contribute to Transit Delay



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Using **High Resolution** Transit Geolocation Data to Analyze Performance and Identify Intersections that Contribute to Transit Delay



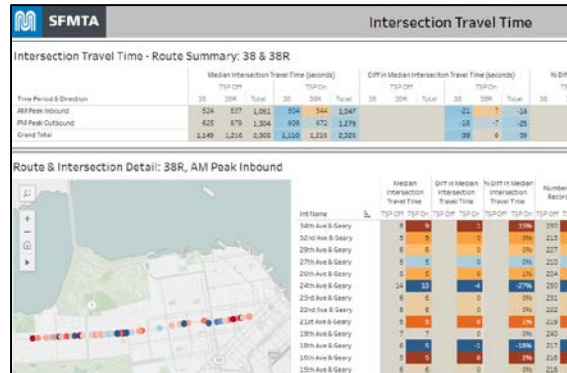
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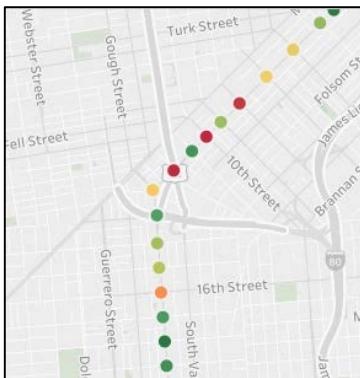
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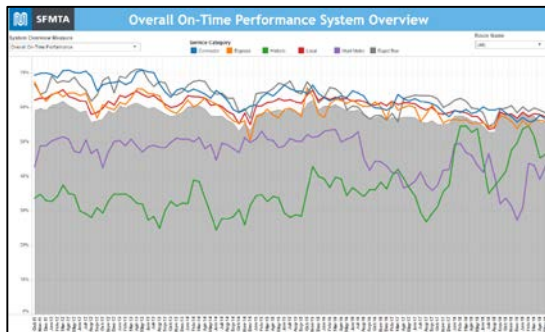


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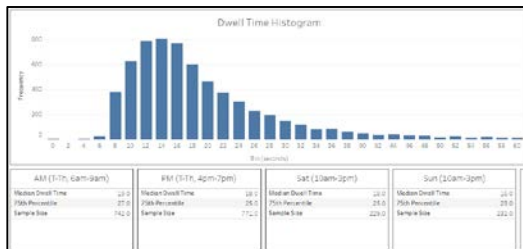
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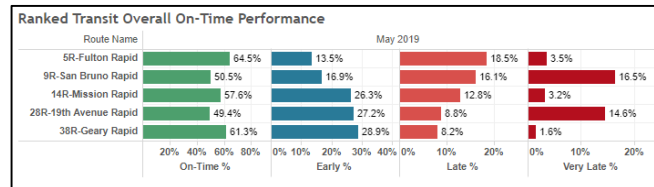
Currently Available Metrics



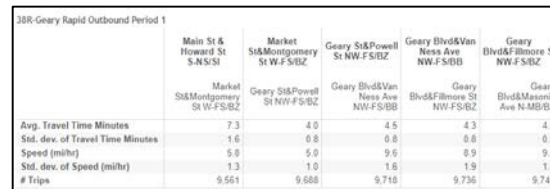
Mode Based On Time Performance



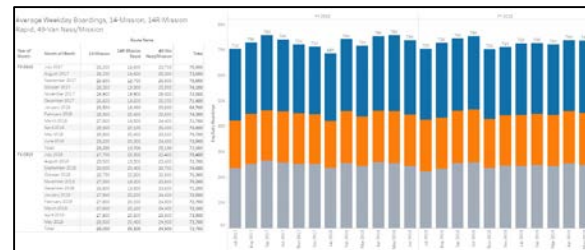
Stop Level Dwell Times



Route Based On Time Performance



Timepoint to Timepoint Travel Time



Route Level Ridership

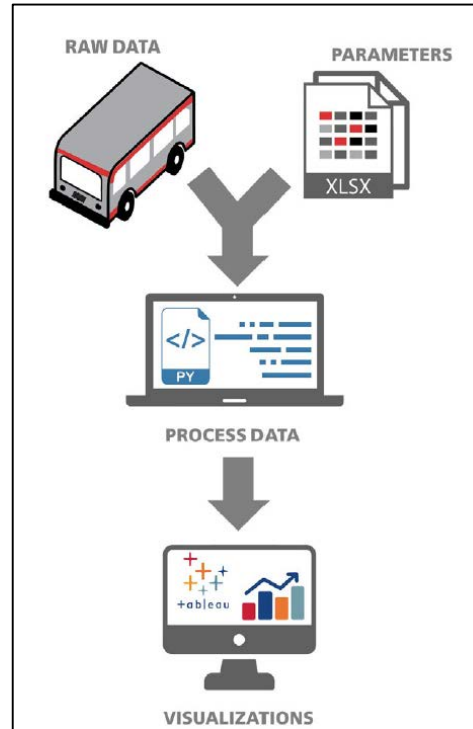
Data Collection Effort

- On board equipment collects data
 - Approximately 500 buses equipped
- Daily data dump in yard over WiFi
- Weekly transmission of all data (~4 GB)
 - Began collecting data in November, 2018
- Data processing occurs weekly
 - Currently on local machine
 - Future: through IT in a production data warehouse



2.7 million GPS points from one day of data

Methodology Overview



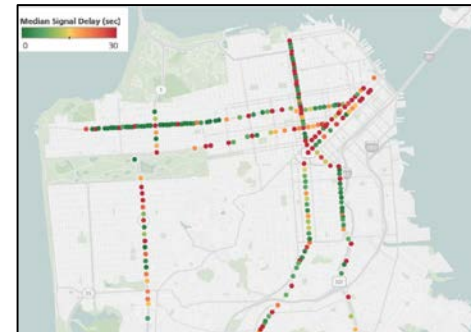
Metric #1 – Transit Signal Delay

- Processing
 - Isolated data to individual trips
 - Matched trip to schedule
 - Determined points of interest, matched to nearest
- Consumable data
 - Pushed to Tableau server
 - Can be pivoted by route, direction, location, time of day, etc.



Processing – target upstream and downstream points

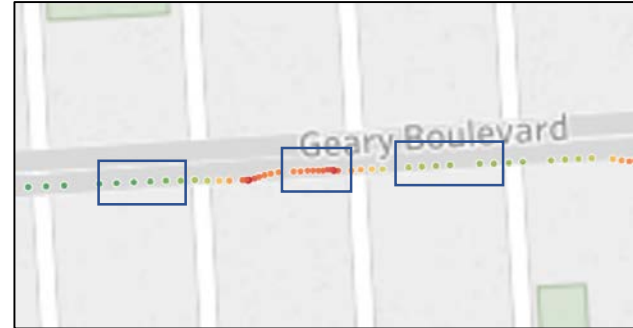
Intersection Name	Median Delay (sec)	Number of Records
25th St & Mission	4	372
24th St & Mission	14	380
23rd St & Mission	11	367
22nd St & Mission	19	365
21st St & Mission	2	350
20th St & Mission	3	347
19th St & Mission	16	343
18th St & Mission	20	346
17th St & Mission	2	343
16th St & Mission	3	339
15th St & Mission	7	329
14th St & Mission	17	332



Analysis – Tabulated and mapped delay information by TOD, direction, route

Metric #2 - % of Buses Stopped at Intersection

- Processing
 - Isolated data to individual trips
 - Matched trip to schedule
 - Draw approach zones for signals
 - Determine if buses traveled <2MPH
- Consumable data
 - Pushed to Tableau server
 - Can be pivoted by route, direction, location, time of day, etc.



Processing – define and utilize approach zones

Intersection Name	% Stopped	Number of Records
25th St & Mission	29%	400
24th St & Mission	47%	408
23rd St & Mission	41%	415
22nd St & Mission	69%	379
21st St & Mission	5%	374
20th St & Mission	14%	378
19th St & Mission	55%	396
18th St & Mission	70%	394
17th St & Mission	4%	377
16th St & Mission	33%	375
15th St & Mission	34%	380
14th St & Mission	59%	362



For More Details...

```
#During Daylight Saving (comment out if not) - March to November
allPoints['TimestampPacific'] = allPoints['TimestampNew'] - pd.Timedelta(
    hours=1)

allPoints['tripID'] = allPoints['tripID'].astype(str)
allPoints['date'] = allPoints['TimestampPacific'].dt.strftime('%Y-%m-%d')
allPoints['trip_Date'] = allPoints['date'] + " " + allPoints['tripID']
```

```
#display(allPoints.tail())
```

```
# Loop through each zone to get minimums in each per trip_Date
```

```
for z, zone in allZones.iterrows():
```

```
    try:
```

```
        pointsInZone = allPoints.loc[(allPoints['route'] == zone['route']) &
                                      (allPoints['DIRECTION'] == zone['DIRECTION']) &
                                      (allPoints['lat'] > zone['lat_min']) &
                                      (allPoints['lat'] < zone['lat_max']) &
                                      (allPoints['lng'] > zone['lng_min']) &
                                      (allPoints['lng'] < zone['lng_max'])]
```

```
        # Go to next zone if filter yielded nothing
```

```
        if pointsInZone.empty:
            continue
```

```
        # Determine the minimum value in each zone
```

```
        minPoints = pointsInZone.loc[pointsInZone.groupby("tripID").min()]
```

```
        # True if stopped, False if didn't stop
```

```
        minPoints['stopped'] = minPoints['speed'] < 2
```

```
        # Add values from current Zone series to each record
```

```
        minPoints['zoneID'] = zone['zoneID']
        minPoints['CNN'] = zone['CNN']
        minPoints['intName'] = zone['intName']
        minPoints['intSequence'] = zone['intSequence']
        minPoints['includesNearsideStop'] = zone['includesNearsideStop']
```

```
#LOOK INTO THIS: at some point, the order of the columns changed
resultsDF = pd.concat([resultsDF, minPoints])
```

```
mergedData['segmentID'] = segment['segmentID']
mergedData['CNN'] = segment['CNN']
mergedData['intName'] = segment['intName']
mergedData['freeFlowTT'] = segment['freeFlowTT']
mergedData['includesNearsideStop'] = segment['includesNearsideStop']
```

```
# Calculate delay
mergedData['delay'] = mergedData['travelTime'] - mergedData['freeFlowTT']
```

```
# Flag extreme outliers (True if > 10 minutes to get through intersection)
mergedData['extremeOutlier'] = (mergedData['travelTime'] > 600)
```

```
# Drop extra data
```

```
mergedData.drop(['PUBLICROUTENAME', 'TRIPID', 'LASTUPDATE', 'TimestampNew'], axis=1, inplace=True)
```

```
#resultsDF = pd.concat([resultsDF, mergedData])
```

```
return mergedData
```

```
except:
```

```
    print("ERROR!")
```

```
    # Add code here to better track errors (create list and add to list)
```

```
if progressBar(chunkCounter, totalChunks, currentSegment, totalSegments, progress=0):
```

```
    newProgress = int(100 * float(currentSegment)/float(totalSegments))
```

```
    if newProgress >= (progress+5):
```

```
        clear_output()
```

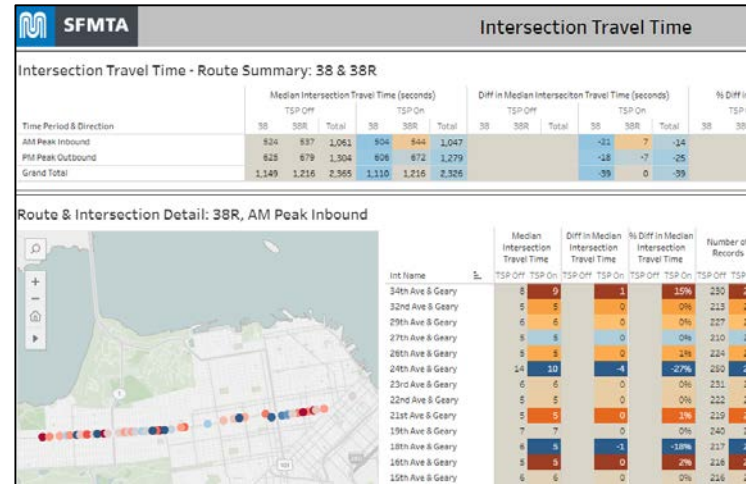
```
        print("Progress: ", newProgress, "% complete (" , currentSegment, " of ", totalSegments, " trips anal")
```

```
        progress = newProgress
```

```
    return progress
```

Next Steps

- Conduct before/after studies
- Productionalize processing efforts
- Include intersections with nearside stops
- Address poor GPS readings in NE due to Urban canyon effect



Acknowledgement



Ian Martin
Student Intern
Summers 2018, 2019

Thank You!



Tal Green, P.E.
Tal.Green@sfmta.com

