Sensor Fusion for Robust Path Duplication in Automated Ground Vehicle Convoys

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Outline



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Motivation



Potential Benefits of Automated Convoys

- Transport efficiency
- Increased safety
- Reduced stress on end user

Design Considerations

- Path following accuracy
 - Reduced footprint
 - Improved fuel savings
- Availability



- Potential for compromised GPS/coms reception
- Mission environment may not be conducive to additional infrastructure
- Reliability
 - Observation redundancy to handle measurement faults



System Description



- Four fully actuated tractor-trailers
 - Novatel PwrPak7 GNSS receiver
 - Interface to onboard CAN bus providing wheel-based speed/yaw rate, etc
 - DSRC radio (exchanging GNSS and vehicle state observations)
 - Delphi Electronic Scanning Radar (ESR)





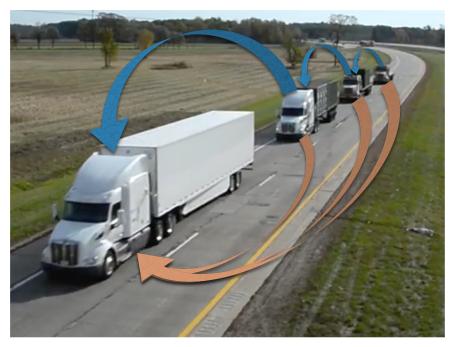


Path Following Approach



- Lane markings are not always available lateral control is realized by duplicating the path of a leading vehicle
- Two main frameworks for convoy networks:
 - Immediate Lead Following stability issues from accumulated errors
 - Ultimate Lead Following large following distances of trailing vehicles



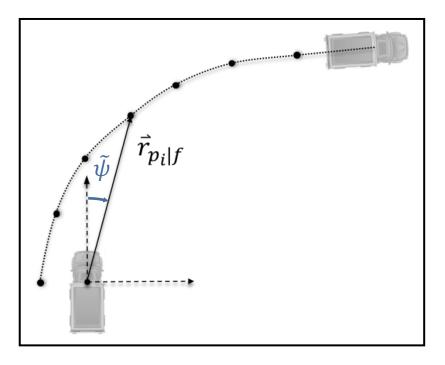


Path Following Approach



- Virtual Lead Following
 - The follower references a "virtual lead" at some userdefined lookahead distance
 - Choice of lookahead distance changes the effective damping
 - Classical heading controller acts to drive the relative bearing (\$\tilde{\mu}\$) to zero

$$\tilde{\psi} = \operatorname{atan2}(y, x)$$



 <u>Prerequisite</u>: Set of relative position vectors (RPVs) representing past positions of lead, relative to the follower vehicle, and resolved in the follower vehicle body frame

Path Generation Approaches



Global Path Generation

- Also known as "breadcrumb method"
- Both vehicles can measure absolute position (e.g. North, East)
- Leader saves path and shares with follower
- Requires base-station infrastructure for sufficient accuracy

Relative Path Generation*

- Accurate inter-vehicle RPV measurements
- RPV measurement combined with vehicle odometry (change in position in time) allows for path generation

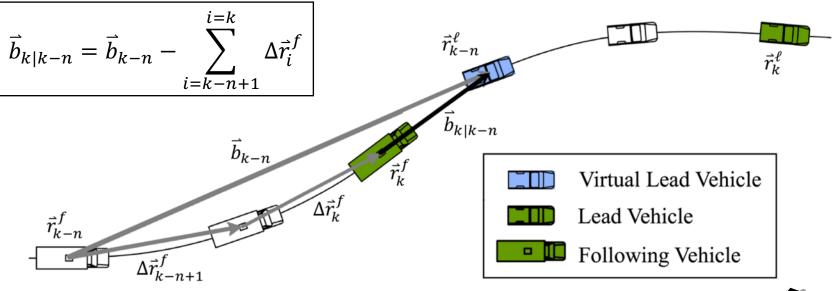
*Main focus of the presented work



Relative Path Generation



- **Odometry** of following vehicle is used to generate a dead reckoning position/orientation (pose) w.r.t. arbitrary origin
- Measurements of inter-vehicle RPV used to form the set of path waypoints
- Since the path is relative, the impact of dead reckoning errors only accumulate for *n* epochs (function of following distance and speed)







Limitations

- Depends on consistent availability of RPV and odometry
- Unbounded error growth with respect to separation distance
 Proposed Solution
- Incorporate additional measurements:
 - Lead vehicle odometry ($\Delta \vec{r}_i^{\ell}$)
 - Redundancy in odometry/RPV (e.g. ego-motion from wheel speed+yaw rate/inter-vehicle RADAR observations)
- Fuse measurements in a graph-based estimation framework



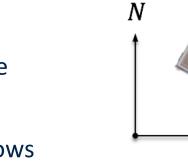
Odometry Methods

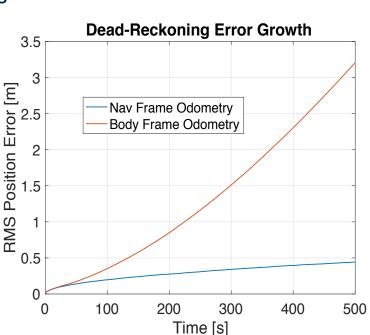


- Change in body frame position $(\Delta x, \Delta y)$ and heading $(\Delta \psi)$ over sampling period
- Calculated using wheel speed and yaw rate
- (+) good availability and high sample rate
- (-) dead reckoning position uncertainty grows relatively fast

Time Differenced Carrier Phase (TDCP)

- Change in navigation frame position $(\Delta N, \Delta E)$ over sampling period
- (+) millimeter level accuracy
- (+) position uncertainty grows linearly
- (-) limited availability





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RPV Methods

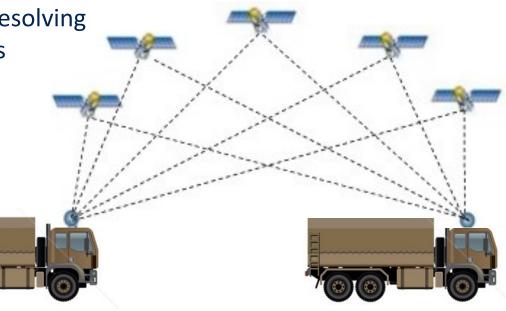


RADAR Observation

- Range/bearing of leader with respect to the follower
- (+) Does not require coms or GPS reception
- (+) High sample rate
- (-) Poor availability (FoV, signal shadowing from neighboring vehicles, etc.)

Dynamic-Base RTK (DRTK)

- Calculates inter-vehicle RPV by resolving carrier phase integer ambiguities
- (+) Centimeter level accuracy over long distances (~10km)
- (-) Poor availability in certain environments



Graph-Based Path Estimation



- Graph-based estimation framework is considered to smooth path and span GPS/coms outages
- Graph edges → measurements (weighted by uncertainty)
- Graph vertices \rightarrow state estimates (2D position)
 - Follower subgraph F with $\{\vec{r}_{k-W}^f, \dots, \vec{r}_k^f\} \in V(F)$
 - Lead subgraph L with $\{\vec{r}_{k-W}^{\ell}, \dots, \vec{r}_{k}^{\ell}\} \in V(L)$
 - > Follower Positions
 - > Leader Positions
 - Relative Position Vector
 - Leader Odometry Vector
 - Follower Odometry Vector

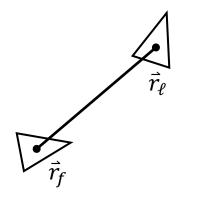
Graph-Based Path Estimation



- Each measurement's contribution is stored in the information matrix ($\Omega \in \mathbb{R}^{n \times n}$) and information vector ($\xi \in \mathbb{R}^{n \times 1}$)
- The normalized innovation is added to the rows/columns that correspond to the adjacent states as shown below

$$\Omega_{[ij]} += \mathbf{H}_{ij}^T \mathbf{R}_{ij}^{-1} \mathbf{H}_{ij}$$
$$\xi_{[ij]} += \mathbf{H}_{ij}^T \mathbf{R}_{ij}^{-1} z_{ij}$$

- R measurement covariance matrix
- H measurement matrix
- z measurement vector
- *i*, *j* adjacent state indices



Example: RPV measurement $H = \begin{bmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 0 & -1 \end{bmatrix}$ $R = E[\nu\nu^{T}]$ $z = (\vec{r}_{\ell} - \vec{r}_{f}) + \nu$



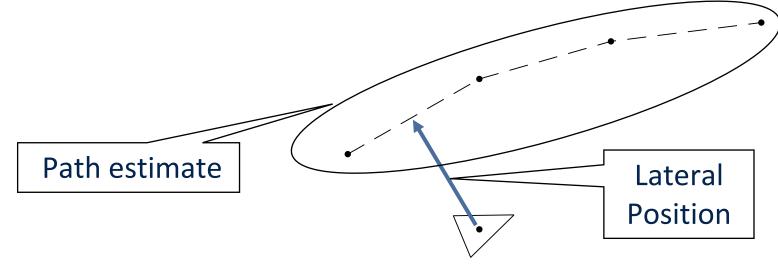
Graph-Based Path Estimation



• After accounting for each measurement in the graph, the estimated state (\hat{x}) is calculated by solving

$$\Omega \, \hat{x} = -\xi$$

- The state estimate contains the relative path of the leader, which is used to calculate the follower vehicle's control input
- This process is repeated at each time epoch for continued control of the follower



Results



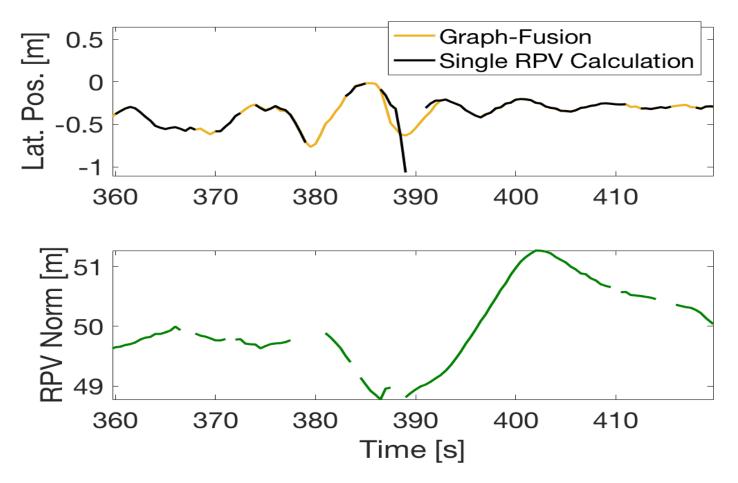
- Data from CACC demonstrations in two environments: 1) Oval test track 2) Forest road
- Lateral position is used as performance metric
- No ground truth in this data set results are based on improved availability
 - Path generation using only DRTK + follower TDCP
 - Path estimate from graph-based fusion
- Three main scenarios to consider:
 - 1) DRTK RPV outliers from false integer ambiguity
 - 2) TDCP odometry outliers from multi-path
 - 3) Complete GNSS outage (e.g. tunnel)



Results: RPV outlier case



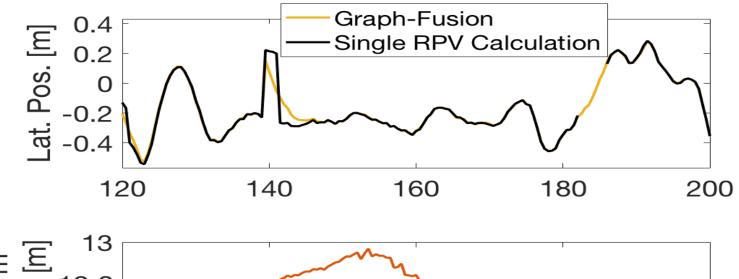
- Spikes/outages in RPV -> spikes/outages in path
- Graph fusion spans gaps and dampens spikes

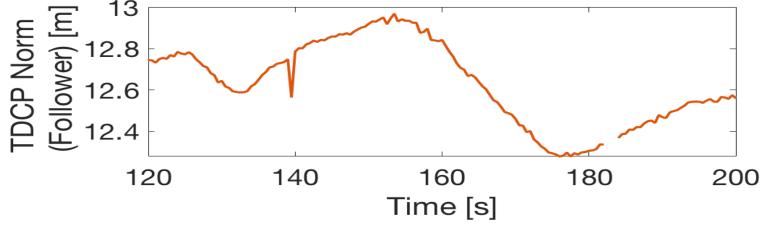


Results: Odometry outlier case



- Odometry outliers -> bias path over entire following distance
- Graph fusion spans outages, but still heavily influenced spikes

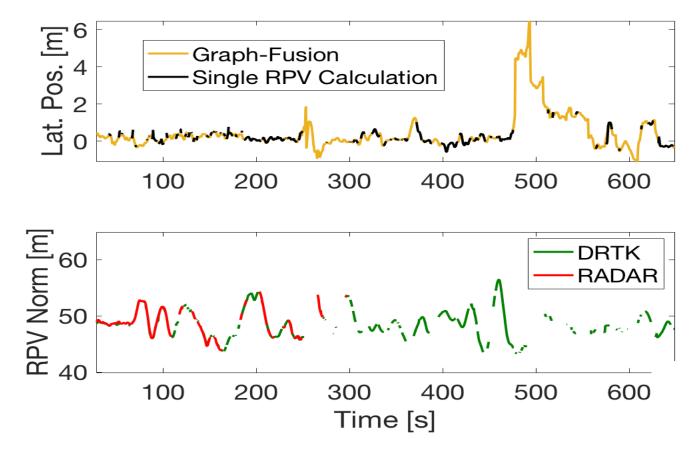




Results: GNSS outage case



- Graph-based solution relies solely on body-centric measurements during outages
- Availability is improved, but accuracy significantly diminished





Testing:

- Better performance analysis with ground truth
- Test with online controller
- Algorithm:
- Graph edge outlier rejection
- Landmark exchange for GNSS-denied SLAM
- Higher resolution path using high rate measurements (RADAR CAN information at 5Hz)





Questions?

Thank You!

