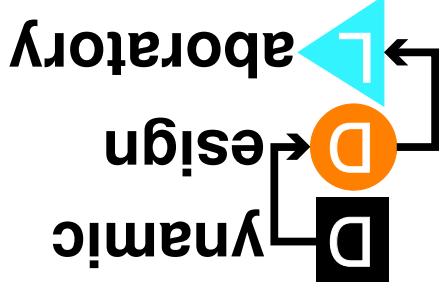


Identifying Tire Pressure Variation by Nonlinear Estimation of Longitudinal Stiffness and Effective Radius

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Goal

Robust identification of tire properties using stock vehicle sensors

Nearly all new cars have ABS

GPS already prevalent and soon standard

Tire longitudinal stiffness depends on many

factors

- Inflation pressure
- Tread depth
- Normal load

Applications

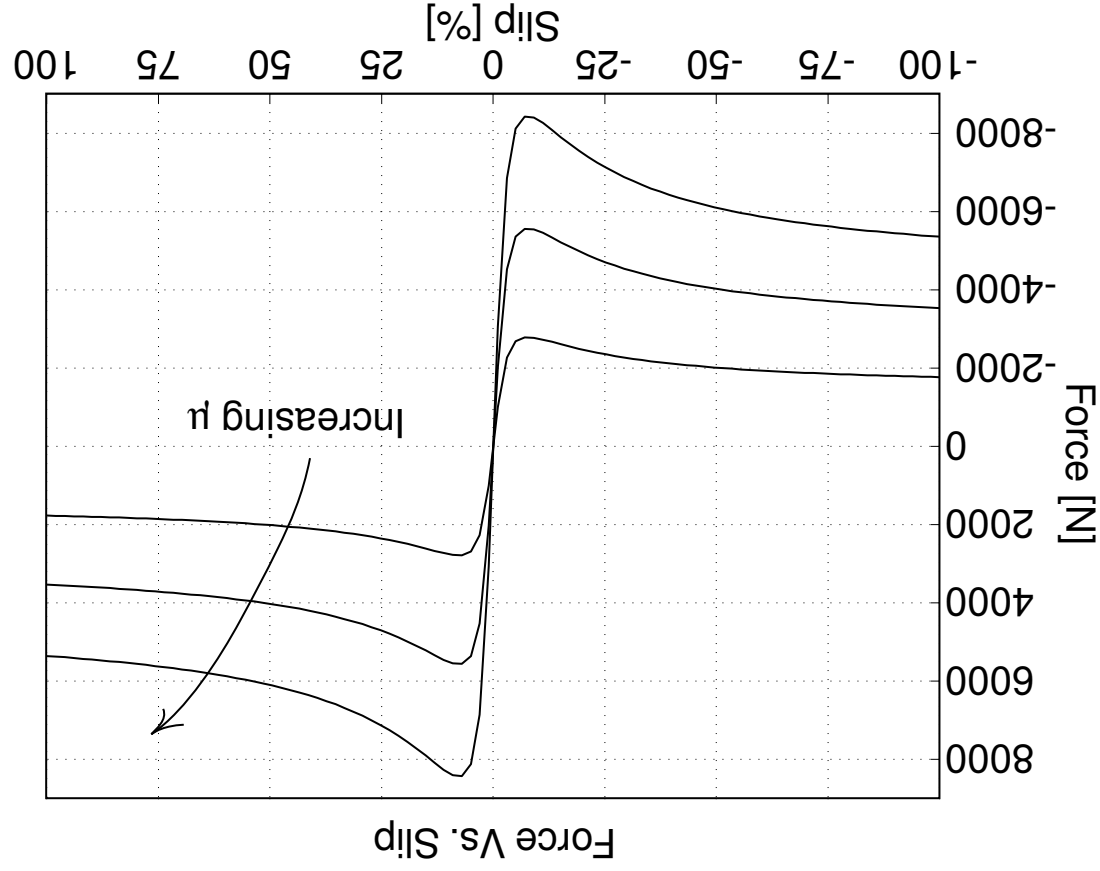
- Onboard diagnostics
 - Pressure sensor
 - Tread wear
- Peak traction force
 - Stability control
 - Driver warning systems
- Tire modeling for simulation parameters

Outline

- Force Slip Model
- Available Sensors
- New Estimator Structure
- Results
- Conclusions
- Much more consistent estimator
- Longitudinal stiffness inversely proportional to inflation pressure

Force-Slip Model

Force Vs. slip
approximately
linear for low
values of slip

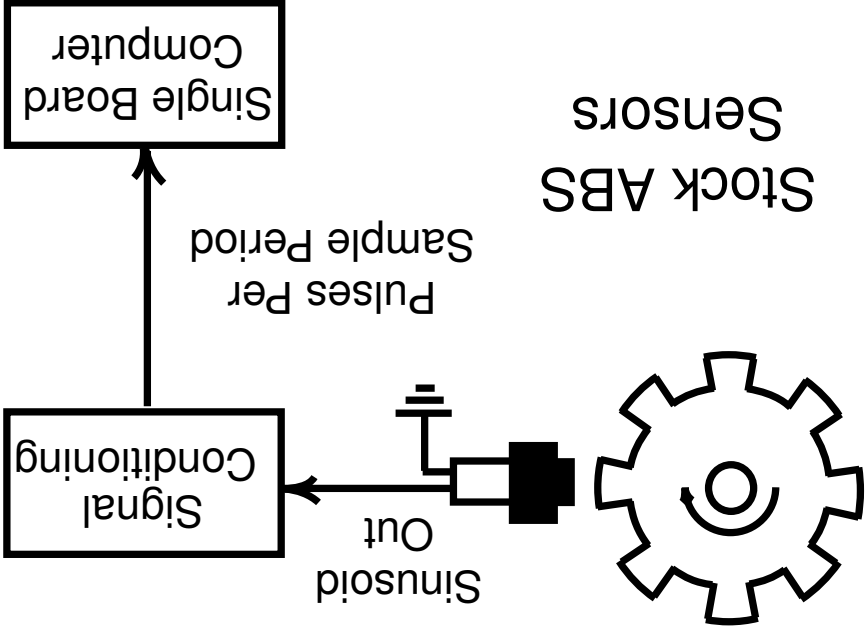


$$F = C_x \left(\frac{V - R\omega}{V} \right)$$

Available Sensors



Novatel GPS Receiver



Measurements

$$\hat{v}^k = R_r \left(\frac{2T}{\hat{\theta}_r^{k+1} - \hat{\theta}_r^{k-1}} \right)$$

Velocity approximation

$$\hat{a}^k = R_r \left(\frac{4T^2}{\hat{\theta}_r^{k+2} - 2\hat{\theta}_r^k + \hat{\theta}_r^{k-2}} \right)$$

Acceleration approximation

$$\hat{w}_d^k = \left(\frac{2T}{\hat{\theta}_f^{k+1} - \hat{\theta}_f^{k-1}} \right)$$

Angular Velocity approximation

Previous Work

$$F = C_x \left(\frac{V - R\omega}{V} \right)$$

\Leftrightarrow

$$\hat{a} = \begin{bmatrix} \frac{1}{m} \\ \hat{\omega}_f \\ \frac{mV}{Rf} \end{bmatrix} \begin{bmatrix} C_x \\ Rf \\ C_x \end{bmatrix}$$

Least squares parameter estimation

Had large variance of C_x estimates for similar test conditions

Energy Formulation

Try reformulating problem as energy balance

$$\frac{dV}{dt} = C_x \left(\frac{V - R\omega}{V} \right)$$

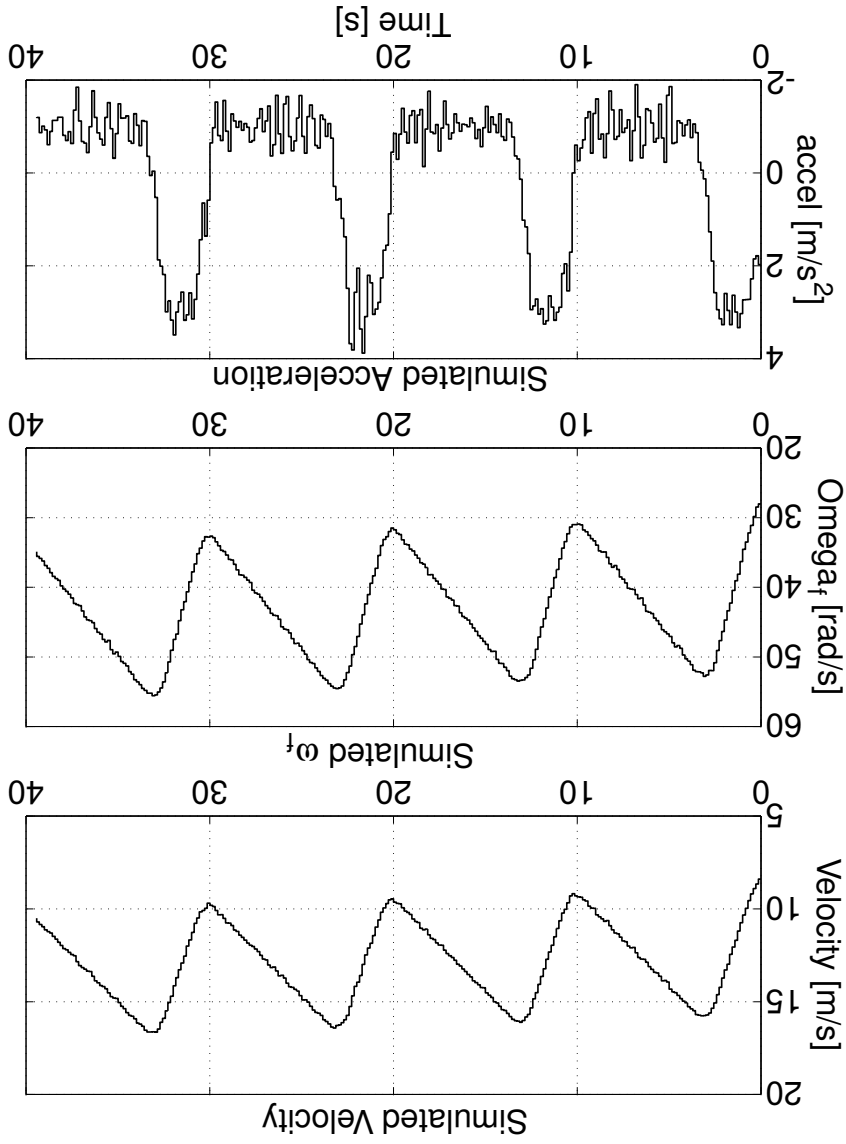
$$m \int V dV = -C_x \int (V - R\omega) dt$$

$$mV^2 - mV_0^2 = -2C_x(S - R\theta)$$

$$mR_r^2 \left(\dot{\theta}_r^2 - \dot{\theta}_{r0}^2 \right) = -2 \left[R_r \hat{\theta}_r - \hat{\theta}_f \right] \begin{bmatrix} R_f C_x \\ C_x \end{bmatrix}$$

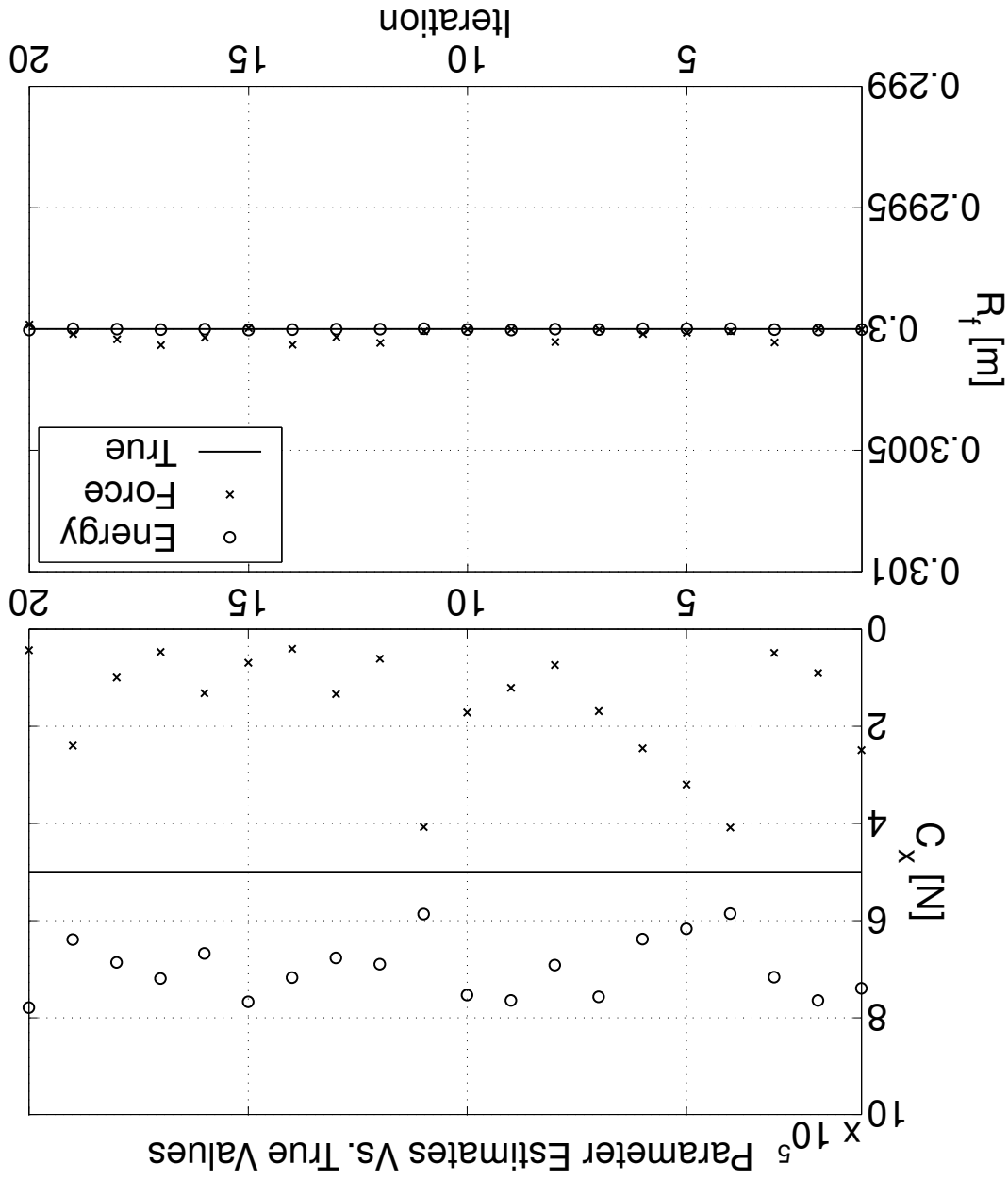
Simulation Setup

Wheel angle
measurement noise
modeled as white



Simulation Results

Linear least squares estimators return estimates with variance dependent biases



Implicit Regularization

$$\hat{a} = \begin{bmatrix} \frac{1}{m} \\ \frac{m\hat{V}}{\hat{\omega}_f} \end{bmatrix} \begin{bmatrix} C_x \\ R_f C_x \end{bmatrix}$$

$$a + \Delta a = \begin{bmatrix} \frac{1}{m} \\ \frac{\omega_f + \Delta\omega_f}{m(V + \Delta V)} \end{bmatrix} \begin{bmatrix} C_x \\ R_f C_x \end{bmatrix}$$

Errors in estimation matrix cause bias in parameter estimates

Recast as a Nonlinear Problem

Write out the measurement errors explicitly

$$m_{R_r}(\theta_r + \Delta\theta_r) - Cx = \left(\frac{R_r(\theta_r + \Delta\theta_r)}{R_f(\theta_f + \Delta\theta_f) - R_r(\theta_r + \Delta\theta_r)} \right)$$

\Leftrightarrow

$$f(\hat{\theta}_f, \hat{\theta}_r, \Delta\theta_f, \Delta\theta_r, R_r, Cx) = 0$$

Nonlinear Optimization

Minimize the measurement errors subject to the force-slip constraint function (orthogonal regression)

$$\begin{aligned} \text{Minimize:} & \quad \left\| \begin{array}{l} \Delta \theta_f \\ \Delta \theta_r \end{array} \right\| \\ \text{Subject to:} & \quad f(\hat{\theta}_f, \hat{\theta}_r, \Delta \theta_f, \Delta \theta_r, R_r, C_x) = 0 \end{aligned}$$

Cost Surfaces are Quasiconvex

Locally

quasiconvex

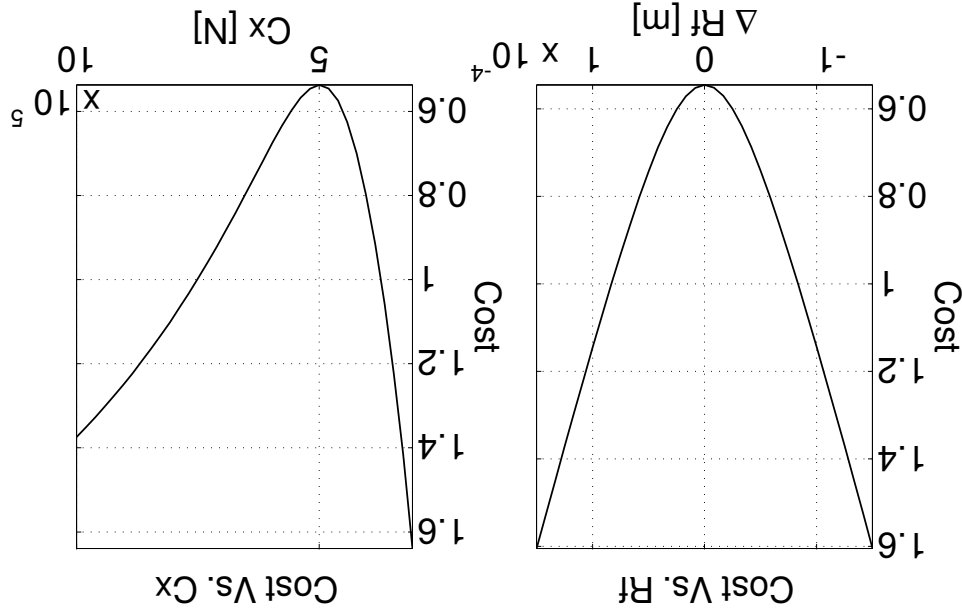
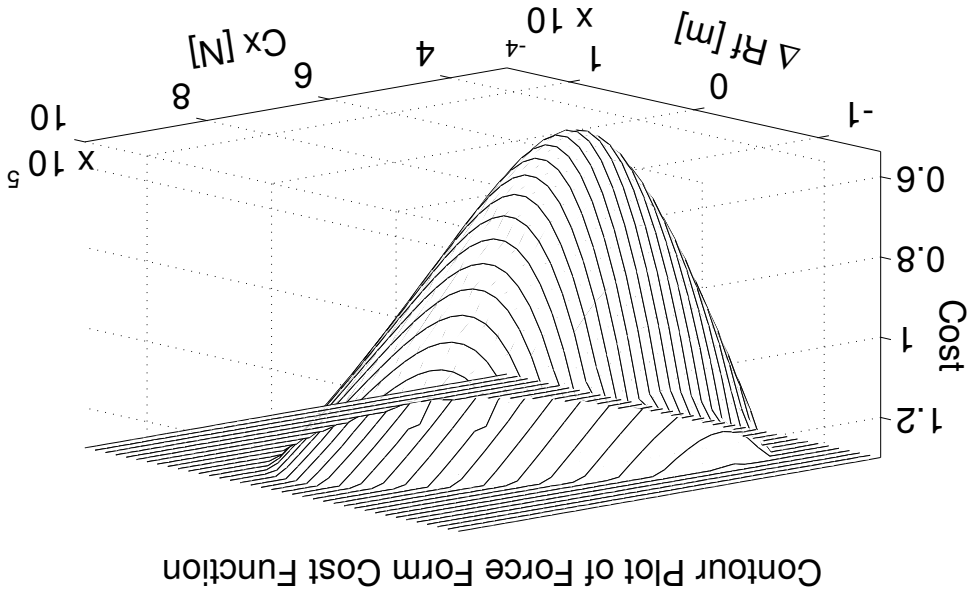
cost function

insures unique

solution for

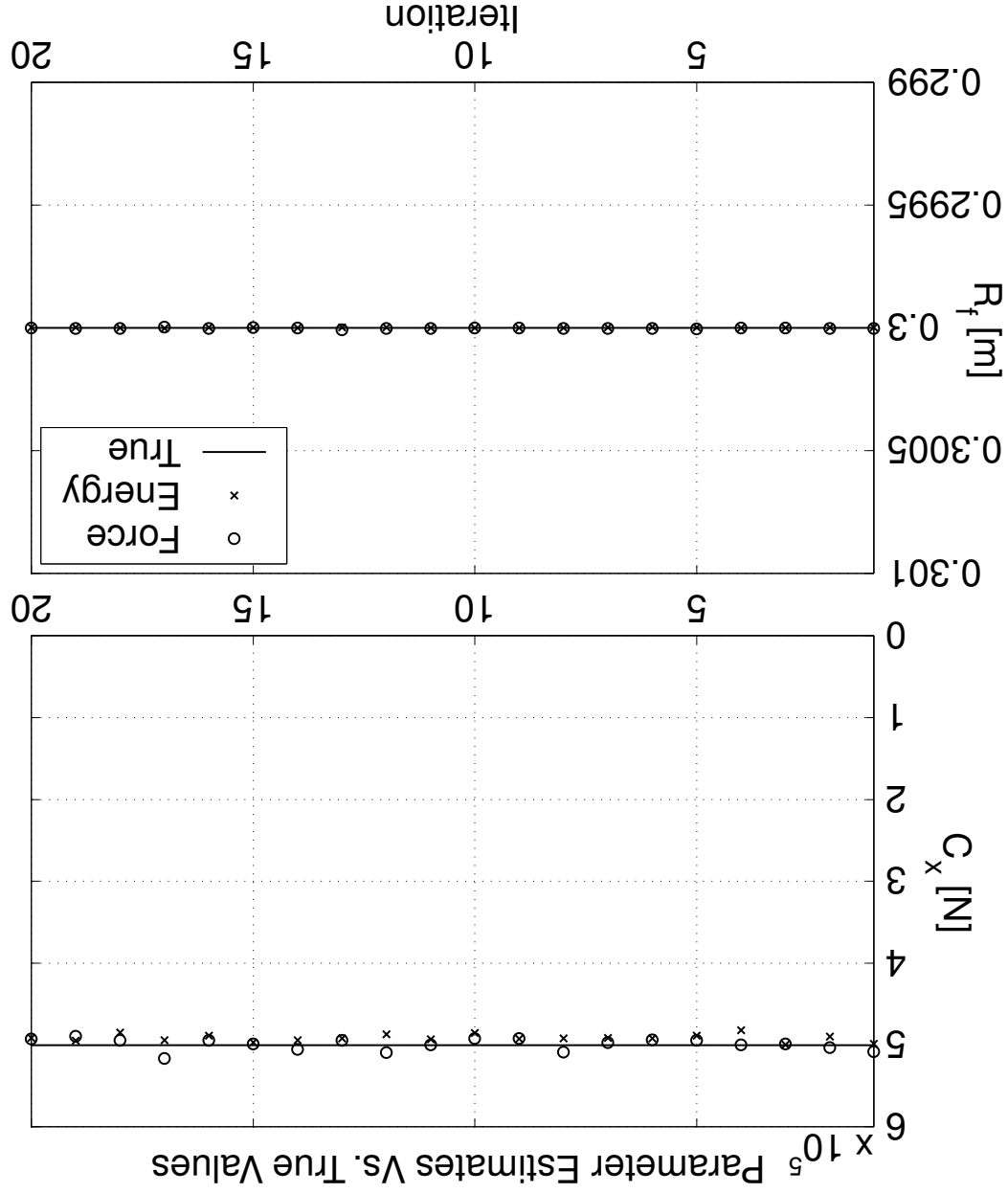
parameter

estimates

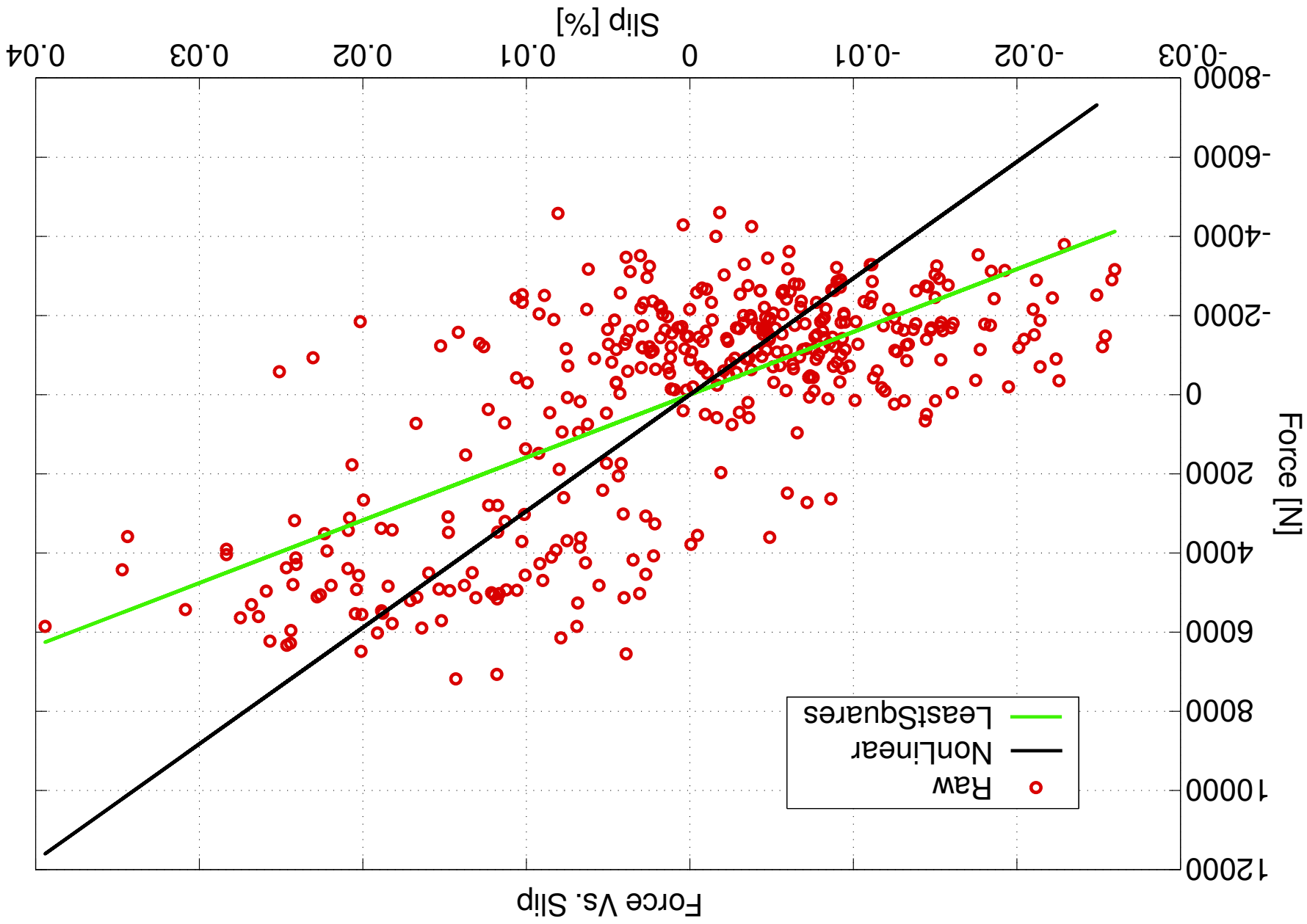


Simulation Results

Nonlinear schemes
return much more
consistent
parameter
estimates



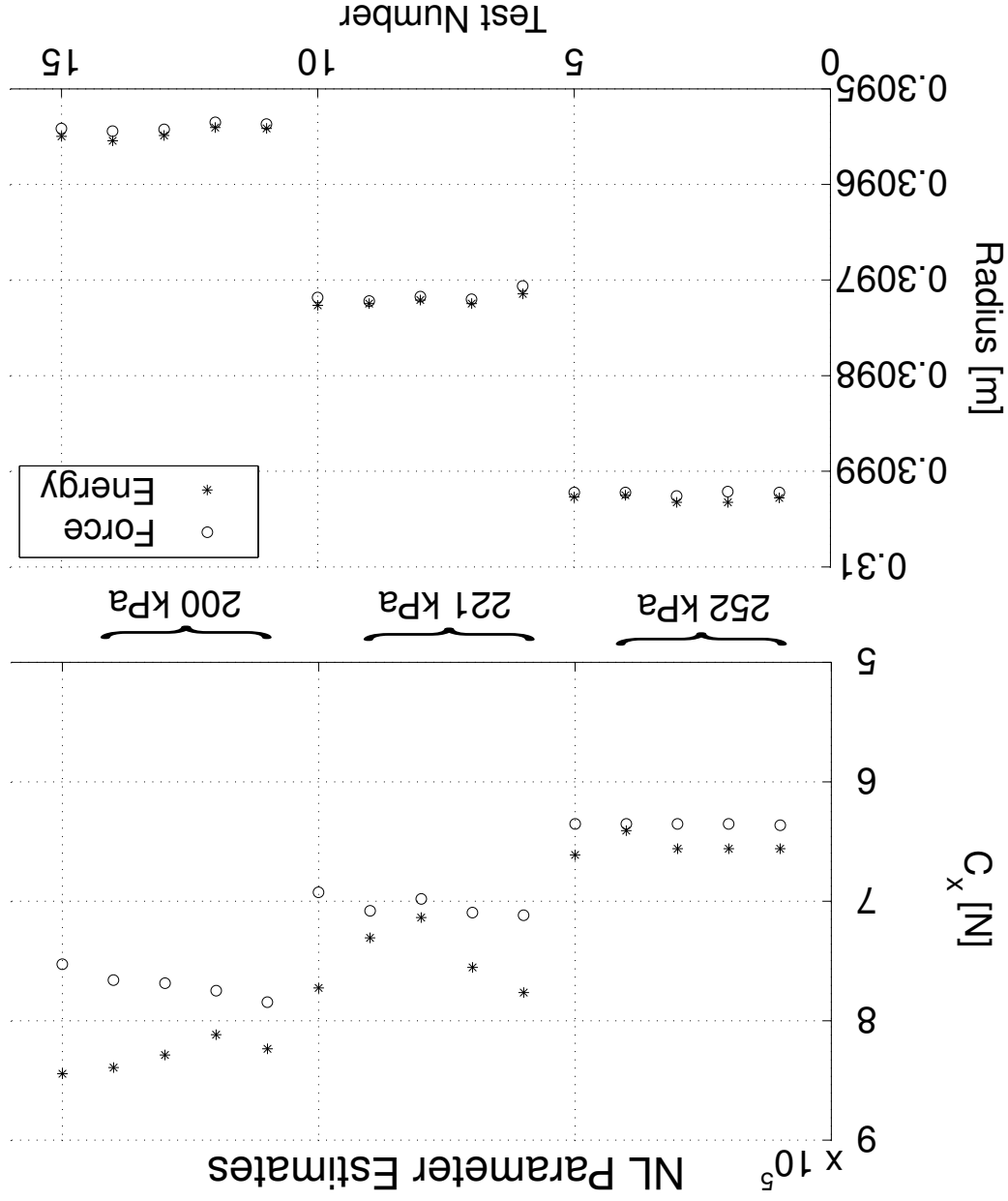
Simulation Results 2



Test Data

Test data for three different tire pressures.

Each data point represents 60-90 seconds of data



Test Data

- Longitudinal Stiffness estimates repeatable to about 5%
- Strong dependence of C_x upon pressure
- Very small change in effective radius, 0.4 [mm] for 50 [kPa]
- Unmodeled factors may account for errors
 - Road grade
 - Normal load
- Other systematic errors may be present

Conclusions

- C_x and R_e sensitive to a number of factors
- Estimating tire properties more difficult than it first appears
- Linear formulations are biased
- Nonlinear schemes appear to work well
- longitudinal stiffness inversely proportional to inflation pressure

Future Work

- Perform experiments on different tires
- Identify C_x sensitivity to and compensate for
 - Tread depth
 - Normal load
 - Peak road friction
- Improve algorithm efficiency by several orders of magnitude
- Combined sideslip, longitudinal slip parameter estimation

Questions ?