

A Review of: *Minimum-Landing-Error Powered-Descent Guidance for Mars Landing Using Convex Optimization*

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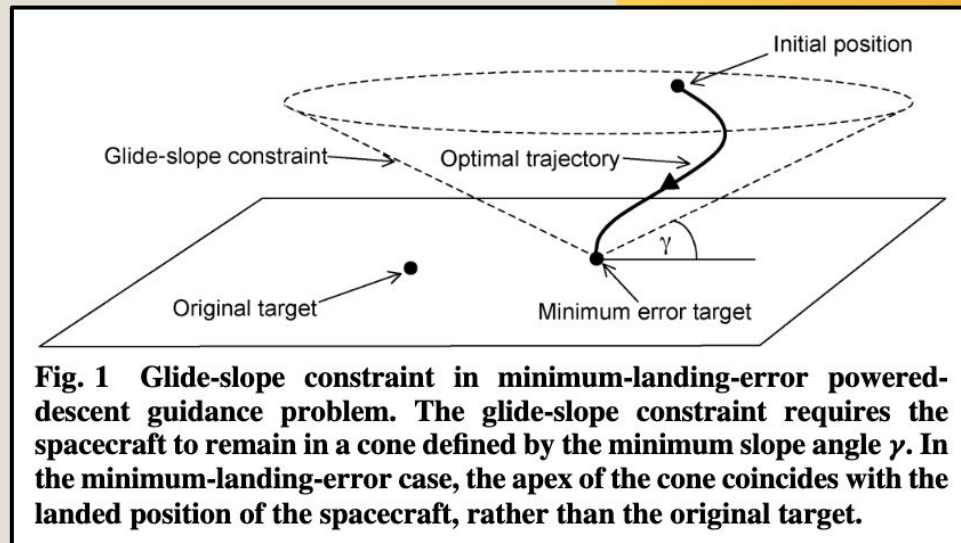
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Description of the System

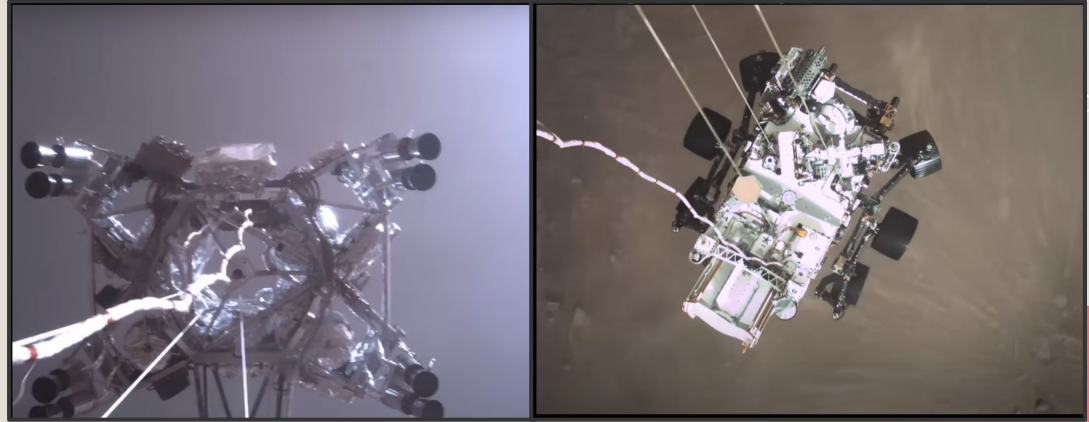
- Mars lander using powered-descent, beginning after parachutes are cut
- Land at some target on the surface
 - Minimum fuel if feasible
 - Minimum landing-error if infeasible
- Improve accuracy of landing guidance algorithms for Mars missions
- Convex-optimization approach → guaranteed globally optimal solution in finite time



Okay, but why is this important?



G-FOLD Diversion Test - NASA Jet Propulsion Laboratory
YouTube



Perseverance rover touching down on Mars. February 18th, 2021

Image: JPL-Caltech/NASA

Minimum-Fuel - Continuous

Cost $\max_{t_f, T_c(\cdot)} m(t_f) = \min_{t_f, T_c(\cdot)} \int_0^{t_f} \|T_c(t)\| dt$

subject to

Dynamics $\ddot{r}(t) = g + T_c(t)/m(t), \quad \dot{m}(t) = -\alpha \|T_c(t)\|$

Min/Max Thrust $0 < \rho_1 \leq \|T_c(t)\| \leq \rho_2, \quad r_1(t) \geq 0$

Glideslope $\|S_j x(t) - v_j\| + c_j^T x(t) + a_j \leq 0$
 $j = 1, \dots, n_s$

Initial/Terminal Conditions $m(0) = m_{\text{wet}}, \quad r(0) = \mathbf{r}_0, \quad \dot{r}(0) = \dot{\mathbf{r}}_0$
 $r(t_f) = \dot{r}(t_f) = 0$

Minimum-Fuel - Discrete Convex

Cost $\min_{N, \eta} \omega^T \eta \rightarrow$ **Fuel usage at each time step**

subject to

Max thrust $\|E_u \Upsilon_k \eta\| \leq e_\sigma^T \Upsilon_k \eta, \quad k = 0, \dots, N$

Min/Max Mass
$$\mu_1(t_k) \left[1 - \{F \mathbf{y}_k - z_0(t_k)\} + \frac{\{F \mathbf{y}_k - z_0(t_k)\}^2}{2} \right] \leq e_\sigma^T \Upsilon_k \eta \leq \mu_2(t_k) (1 - \{F \mathbf{y}_k - z_0(t_k)\})$$

 $k = 1, \dots, N$

$\ln(m_{\text{wet}} - \alpha \rho_2 t_k) \leq F \mathbf{y}_k \leq \ln(m_{\text{wet}} - \alpha \rho_1 t_k)$
 $k = 1, \dots, N$

Glideslope $\|S_j E \mathbf{y}_k - v_j\| + c_j^T E \mathbf{y}_k + a_j \leq 0$
 $k = 1, \dots, N, \quad j = 1, \dots, n_s$

Minimum-Landing-Error - Discrete Convex

Cost $\min_{N,\eta} \|E_r y_N\|^2 \rightarrow$ **Distance from origin**

subject to

Max thrust $\|E_u \Upsilon_k \eta\| \leq \mathbf{e}_4^T \Upsilon_k \eta, \quad k = 0, \dots, N$

Min/Max Mass $\rho_1 e^{-z_0(t_k)} \left[1 - (F \mathbf{y}_k - z_0(t_k)) + \frac{(F \mathbf{y}_k - z_0(t_k))^2}{2} \right]$
 $\leq \mathbf{e}_4^T \Upsilon_k \eta \leq \rho_2 e^{-z_0(t_k)} [1 - (F \mathbf{y}_k - z_0(t_k))]$
 $k = 1, \dots, N$

$E_r \mathbf{y}_k \in \mathbf{X}, \quad k = 1, \dots, N$ **Glideslope**

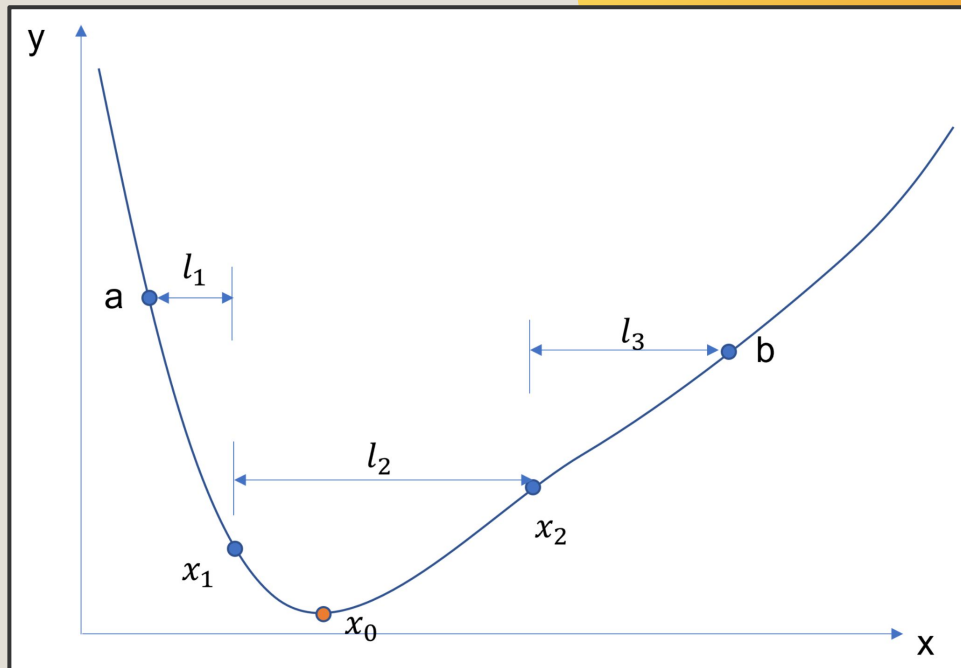
$F y_N \geq \ln m_{\text{dry}}$ **Max Fuel Usage**

$y_N^T \mathbf{e}_1 = 0, \quad E_v y_N^T = 0$ **Final Alt./Vel. = 0**

Dynamics $\mathbf{y}_k = \Phi_k \begin{bmatrix} \mathbf{r}_0 \\ \dot{\mathbf{r}}_0 \\ \ln m_{\text{wet}} \end{bmatrix} + \Lambda_k \begin{bmatrix} \mathbf{g} \\ 0 \end{bmatrix} + \Psi_k \eta \quad k = 1, \dots, N$

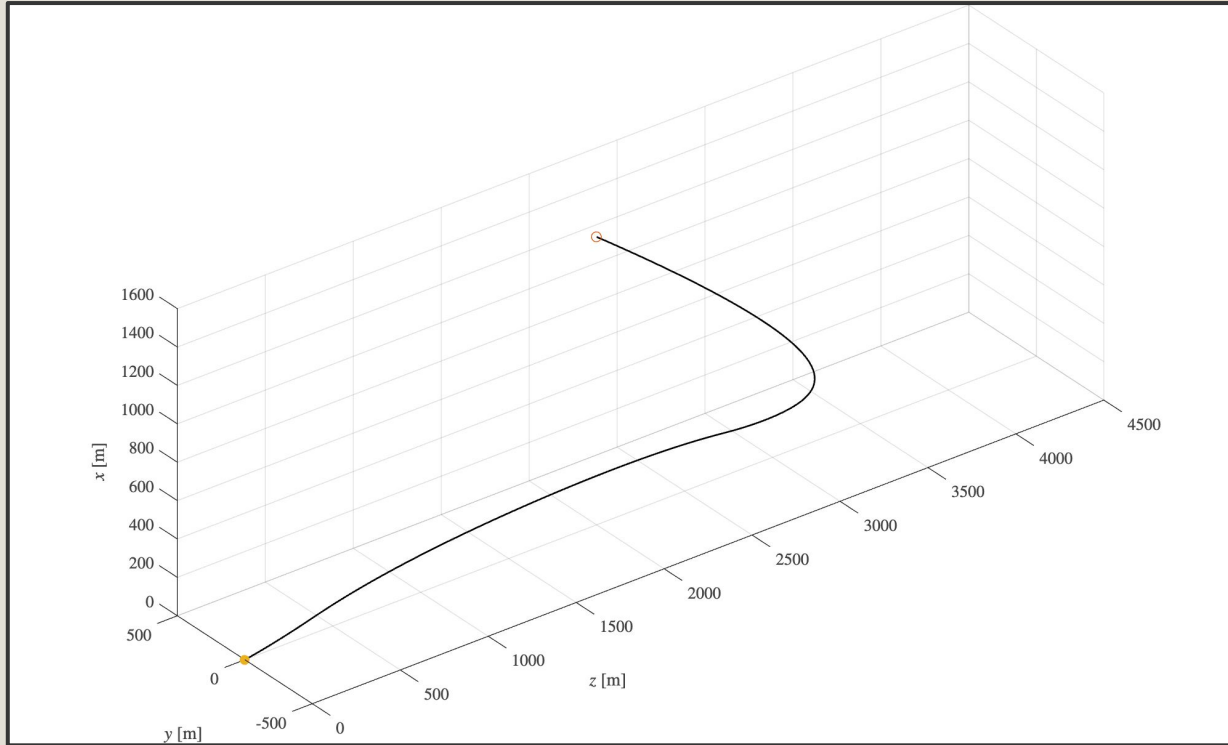
Simulation Methods

- CVX interface in MATLAB to solve second-order cone program
- Optimal fuel/error depends on chosen final time
 - Golden-section search for time
- ~20 seconds from initialization to solution
 - Flight software would use C/C++ → massive speedup
 - Lots of room for optimization in the code still
- Just one part of the larger G-FOLD algorithm

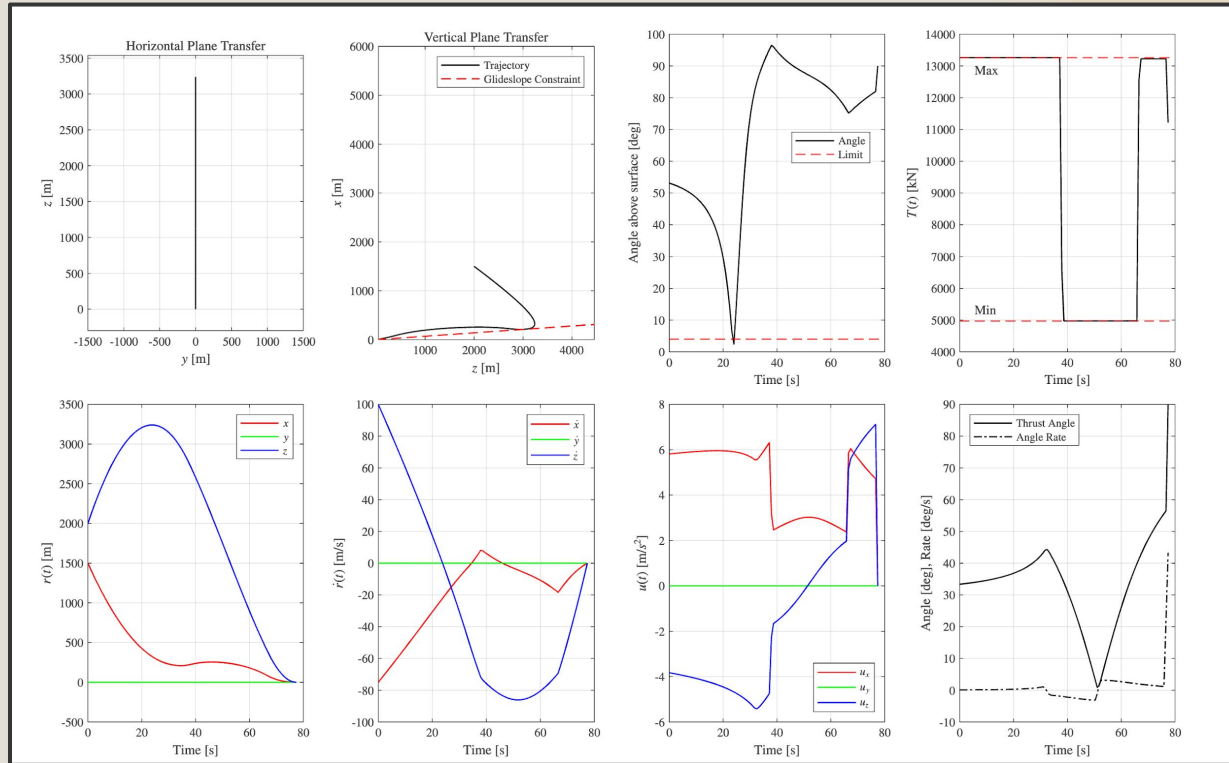


Golden-section search illustration
Yanning Liu, CU Denver

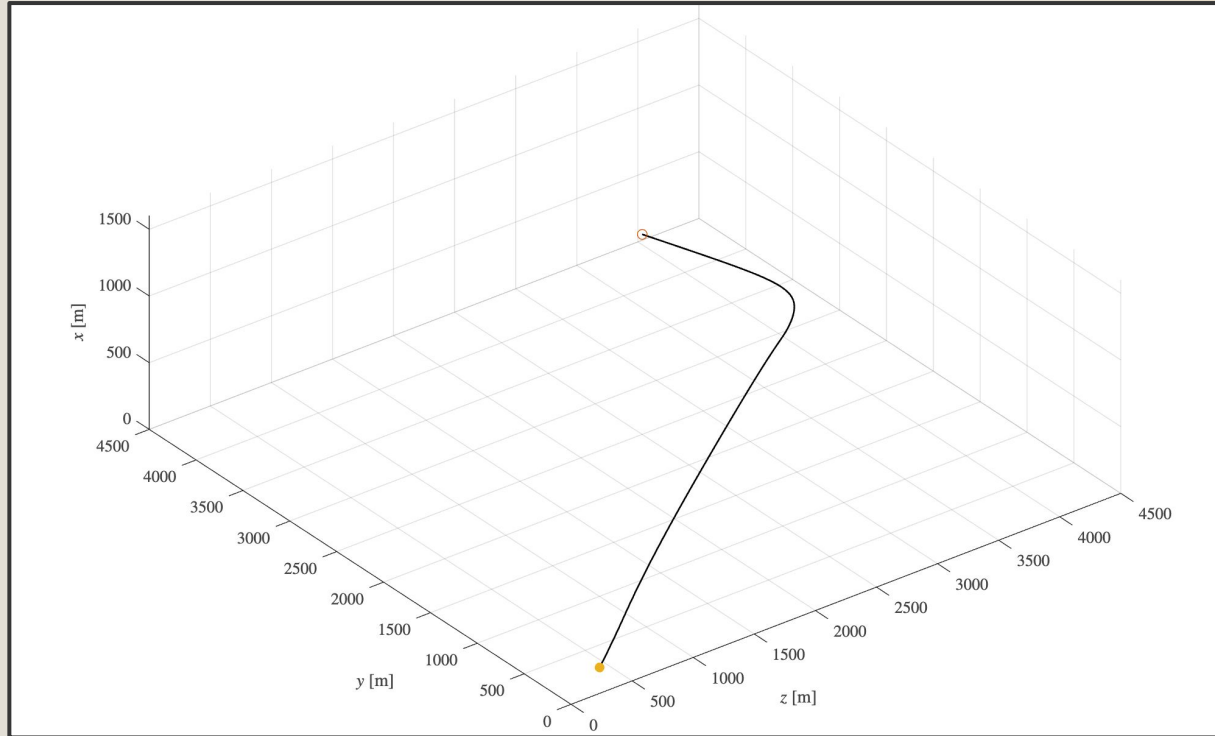
Simulation Results - Fuel



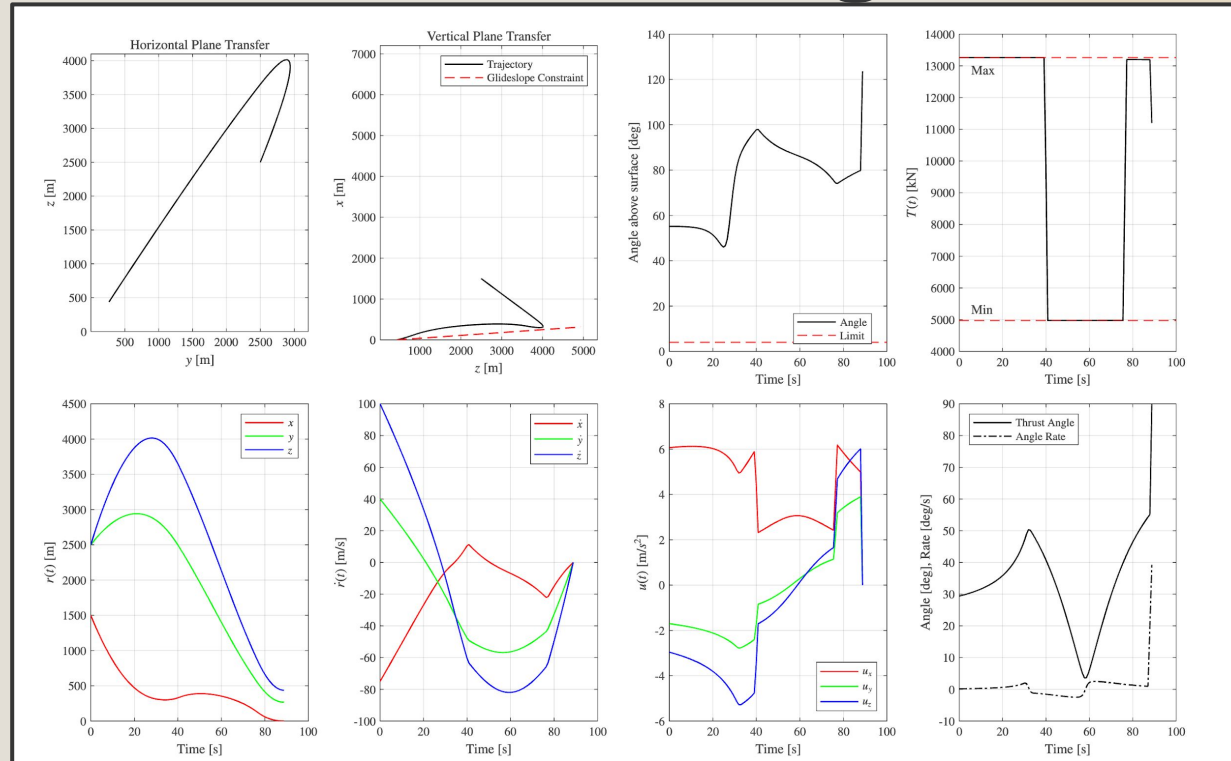
Simulation Results - Fuel



Simulation Results - Landing-Error



Simulation Results - Landing-Error





Thank you