

OpenSim Moco Cheat Sheet for the Matlab Interface

MocoStudy

MocoProblem

Access the MocoProblem from the study.

```
problem = study.updProblem();
```

Set the model.

```
problem.setModel(Model('model_file.osim'));
```

Set variable bounds.

Set initial time to 0; final time between 0.5 and 1.5 s.

```
problem.setTimeBounds(MocoInitialBounds(0),  
MocoFinalBounds(0.5, 1.5));
```

The coordinate value must be between 0 and π over the phase, and its initial value is 0 and its final value is $\pi/2$.

```
problem.setStateInfo('/jointset/j0/q0/value',  
[0, pi], 0, pi/2);
```

The control for actuator '/tau0' must be within [-50, 50] over the phase.

```
problem.setControlInfo('/tau0', [-50, 50]);
```

Optimize static model properties.

Create parameter 'myparam' to optimize the mass of Body '/bodyset/b0' within [0.1, 0.5].

```
problem.addParameter(MocoParameter('myparam',  
'/bodyset/b0', 'mass', MocoBounds(0.1, 0.5)));
```

Add goals to the problem.

Control · ControlTracking · FinalTime
StateTracking · MarkerTracking · TranslationTracking
OrientationTracking · JointReaction

Minimize the sum of squared controls with weight 1.5.

```
problem.addGoal(MocoControlGoal('effort', 1.5));
```

Add path constraints to the problem.

Define time-dependent bounds for controls.

```
pathCon = MocoControlBoundConstraint();  
problem.addPathConstraint(pathCon);
```

MocoSolver

Initialize the CasADI or Tropter solver.

```
solver = study.initCasADiSolver();  
% alternative: solver = study.initTropterSolver();
```

Settings for Tropter and CasADI solvers.

Solve the problem on a grid of 50 mesh intervals.

```
solver.set_num_mesh_intervals(50);
```

Transcribe the optimal control problem with the Hermite-Simpson scheme (alternative: 'trapezoidal').

```
solver.set_transcription_scheme('hermite-simpson');
```

Loosen the convergence and constraint tolerances.

```
solver.set_optim_convergence_tolerance(1e-3);  
solver.set_optim_constraint_tolerance(1e-3);
```

Stop optimization after 500 iterations.

```
solver.set_optim_max_iterations(500);
```

By default, the Hessian is approximated from first derivatives. Set to 'exact' to use an exact Hessian.

```
solver.set_optim_hessian_approximation('exact');
```

Create a guess, randomize it, then set the guess.

```
guess = solver.createGuess(); guess.randomizeAdd();  
solver.setGuess(guess);
```

Set the guess from a MocoTrajectory or MocoSolution file.

```
solver.setGuessFile('previous_solution.sto');
```

Settings for only CasADI solver.

By default, CasADI uses 'central' finite differences; 'forward' differences are faster but less accurate.

```
solver.set_finite_difference_scheme('forward');
```

Turn off parallel calculations.

```
solver.set_parallel(0);
```

Monitor solver progress by writing every 10th iterate to file.

```
solver.set_output_interval(10);
```

Solve the study and obtain a MocoSolution.

```
solution = study.solve();
```

Visualize the solution.

```
study.visualize(solution);
```

Compute outputs from the solution.

```
outputs = StdVectorString();  
outputs.add('.*active_force_length_multiplier');  
table = study.analyze(solution, outputs);
```

MocoTrajectory and MocoSolution

Create a MocoTrajectory.

```
traj = MocoTrajectory('MocoStudy_solution.sto');
traj = MocoTrajectory.createFromStatesControlsTables(
    states, controls);
```

Get time information.

```
traj.getNumTimes();
traj.getInitialTime(); traj.getFinalTime();
traj.getTimeMat();
```

Get names of variables.

```
traj.getStateNames(); traj.getControlNames();
traj.getMultiplierNames(); traj.getParameterNames();
```

Get the trajectory/value for a single variable by name.

```
traj.getStateMat(name); traj.getControlMat(name);
traj.getMultiplierMat(name); traj.getParameter(name);
```

Get the trajectories/values for all variables of a given type.

```
traj.getStatesTrajectoryMat();
traj.getControlsTrajectoryMat();
traj.getMultipliersTrajectoryMat();
traj.getParametersMat();
```

Change the number of times in the trajectory.

```
traj.resampleWithNumTimes(150);
```

Set variable values.

```
traj.setTime(times)
traj.setState(stateTraj); traj.setControl(controlTraj);
traj.setParameter(value);
traj.setStatesTrajectory(statesTraj);
traj.insertStatesTrajectory(subsetStates);
```

Randomize the variable values.

```
traj.randomizeAdd();
```

Export the trajectory.

```
traj.write('mocotrajectory.sto');
traj.exportToStatesTable()
traj.exportToStatesTrajectory(mocoProblem)
```

Compare two trajectories.

```
traj.isNumericallyEqual(otherTraj);
traj.compareContinuousVariablesRMS(otherTraj);
traj.compareParametersRMS(otherTraj);
```

Functions on only MocoSolution.

```
solution.success(); solution.getStatus();
solution.getObjective(); solution.getNumIterations();
solution.getSolverDuration();
solution.unseal(); % Access a failed solution.
```

The Moco Optimal Control Problem

$$\text{minimize} \quad \sum_j w_j J_j \left(t_0, t_f, y_0, y_f, x_0, x_f, \lambda_0, \lambda_f, p, \int_{t_0}^{t_f} s_{c,j}(t, y, x, \lambda, p) dt \right) \quad \text{problem.addGoal()}$$

subject to $\dot{q} = u$

$$M(q, p)\dot{u} + G(q, p)^T \lambda = f_{\text{app}}(t, y, x, p) - f_{\text{bias}}(q, u, p)$$

$$\dot{z}_{\text{ex}}(t) = f_{\text{aux,ex}}(t, y, x, \lambda, p) \quad 0 = f_{\text{aux,im}}(t, y, \dot{z}_{\text{im}}, x, \lambda, p)$$

$$0 = \phi(q, p)$$

$$0 = \nu(q, u, p)$$

$$0 = \alpha(q, u, \dot{u}, p)$$

$$g_L \leq g(t, y, x, \lambda, p) \leq g_U$$

`problem.setModel()`

`problem.addPathConstraint()`

`problem.addGoal()`

`problem.setStateInfo()`

`problem.setControlInfo()`

`problem.setTimeBounds()`

`problem.addParameter()`

with respect to $y \in [y_L, y_U]$ $x \in [x_L, x_U]$

$$t_0 \in [t_{0,L}, t_{0,U}] \quad t_f \in [t_{f,L}, t_{f,U}]$$

$$p \in [p_L, p_U]$$

t time
 $q(t)$ generalized coordinates
 $u(t)$ generalized speeds

w_j	weight for the j -th cost
J_j	the j -th cost
$s_{c,j}$	integrand used in the j -th cost
M	mass matrix
f_{bias}	centripetal and coriolis forces
G	kinematic constraint Jacobian
f_{app}	applied forces
f_{aux}	auxiliary dynamics (explicit & implicit)

ϕ	position-level constraints
ν	velocity-level constraints
α	acceleration-level constraints
g	path constraints
V_k	the k -th endpoint constraint
K	number of endpoint constraints
$s_{e,k}$	integrand in k -th endpoint con.
$()_U$	an upper bound
$()_L$	a lower bound

$y(t)$ $(q(t), u(t), z(t))$

$x(t)$ controls

p constant parameters

λ kinematic constraint multipliers