

# MocoExtendProblem: Interface Between OpenSim and MATLAB for Rapidly Developing Direct Collocation Goals in Moco

Aravind Sundararajan<sup>1</sup>✉, Varun Joshi<sup>2</sup>, Brian Umberger<sup>2</sup>, and Matthew O'Neill<sup>1</sup>

<sup>1</sup> Department of Anatomy, Midwestern University, Glendale Arizona, USA <sup>2</sup> School of Kinesiology, University of Michigan, Ann Arbor, Michigan, USA ✉ Corresponding author

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## Summary

optimal control problems using OpenSim Moco ([Dembia, 2021](#)) and MATLAB (The Mathworks, Inc., Natick, MA, USA). MEP features several templates for testing and prototyping novel MocoGoals in lieu of rebuilding OpenSim or generating an .omoco file from C++ to load the problem into MATLAB. Instead, users structure custom goals, build them, and call custom goals from MATLAB scripts.

This repository features:

- A build.m script that compiles goals in the custom\_goals directory and procedurally constructs the C++/MATLAB class implementations and compiles the MEX interface.
- Compatibility tested with OpenSim 4.2-4.5.
- OpenSim versions lower than 4.5 require unique modifications to the build pipeline since booleans for division by duration, distance and mass were migrated to the abstract MocoGoal.
- The ability to include MEP as a submodule, build, and use valid custom goals.
- Three example custom goals in the custom\_goals and custom\_goals\_compat directories.

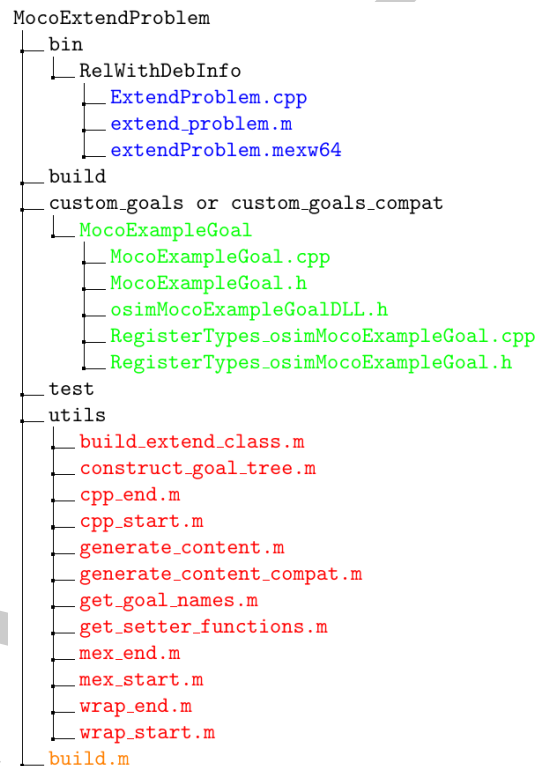
## Statement of need

OpenSim is an open-source software platform for modeling musculoskeletal structures and creating dynamic simulations of movement ([Seth, 2018](#)). OpenSim enables researchers and clinicians to investigate how biological and non-biological structures respond to different loads, postures and activities in both static and dynamic situations. OpenSim has been used to study a wide range of biomechanical problems, such as the mechanics of walking and running ([Falisse et al., 2019](#)), the impact of injury or disease on movement ([Johnson, 2022](#)), and the effectiveness of rehabilitation exercises ([Spomer et al., 2023](#)).

OpenSim Moco ([Dembia, 2021](#)) employs an optimization paradigm called direct collocation to solve trajectory optimization problems that range from solving for muscle forces, to tracking experimental data, and fully predictive simulations. Direct collocation is a numerical optimal control method ([Kelly, 2017](#)) that is computationally efficient and is used extensively in computational approaches to understanding biological movement. While direct collocation is powerful, Moco only provides a fixed set of optimization goals. It can be daunting for many users to develop custom goals in C++. We developed MEP so Moco users without experience compiling C++ can still write and test custom goals. The OpenSim interfaces are created with SWIG, as opposed to MEX, which can be daunting for even experienced biomechanists.

40 MocoExtendProblem was developed using MATLAB versions 2022a. Running build.m will  
41 compile MocoGoals in the custom\_goals directory, or in the custom\_goals\_compat directory  
42 for OpenSim versions pre-4.5.

43 CMake and msbuild from Visual Studio 2019 or higher must be added to the system PATH.  
44 build.m will procedurally construct both extend\_problem.m and ExtendProblem.cpp by parsing  
45 the header files of the discovered goals within the custom\_goals directory. Both ExtendProb-  
46 lem.cpp and extend\_problem.m generate bindings to instantiate custom goals placed in the  
47 custom\_goals directory. Custom goals can be compiled with Visual Studio 2019 or higher  
48 and then MATLAB's MEX compiler is used to compile ExtendProblem. ExtendProblem.cpp  
49 leverages the C++ library mexplus (Yamaguchi, 2018) to gain access to MEX entry points  
50 through C++ macros.



**Figure 1:** MEP Framework. The researcher runs the build.m script (orange) that subsequently calls methods in the utils folder (red) which are tasked with reading the custom\_goals folder (green) and procedurally construct the mex and the interface class that calls the mex (blue). Each custom goal (green) is handled as its own compiled plugin.

51 To create a new goal with MEP:

- 52 1. OpenSim 4.5+ users should copy a goal in the custom\_goals directory while 4.2-4.4
- 53 users should copy a goal in custom\_goals\_compat.
- 54 2. Replace mentions of the original goal name to that of your new custom goal name in
- 55 each of the 5 files and file names, being careful to also modify the include guards in the
- 56 dll and register types header files.
- 57 3. Reimplement constructProperties(), initializeOnModelImpl(), calcIntegrandImpl(), cal-
- 58 cGoalImpl() such that they describe your custom goal.

59 To incorporate extend\_problem goals into an existing MATLAB script, a C-style pointer to the  
60 instantiated MocoProblem is passed as a constructor argument to the extend\_problem.m class  
61 that wraps the MEP MEX. Class methods of extend\_problem.m (Figure 1; blue) are then used

62 to add custom goals to the MocoProblem.

```
63 cptr = uint64(problem.getCPtr(problem));
64 ep = extend_problem(cptr);
65 ep.addMocoCustomGoal('custom_goal',weight,power,divide_by_distance);
```

66 This paradigm has implications for OpenSim and MATLAB developers beyond the scope of  
67 just incorporating novel MocoGoals; these same tools can be used to extend other classes  
68 and easily incorporate them into existing MATLAB-OpenSim scripts. We have posted all  
69 tools, instructions and simulation results related to this project on GitHub and SimTK.org  
70 ([simtk.org/projects/moco-ep](https://simtk.org/projects/moco-ep)).

## 71 Requirements

- 72 ■ Download and install OpenSim from <https://simtk.org> and follow the documentation for  
73 setting up OpenSim's MATLAB scripting environment.
- 74 ■ Follow the instructions (OpenSim) to download necessary dependencies for both scripting  
75 in MATLAB and C++ development.
- 76 ■ In MATLAB, configure MEX with `mex -setup C++` to use the MS VisualStudio 2019+.

## 77 Showcases

78 To demonstrate the utility of this framework, we generated a two-dimensional (2-D) walking  
79 simulation using the MATLAB-OpenSim API ([Denton & Umberger, 2023](#)). The base code  
80 uses the built-in MocoControlEffortGoal and MocoAverageSpeedGoal to generate tracking and  
81 predictive simulations of minimum effort walking at an average speed of 1.3 m s<sup>-1</sup>. Additionally,  
82 each objective function includes implicit acceleration and auxiliary derivative terms that are  
83 minimized to ensure smooth trajectories.

84 Since Moco lacks built-in gait stability goals, we developed three stability goals using MEP  
85 build.m to create an ExtendProblem class that adds these to an existing MocoProblem ([Figure 1](#);  
86 blue). The first is a base of support ([Equation 1](#)) criterion in which the whole-body center  
87 of mass (COM) is optimized to lay between the two hindfeet COMs projected to the ground  
88 reference frame, the second is a zero-moment-point goal ([Equation 2](#)) where the center of  
89 mass tracks the computed zero-tilting moment location, and the third is a marker acceleration  
90 minimization goal ([Equation 3](#)) that minimizes the explicit accelerations of a marker placed on  
91 the head (marker location is arbitrary and can be set by the user).

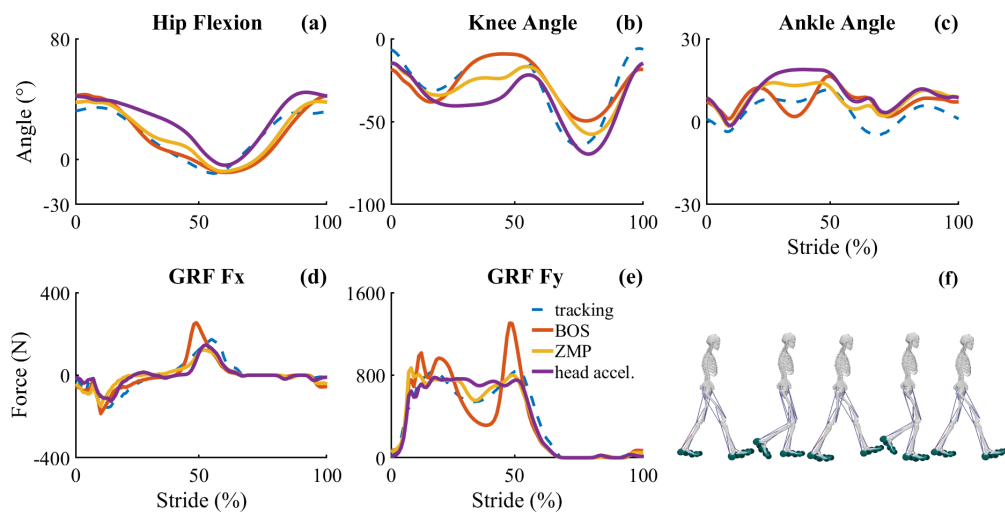
92 MEP's build.m was used to generate an ExtendProblem class that adds these new stability  
93 cost terms:

$$J_{BOS} = W_1 EFF^2 + W_2 ACC_{smoothing} + W_3 BOS \quad (1)$$

$$J_{zmp} = W_1 EFF^2 + W_2 ACC_{smoothing} + W_3 ZMP \quad (2)$$

$$J_{acc} = W_1 EFF^2 + W_2 ACC_{smoothing} + W_3 ACC_{marker} \quad (3)$$

94 The results of each multi-objective predictive simulation, in which the stability criterion was  
95 compiled using MEP, is shown against the results from a tracking simulation [[Figure 2](#); ??]  
96 that closely-matched experimental data ([Denton & Umberger, 2023](#)). As the purpose was  
97 to demonstrate the utility of MEP, we did not tune the stability term weights to match the  
98 tracking result as closely as possible.



**Figure 2:** Sagittal plane hip, knee and ankle angles (a-c), vertical and A-P ground reaction forces (d-e), the 11 degree-of-freedom, 18 muscle sagittal plane human walking model used for tracking and predictive simulations (f)

	Objective cost	Effort cost	Smoothing cost	Stability cost
$J_{BOS}$	3.759046	2.270912	0.683608	0.794155
$J_{ZMP}$	4.184254	2.751212	0.725837	0.686290
$J_{accel}$	4.774932	3.797785	0.793123	0.174308

99 [Table: Objective cost and term breakdown for three predictive simulations using MEP {table1}.]

While these examples used planar gait simulations, MEP is agnostic to model complexity or task, and is being used successfully in our ongoing research (Sundararajan, 2023; Joshi2022?) of locomotor performance in humans and other animals. GNU Octave support would require minimal syntactical modification. An additional benefit of sequestering novel goals into ExtendProblem is being able to back-port goals from a newer OpenSim version to an older version (i.e. taking a goal from OpenSim 4.4 and bringing that functionality to 4.2). Ultimately, MEP offers a modular framework to rapidly develop, test and compare novel MocoGoals for features beyond OpenSim Moco's current scope.

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