```
In [2]: # here is how we activate an environment in our current directory
import Pkg; Pkg.activate(@__DIR__)

# instantate this environment (download packages if you haven't)
Pkg.instantiate();

# let's load LinearAlgebra in
using LinearAlgebra
using Test
```

Activating environment at `~/Dropbox/My Mac (MacBook Pro (2))/Desktop/CM				
U/Optimal Control/HWO_S23/Project.toml`				
<pre>Updating registry at `~/.julia/registries/General`</pre>				
Installed	Zstd_jll ————	v1.5.2+0		
Installed	Cairo_jll ————	v1.16.1+1		
	FriBidi_jll			
	LogExpFunctions ————			
	BenchmarkTools ————			
	FiniteDiff ————			
	ForwardDiff ————			
	Plots ————			
	Xorg_libXext_jll			
	Xorg_libXi_jll ————			
	FFMPEG ————			
	JLFzf ————			
	Showoff —			
	FixedPointNumbers ————			
Installed	Requires ————	v1.3.0		
	Libglvnd_jll			
	Libffi_jll			
Installed	Libtiff_jll	v4.4.0+0		
	OpenSSL_jll			
	DiffRules ————			
	Scratch ————			
Installed	CommonSubexpressions ————	v0.3.0		
Installed	XML2_jll ———————————————————————————————————	v2.10.3+0		
	SimpleBufferStream ————			
	PlotThemes ————			
	OpenSpecFun_jll			
	ConstructionBase ————			
	OrderedCollections ————			
Installed	Xorg_libXinerama_jll	v1.1.4+4		
Installed	Grisu ————	v1.0.2		
	Libmount_jll			
	StaticArraysCore ————			
	Measures ————			
	BitFlags ————			
	Colors ————			
	<pre>Xorg_libXfixes_jll</pre>			
	RecipesBase ————			
Installed	<pre>Xorg_xcb_util_wm_jll</pre>	v0.4.1+1		
Installed	XSLT_jll ———————————————————————————————————	v1.1.34+0		
Installed	Bzip2_jll	v1.0.8+0		
Installed	Ogg_jll	v1.3.5+1		
	Xorg_libXdmcp_jll			
	StatsAPI ————			
	LAME_jll			
	Xorg_libXrender_jll ————			
Installed	HarfBuzz_jll	v2.8.1+1		
Installed	libvorbis_jll	v1.3.7+1		
Installed	DataAPI ————	v1.14.0		
	LoggingExtras ————			
	ChainRulesCore ————			
	Xorg_libX11_jll			
Installed	Glib_jll	v2.74.0+2		

Q1

Installed	x264_jll	v2021.5.5+0
	libpng_jll ————	
	IrrationalConstants ————	
	ChangesOfVariables ————	
	Setfield ————	
	Formatting ————	
	GLFW_jll ———	
	libass_jll ————	
	<pre>Xorg_xcb_util_renderutil_jll -</pre>	
	Pipe —————	
	DiffResults ————	
	TensorCore ———	
Installed	GR ———	v0.71.3
	LaTeXStrings ————	
	CodecZlib ————	
	Xorg_libXau_jll ————	
	x265_jll	
	Missings —	
	Unzip ————	
Installed	DocStringExtensions ————	v0.9.3
	DataStructures ————	
Installed	<pre>Xorg_xcb_util_jll</pre>	v0.4.0+1
	Xorg_xkbcomp_jll —	
	xkbcommon_jll	
Installed	Graphite2_jll	v1.3.14+0
Installed	ColorTypes ————	v0.11.4
Installed	RelocatableFolders ————	v1.0.0
Installed	Contour —	v0.6.2
	HTTP ————	
	Gettext_jll	
	NaNMath ————	
	TranscodingStreams ————	
Installed	SortingAlgorithms ————	v1.1.0
Installed	InverseFunctions ————	v0.1.8
	Latexify ————	
Installed	StaticArrays —	v1.5.12
	libfdk_aac_jll ————	
	Xorg_libxcb_jll —	
	<pre>Xorg_xcb_util_keysyms_jll</pre>	
	JpegTurbo_jll ————	
	FFMPEG_jll ———————————————————————————————————	
	StatsBase ——————	
	Xorg_libxkbfile_jll ————	
Installed	Onus ill	VI.I.0+4
Instatted	Opus_jll ———————————————————————————————————	V1.5.2+0
	Libgpg_error_jll —————	
	Xorg_libXcursor_jll ————	
	fzf_jll ————	
	UnicodeFun ———————	
	Xorg_xcb_util_image_jll ———	
Installed	Libacrynt ill —————	v1.8.7+0
Installed	Libgcrypt_jll ———————————————————————————————————	v2.10.1+0
	ArrayInterfaceCore ————	
	PlotUtils ———	
	OpenSSL ————	
	•	

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```
Installed URIs ----- v1.4.1

        Installed FreeType2_jll
        v2.10.4+0

        Installed libaom_jll
        v3.4.0+0

    Installed Expat jll — v2.4.8+0
   Installed Expat_jtt V2.4.8+0
Installed Xorg_xtrans_jtl V1.4.0+3
Installed Libuuid_jtl V2.36.0+0
Installed ColorSchemes V3.20.0
Installed SpecialFunctions V2.1.7

        Installed Special Functions
        V2.1.7

        Installed Fontconfig_jll
        V2.13.93+0

        Installed Wayland_jll
        V1.21.0+0

        Installed ColorVectorSpace
        V0.9.10

        Installed Compat
        V4.5.0

        Installed GR_jll
        V0.71.3+0

        Installed GR_jll
        v0./1.3+0

        Installed Reexport
        v1.2.2

        Installed MacroTools
        v0.5.10

        Installed RecipesPipeline
        v0.6.11

        Installed LERC_jll
        v3.0.0+1

        Installed Libiconv_jll
        v1.16.1+2

        Installed Xorg_xkeyboard_config_jll
        v2.27.0+4

   Updating `~/Dropbox/My Mac (MacBook Pro (2))/Desktop/CMU/Optimal Contro
l/HW0 S23/Project.toml`
   [6e4b80f9] + BenchmarkTools v1.3.2
   [6a86dc24] + FiniteDiff v2.17.0
   [f6369f11] + ForwardDiff v0.10.34
   [91a5bcdd] + Plots v1.38.2
     Updating `~/Dropbox/My Mac (MacBook Pro (2))/Desktop/CMU/Optimal Contro
l/HW0 S23/Manifest.toml`
   [30b0a656] + ArrayInterfaceCore v0.1.28
   [6e4b80f9] + BenchmarkTools v1.3.2
   [d1d4a3ce] + BitFlags v0.1.7
   [d360d2e6] + ChainRulesCore v1.15.7
   [9e997f8a] + ChangesOfVariables v0.1.4
   [944b1d66] + CodecZlib v0.7.0
   [35d6a980] + ColorSchemes v3.20.0
   [3da002f7] + ColorTypes v0.11.4
   [c3611d14] + ColorVectorSpace v0.9.10
   [5ae59095] + Colors v0.12.10
   [bbf7d656] + CommonSubexpressions v0.3.0
   [34da2185] + Compat v4.5.0
   [187b0558] + ConstructionBase v1.4.1
   [d38c429a] + Contour v0.6.2
   [9a962f9c] + DataAPI v1.14.0
   [864edb3b] + DataStructures v0.18.13
   [163ba53b] + DiffResults v1.1.0
   [b552c78f] + DiffRules v1.12.2
   [ffbed154] + DocStringExtensions v0.9.3
   [c87230d0] + FFMPEG v0.4.1
   [6a86dc24] + FiniteDiff v2.17.0
   [53c48c17] + FixedPointNumbers v0.8.4
   [59287772] + Formatting v0.4.2
   [f6369f11] + ForwardDiff v0.10.34
   [28b8d3ca] + GR v0.71.3
   [42e2da0e] + Grisu v1.0.2
```

```
[cd3eb016] + HTTP v1.7.3
[83e8ac13] + IniFile v0.5.1
[3587e190] + InverseFunctions v0.1.8
[92d709cd] + IrrationalConstants v0.1.1
[1019f520] + JLFzf v0.1.5
[692b3bcd] + JLLWrappers v1.4.1
[682c06a0] + JSON v0.21.3
[b964fa9f] + LaTeXStrings v1.3.0
[23fbe1c1] + Latexify v0.15.18
[2ab3a3ac] + LogExpFunctions v0.3.19
[e6f89c97] + LoggingExtras v1.0.0
[1914dd2f] + MacroTools v0.5.10
[739be429] + MbedTLS v1.1.7
[442fdcdd] + Measures v0.3.2
[e1d29d7a] + Missings v1.1.0
[77ba4419] + NaNMath v1.0.1
[4d8831e6] + OpenSSL v1.3.3
[bac558e1] + OrderedCollections v1.4.1
[69de0a69] + Parsers v2.5.3
[b98c9c47] + Pipe v1.3.0
[ccf2f8ad] + PlotThemes v3.1.0
[995b91a9] + PlotUtils v1.3.2
[91a5bcdd] + Plots v1.38.2
[21216c6a] + Preferences v1.3.0
[3cdcf5f2] + RecipesBase v1.3.3
[01d81517] + RecipesPipeline v0.6.11
[189a3867] + Reexport v1.2.2
[05181044] + RelocatableFolders v1.0.0
[ae029012] + Requires v1.3.0
[6c6a2e73] + Scratch v1.1.1
[efcf1570] + Setfield v1.1.1
[992d4aef] + Showoff v1.0.3
[777ac1f9] + SimpleBufferStream v1.1.0
[66db9d55] + SnoopPrecompile v1.0.3
[a2af1166] + SortingAlgorithms v1.1.0
[276daf66] + SpecialFunctions v2.1.7
[90137ffa] + StaticArrays v1.5.12
[1e83bf80] + StaticArraysCore v1.4.0
[82ae8749] + StatsAPI v1.5.0
[2913bbd2] + StatsBase v0.33.21
[62fd8b95] + TensorCore v0.1.1
[3bb67fe8] + TranscodingStreams v0.9.11
[5c2747f8] + URIs v1.4.1
[1cfade01] + UnicodeFun v0.4.1
[41fe7b60] + Unzip v0.1.2
[6e34b625] + Bzip2_jll v1.0.8+0
[83423d85] + Cairo_jll v1.16.1+1
[2e619515] + Expat_jll v2.4.8+0
[b22a6f82] + FFMPEG_jll v4.4.2+2
[a3f928ae] + Fontconfig ill v2.13.93+0
[d7e528f0] + FreeType2_jll v2.10.4+0
[559328eb] + FriBidi_jll v1.0.10+0
[0656b61e] + GLFW jll v3.3.8+0
[d2c73de3] + GR_jll v0.71.3+0
[78b55507] + Gettext_jll v0.21.0+0
[7746bdde] + Glib jll v2.74.0+2
```

O1

```
[3b182d85] + Graphite2 jll v1.3.14+0
[2e76f6c2] + HarfBuzz jll v2.8.1+1
[aacddb02] + JpeqTurbo jll v2.1.2+0
[c1c5ebd0] + LAME_jll v3.100.1+0
[88015f11] + LERC_jll v3.0.0+1
[dd4b983a] + LZO jll v2.10.1+0
[e9f186c6] + Libffi_jll v3.2.2+1
[d4300ac3] + Libgcrypt_jll v1.8.7+0
[7e76a0d4] + Libglvnd jll v1.6.0+0
[7add5ba3] + Libgpg_error_jll v1.42.0+0
[94ce4f54] + Libiconv_jll v1.16.1+2
[4b2f31a3] + Libmount jll v2.35.0+0
[89763e89] + Libtiff_jll v4.4.0+0
[38a345b3] + Libuuid jll v2.36.0+0
[e7412a2a] + Ogg jll v1.3.5+1
[458c3c95] + OpenSSL jll v1.1.19+0
[efe28fd5] + OpenSpecFun_jll v0.5.5+0
[91d4177d] + Opus_jll v1.3.2+0
[30392449] + Pixman jll v0.40.1+0
[ea2cea3b] + Qt5Base_jll v5.15.3+2
[a2964d1f] + Wayland_jll v1.21.0+0
[2381bf8a] + Wayland protocols jll v1.25.0+0
[02c8fc9c] + XML2_jll v2.10.3+0
[aed1982a] + XSLT_jll v1.1.34+0
[4f6342f7] + Xorg libX11 jll v1.6.9+4
[0c0b7dd1] + Xorq libXau jll v1.0.9+4
[935fb764] + Xorg_libXcursor_jll v1.2.0+4
[a3789734] + Xorq libXdmcp jll v1.1.3+4
[1082639a] + Xorg_libXext_jll v1.3.4+4
[d091e8ba] + Xorg_libXfixes_jll v5.0.3+4
[a51aa0fd] + Xorq libXi jll v1.7.10+4
[d1454406] + Xorq libXinerama jll v1.1.4+4
[ec84b674] + Xorg_libXrandr_jll v1.5.2+4
[ea2f1a96] + Xorg libXrender jll v0.9.10+4
[14d82f49] + Xorg_libpthread_stubs_jll v0.1.0+3
[c7cfdc94] + Xorq libxcb jll v1.13.0+3
[cc61e674] + Xorq libxkbfile jll v1.1.0+4
[12413925] + Xorg xcb util image jll v0.4.0+1
[2def613f] + Xorg_xcb_util_jll v0.4.0+1
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[0d47668e] + Xorg_xcb_util_renderutil_jll v0.3.9+1
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[33bec58e] + Xorg xkeyboard config jll v2.27.0+4
[c5fb5394] + Xorg_xtrans_jll v1.4.0+3
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[214eeab7] + fzf_jll v0.29.0+0
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[dfaa095f] + x265 ill v3.5.0+0
[d8fb68d0] + xkbcommon_jll v1.4.1+0
[0dad84c5] + ArgTools
```

O1

```
[56f22d72] + Artifacts
[2a0f44e3] + Base64
[ade2ca70] + Dates
[8bb1440f] + DelimitedFiles
[f43a241f] + Downloads
[9fa8497b] + Future
[b77e0a4c] + InteractiveUtils
[b27032c2] + LibCURL
[76f85450] + LibGit2
[8f399da3] + Libdl
[37e2e46d] + LinearAlgebra
[56ddb016] + Logging
[d6f4376e] + Markdown
[a63ad114] + Mmap
[ca575930] + NetworkOptions
[44cfe95a] + Pkg
[de0858da] + Printf
[9abbd945] + Profile
[3fa0cd96] + REPL
[9a3f8284] + Random
[ea8e919c] + SHA
[9e88b42a] + Serialization
[6462fe0b] + Sockets
[2f01184e] + SparseArrays
[10745b16] + Statistics
[4607b0f0] + SuiteSparse
[fa267f1f] + TOML
[a4e569a6] + Tar
[8dfed614] + Test
[cf7118a7] + UUIDs
[4ec0a83e] + Unicode
[e66e0078] + CompilerSupportLibraries_jll
[deac9b47] + LibCURL_jll
[29816b5a] + LibSSH2 jll
[c8ffd9c3] + MbedTLS_jll
[14a3606d] + MozillaCACerts_jll
[05823500] + OpenLibm jll
[efcefdf7] + PCRE2 jll
[83775a58] + Zlib_jll
[8e850ede] + nghttp2_jll
[3f19e933] + p7zip_jll
```

Q1

Question 1: Differentiation in Julia (10 pts)

Julia has a fast and easy to use forward-mode automatic differentiation package called ForwardDiff.jl that we will make use of throughout this course. In general it is easy to use and very fast, but there are a few quirks that are detailed below. This notebook will start by walking through general usage for the following cases:

- functions with a single input
- functions with multiple inputs
- composite functions

as well as a guide on how to avoid the most common ForwardDiff.jl error caused by creating arrays inside the function being differentiated. First, let's look at the ForwardDiff.jl functions that we are going to use:

- FD.derivative(f,x) derivative of scalar or vector valued f wrt scalar x
- FD.jacobian(f,x) jacobian of vector valued f wrt vector x
- FD.gradient(f,x) gradient of scalar valued f wrt vector x
- FD.hessian(f,x) hessian of scalar valued f wrt vector x

Note on gradients:

For an arbitrary function $f(x):\mathbb{R}^N o \mathbb{R}^M$, the jacobian is the following:

$$rac{\partial f(x)}{\partial x} = \left[egin{array}{cccc} rac{\partial f_1}{\partial x_1} & \cdots & rac{\partial f_1}{\partial x_n} \ dots & \ddots & dots \ rac{\partial f_m}{\partial x_1} & \cdots & rac{\partial f_m}{\partial x_n} \end{array}
ight]$$

Now if we have a scalar valued function (like a cost function) $f(x):\mathbb{R}^N \to \mathbb{R}$, the jacobian is the following row vector:

$$rac{\partial f(x)}{\partial x} = \left[egin{array}{ccc} rac{\partial f_1}{\partial x_1} & \cdots & rac{\partial f_1}{\partial x_n} \end{array}
ight]$$

The transpose of this jacobian for scalar valued functions is called the gradient:

$$abla f(x) = \left[rac{\partial f(x)}{\partial x}
ight]^T$$

TLDR:

- the jacobian of a scalar value function is a row vector
- the gradient is the transpose of this jacobian, making the gradient a column vector
- ForwardDiff.jl will give you an error if you try to take a jacobian of a scalar valued function, use the gradient function instead

Part (a): General usage (2 pts)

The API for functions with one input is detailed below:

In [5]: # NOTE: this block is a tutorial, you do not have to fill anything out.
#-----load the package-----# using ForwardDiff # this puts all exported functions into our namespace
import ForwardDiff # this means we have to use ForwardDiff.<function name>
import ForwardDiff as FD # this let's us do FD.

```
function foo1(x)
                              #scalar input, scalar output
                               return sin(x)*cos(x)^2
                     end
                     function foo2(x)
                              # vector input, scalar output
                               return sin(x[1]) + cos(x[2])
                    end
                     function foo3(x)
                              # vector input, vector output
                               return [\sin(x[1])*x[2];\cos(x[2])*x[1]]
                     end
                    let # we just use this to avoid creating global variables
                              # evaluate the derivative of fool at x1
                              x1 = 5*randn():
                              @show \partial foo1_{\partial x} = FD_{derivative}(foo1, x1);
                              # evaluate the gradient and hessian of foo2 at x2
                              x2 = 5*randn(2);
                              @show \nablafoo2 = FD.gradient(foo2, x2);
                              @show \nabla^2 foo2 = FD.hessian(foo2, x2);
                              # evluate the jacobian of foo3 at x2
                              @show \partial foo3_{\partial x} = FD_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo3_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(foo)_{ijacobian}(fo
                    end
                    \partial foo1_{\partial x} = FD.derivative(foo1, x1) = -0.9846052886550892
                    \nablafoo2 = FD.gradient(foo2, x2) = [-0.5003595293384151, -0.23209750704668589]
                    \nabla^2 foo2 = FD.hessian(foo2, x2) = [0.8658177298948317 0.0; 0.0 -0.97269252450
                    232881
                     \partial foo3_{\partial x} = FD_{iacobian}(foo3, x2) = [3.0266506678606255 -0.8658177298948317;
                    0.9726925245023288 - 0.9721113975363671
Out[5]: 2×2 Matrix{Float64}:
                       3.02665 -0.865818
                       0.972693 - 0.972111
In [8]: # here is our function of interest
                     function foo4(x)
                              Q = diagm([1;2;3.0]) # this creates a diagonal matrix from a vector
                               return 0.5*x'*0*x/x[1] - log(x[1])*exp(x[2])^x[3]
                     end
                     function foo4 expansion(x)
                              # TODO: this function should output the hessian H and gradient g of the
                              # TODO: calculate the gradient of foo4 evaluated at x
                              g = FD.gradient(foo4, x);
                              # TODO: calculate the hessian of foo4 evaluated at x
                              H = FD.hessian(foo4, x);
```

```
return g, H
         end
 Out[8]: foo4_expansion (generic function with 1 method)
In [10]: @testset "1a" begin
             x = [.2; .4; .5]
             g,H = foo4\_expansion(x)
             @test isapprox(g,[-18.98201379080085, 4.982885952667278, 8.2863087621338
             @test norm(H - [164.2850689540042 -23.053506895400425 -39.942805516320334
                                       -23.053506895400425 10.491442976333639 2.358926
                                       -39.94280551632034 2.3589262864014673 15.314523
         end
         q = FD.qradient(foo4, x) = [-18.98201379080085, 4.982885952667278, 8.286308]
         H = FD.hessian(foo4, x) = [164.2850689540042 -23.053506895400425 -39.942805]
         516320334; -23.053506895400425 10.491442976333639 2.3589262864014673; -39.9
         4280551632034 2.3589262864014673 15.314523504853529]
         Test Summary: | Pass Total
                            2
Out[10]: Test.DefaultTestSet("1a", Any[], 2, false, false)
```

Part (b): Derivatives for functions with multiple input arguments (2 pts)

```
In [9]: # NOTE: this block is a tutorial, you do not have to fill anything out.
        # calculate derivatives for functions with multiple inputs
        function dynamics(x,a,b,c)
             return [x[1]*a; b*c*x[2]*x[1]]
        end
        let
            x1 = randn(2)
            a = randn()
            b = randn()
            c = randn()
            # this evaluates the jacobian with respect to x, given a, b, and c
            A1 = FD.jacobian(dx \rightarrow dynamics(dx, a, b, c), x1)
            # it doesn't matter what we call the new variable
            A2 = FD.jacobian(_x \rightarrow dynamics(_x, a, b, c), x1)
            # alternatively we can do it like this using a closure
            dynamics just x(x) = dynamics(x, a, b, c)
            A3 = FD.jacobian(dynamics_just_x, x1)
            \alphatest norm(A1 - A2) < 1e-13
            (atest norm(A1 - A3) < 1e-13)
        end
```

Out[9]: Test Passed

```
In [11]:
    function eulers(x,u,J)
        # dynamics when x is angular velocity and u is an input torque
        x = J\(u - cross(x,J*x))
        return x
end

function eulers_jacobians(x,u,J)
        # given x, u, and J, calculate the following two jacobians

# TODO: fill in the following two jacobians

# dx/dx
    A = FD.jacobian(dx -> eulers(dx, u, J), x);

# dx/du
    B = FD.jacobian(du -> eulers(x, du, J), u);

return A, B
end
```

Out[11]: eulers_jacobians (generic function with 1 method)

```
In [12]: @testset "1b" begin

x = [.2;-7;.2]
u = [.1;-.2;.343]
J = diagm([1.03;4;3.45])

A,B = eulers_jacobians(x,u,J)

skew(v) = [0 -v[3] v[2]; v[3] 0 -v[1]; -v[2] v[1] 0]
@test isapprox(A,-J\(skew(x)*J - skew(J*x)), atol = 1e-8)

@test norm(B - inv(J)) < 1e-8

end</pre>
```

```
1b | 2 2
Out[12]: Test.DefaultTestSet("1b", Any[], 2, false, false)
```

Part (c): Derivatives of composite functions (1 pts)

```
In [20]: # NOTE: this block is a tutorial, you do not have to fill anything out.
function f(x)
    return x[1]*x[2]
end
function g(x)
    return [x[1]^2; x[2]^3]
end
```

Test Summary: | Pass Total

```
x1 = 2*randn(2)
              # using gradient of the composite function
              \nabla f_1 = FD.gradient(dx \rightarrow f(g(dx)), x1)
              # using the chain rule
              J = FD.jacobian(q, x1)
              \nabla f 2 = J'*FD.gradient(f, g(x1))
              @show norm(\nabla f_1 - \nabla f_2)
          end
          norm(\nabla f_1 - \nabla f_2) = 0.0
Out[20]: 0.0
In [21]: function f2(x)
              return x*sin(x)/2
          end
          function q2(x)
               return cos(x)^2 - tan(x)^3
          end
          function composite_derivs(x)
              # TODO: return \partial y/\partial x where y = g2(f2(x))
              # (hint: this is 1D input and 1D output, so it's ForwardDiff.derivative)
              return FD.derivative(dx -> g2(f2(dx)), x)
          end
Out[21]: composite_derivs (generic function with 1 method)
In [22]: @testset "1c" begin
              x = 1.34
              deriv = composite_derivs(x)
              @test isapprox(deriv,-2.390628273373545,atol = 1e-8)
          end
          Test Summary: | Pass Total
                              1
Out[22]: Test.DefaultTestSet("1c", Any[], 1, false, false)
```

O1

Part (d): Fixing the most common ForwardDiff error (2 pt)

First we will show an example of this error:

```
In [23]: # NOTE: this block is a tutorial, you do not have to fill anything out.
function f_zero_1(x)
    println("-----types of input x----")
    @show typeof(x) # print out type of x
    @show eltype(x) # print out the element type of x
```

```
xdot = zeros(length(x)) # this default creates zeros of type Float64
    println("----types of output xdot----")
    @show typeof(xdot)
    @show eltype(xdot)
    # these lines will error because i'm trying to put a ForwardDiff.dual
    # inside of a Vector{Float64}
    xdot[1] = x[1]*x[2]
    xdot[2] = x[2]^2
    return xdot
end
let
    # try and calculate the jacobian of f_zero_1 on x1
    x1 = randn(2)
    @info "this error is expected:"
    try
        FD.jacobian(f_zero_1,x1)
    catch e
        buf = IOBuffer()
        showerror(buf,e)
        message = String(take!(buf))
        Base.showerror(stdout,e)
    end
end
[ Info: this error is expected:
```

```
-----types of input x-----
typeof(x) = Vector{ForwardDiff.Dual{ForwardDiff.Tag{typeof(f_zero_1), Float
64}, Float64, 2}}
eltype(x) = ForwardDiff.Dual{ForwardDiff.Tag{typeof(f_zero_1), Float64}, Fl
oat64, 2}
-----types of output xdot-----
typeof(xdot) = Vector{Float64}
eltype(xdot) = Float64
MethodError: no method matching Float64(::ForwardDiff.Dual{ForwardDiff.Tag
{typeof(f_zero_1), Float64}, Float64, 2})
Closest candidates are:
  (::Type{T})(::Real, ::RoundingMode) where T<:AbstractFloat at rounding.j</pre>
1:200
  (::Type{T})(::T) where T<:Number at boot.jl:760
  (::Type{T})(::AbstractChar) where T<:Union{AbstractChar, Number} at char.</pre>
jl:50
```

This is the most common ForwardDiff error that you will encounter. ForwardDiff works by pushing ForwardDiff.Dual variables through the function being differentiated. Normally this works without issue, but if you create a vector of Float64 (like you would with xdot = zeros(5), it is unable to fit the ForwardDiff.Dual 's in with the Float64 's. To get around this, you have two options:

Option 1

Our first option is just creating xdot directly, without creating an array of zeros to index into.

```
In [18]: # NOTE: this block is a tutorial, you do not have to fill anything out.
         function f_zero_1(x)
             # let's create xdot directly, without first making a vector of zeros
             xdot = [x[1]*x[2], x[2]^2]
             # NOTE: the compiler figures out which type to make xdot, so when you ca
             # it's a Float64, and when it's being diffed, it's automatically promote
             println("-----types of input x-----")
             @show typeof(x) # print out type of x
             (ashow eltype(x) # print out the element type of x)
             println("-----types of output xdot-----")
             @show typeof(xdot)
             @show eltype(xdot)
             return xdot
         end
         let
             # try and calculate the jacobian of f_zero_1 on x1
             x1 = randn(2)
             FD.jacobian(f_zero_1,x1) # this will work
         end
         -----types of input x-----
         typeof(x) = Vector{ForwardDiff.Dual{ForwardDiff.Tag{typeof(f_zero_1), Float
         64}, Float64, 2}}
         eltype(x) = ForwardDiff.Dual{ForwardDiff.Tag{typeof(f_zero_1), Float64}, Fl
         oat64, 2}
         -----types of output xdot-----
         typeof(xdot) = Vector{ForwardDiff.Dual{ForwardDiff.Tag{typeof(f_zero_1), Fl
         oat64}, Float64, 2}}
         eltype(xdot) = ForwardDiff.Dual{ForwardDiff.Tag{typeof(f zero 1), Float64},
         Float64, 2}
Out[18]: 2×2 Matrix{Float64}:
          -0.322914 -0.376712
                     -0.645828
          -0.0
```

Option 2

The second option is to create the array of zeros in a way that accounts for the input type. This can be done by replacing zeros(length(x)) with zeros(eltype(x), length(x)). The first argument eltype(x) simply creates a vector of zeros that is the same type as the element type in vector x.

```
In [24]: # NOTE: this block is a tutorial, you do not have to fill anything out.
         function f zero 1(x)
             xdot = zeros(eltype(x), length(x))
             xdot[1] = x[1]*x[2]
             xdot[2] = x[2]^2
             println("-----types of input x-----")
             @show typeof(x) # print out type of x
             Oshow eltype(x) # print out the element type of x
             println("-----types of output xdot-----")
             @show typeof(xdot)
             @show eltype(xdot)
             return xdot
         end
         let
             # try and calculate the jacobian of f zero 1 on x1
             x1 = randn(2)
             FD.jacobian(f zero 1,x1) # this will fail!
         end
         -----types of input x-----
         typeof(x) = Vector{ForwardDiff.Dual{ForwardDiff.Tag{typeof(f zero 1), Float
         64}, Float64, 2}}
         eltype(x) = ForwardDiff.Dual{ForwardDiff.Tag{typeof(f zero 1), Float64}, Fl
         oat64, 2}
         ----types of output xdot----
         typeof(xdot) = Vector{ForwardDiff.Dual{ForwardDiff.Tag{typeof(f zero 1), Fl
         oat64}, Float64, 2}}
         eltype(xdot) = ForwardDiff.Dual{ForwardDiff.Tag{typeof(f_zero_1), Float64},
         Float64, 2}
Out[24]: 2×2 Matrix{Float64}:
          -1.30403 -1.0754
                    -2.60807
          -0.0
         Now you can show that you understand these two options by fixing two broken
         functions.
In [67]: # TODO: fix this error when trying to diff through this function
         # hint: you can use promote type(eltype(x), eltype(u)) to return the correct
         function dynamics(x,u)
             \dot{x} = zeros(promote_type(eltype(x), eltype(u)), length(x)) # use correct t
             \dot{x}[1] = x[1]*\sin(u[1])
             \dot{x}[2] = x[2]*cos(u[2])
             return \dot{x}
         end
```

Out[67]: dynamics (generic function with 1 method)

Finite Difference Derivatives

If you ever have trouble working through a ForwardDiff error, you should always feel free to use the FiniteDiff.jl FiniteDiff.jl package instead. This computes derivatives through a finite difference method. This is slower and less accurate than ForwardDiff, but it will always work so long as the function works.

Before with ForwardDiff we had this:

- FD.derivative(f,x) derivative of scalar or vector valued f wrt scalar x
- FD.jacobian(f,x) jacobian of vector valued f wrt vector x
- FD.gradient(f,x) gradient of scalar valued f wrt vector x
- FD.hessian(f,x) hessian of scalar valued f wrt vector x

Now with FiniteDiff we have this:

- FD2.finite_difference_derivative(f,x) derivative of scalar or vector valued f wrt scalar x
- FD2.finite_difference_jacobian(f,x) jacobian of vector valued f wrt vector x
- FD2.finite_difference_gradient(f,x) gradient of scalar valued f wrt vector x
- FD2.finite_difference_hessian(f,x) hessian of scalar valued f wrt vector x

```
In [31]: # NOTE: this block is a tutorial, you do not have to fill anything out.

# load the package
import FiniteDiff as FD2

function foo1(x)
    #scalar input, scalar output
    return sin(x)*cos(x)^2
end

function foo2(x)
    # vector input, scalar output
```

```
return sin(x[1]) + cos(x[2])
          end
          function foo3(x)
               # vector input, vector output
               return [\sin(x[1])*x[2];\cos(x[2])*x[1]]
          end
          let # we just use this to avoid creating global variables
               # evaluate the derivative of fool at x1
               x1 = 5*randn();
               @show \partial foo1 \partialx = FD2.finite difference derivative(foo1, x1);
               # evaluate the gradient and hessian of foo2 at x2
               x2 = 5*randn(2);
               @show \nablafoo2 = FD2.finite_difference_gradient(foo2, x2);
               @show \nabla^2 foo2 = FD2.finite_difference_hessian(foo2, x2);
               # evluate the jacobian of foo3 at x2
               @show \partial foo3_{\partial x} = FD2_{finite_difference_jacobian(foo3,x2)};
               @test norm(\partialfoo1_\partialx - FD.derivative(foo1, x1)) < 1e-4
               @test norm(\nablafoo2 - FD.gradient(foo2, x2)) < 1e−4
               @test norm(\nabla^2foo2 - FD.hessian(foo2, x2)) < 1e-4
               @test norm(\partialfoo3 \partialx - FD.jacobian(foo3, x2)) < 1e-4
          end
          \partial foo1 \partial x = FD2.finite difference derivative(foo1, x1) = 0.3341157713242085
          \nablafoo2 = FD2.finite_difference_gradient(foo2, x2) = [0.885051153162438, 0.08
          516940844437484]
          \nabla^2 foo2 = FD2.finite difference hessian(foo2, x2) = [-0.4654937833547592 0.
          0; 0.0 -0.996366485953331]
          \partial foo3_{\partial x} = FD2_{finite\_difference\_jacobian(foo3, x2)} = [-0.07547071296721697]
          0.4654937768355012; 0.996366485953331 0.04123838245868683]
Out[31]: Test Passed
```

In []: