American Marten Habitat Suitability

Goal

To identify if Pennsylvania suitable habitat areas for American marten.

Description

Suitable habitat areas will be determined using multiple natural resource rasters. Each raster will be given Habitat Suitability Indices (HSI) based off Tom Keller and other expert opinions. A sum of the rasters will provide areas of high suitability on the landscape. Using suitable areas, site visits will be conducted for field verification. If it is determined that suitability areas exist, reintroduction of the species could be discussed as an option.

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image credit: https://i2.wp.com/24.media.tumblr.com/tumblr_ma7nv7qdtg1rzpxtno1_400.jpg

Table of Contents

I. Data

- Data Source Links
- Study Area

II. Data Prep

- Merging Data
- Average Snow Accumulation



III. Defining HSI Values

- What is Habitat Suitability Index (HSI)?
- HSI Values for American marten parameters

IV. Analyses

- Analysis
- Analyses
- Truthing the Model Against Known Marten Populations

V. Documentation References

I. Data

Data Source Links

1. Land Cover

- NLCD Land Cover will be used as our land cover data
 - 2019 CONUS NLCD was used

2. Snow Cover

The National Weather Service, National Operational Hydrologic Remote Sensing Center, National Snowfall Analysis was used

- NWS National Snowfall Analysis
 - Downloaded yearly snowfall data from 2009 2019 then calculated an average over those years

3. Percent Canopy Cover

Land Fire's "Existing Vegetation Cover" was used for Percent Canopy Cover

- Land Fire Data Download will be used for percent canopy cover
 - LF 2020us_210 Existing Vegetation Cover was used
- Existing Vegetation Cover Description

4. Stand Age/ Tree Height

Land Fire's "Existing Vegetation Height" was used for Percent Canopy Cover

- Land Fire Data Download will be used for percent canopy cover
 - LF 2020us_210 Existing Vegetation Height was used
 - Vegetation Height was used as a metric for stand age
- Existing Vegetation Height Description

Study Area

AmMarten_HSI_FinalAnalysis_PDF

- What is the study area?
 - Pennsylvania, Maine, New Hampshire, Vermont, New York, and Michigan
 - States other than Pennsylvania were used to compaire existing marten populations to areas of high suitability in Pennslyvania



II. Data Prep

Merging Data

In [5]:**import** os

```
# In this step, I am setting up my folders and work spaces. By naming the folder
locations and only using the name in the code below,
# I can use this code in other projects and link to new datasets by changing the
locations in 1 spot in the code.
#Data Location - Where data used in the project was downloaded to
data_location = r"P:\Projects\BWM_AmericanMartenHSI\Data"
#Geodatabase Location - Where any feature classes for the project are saved
gdb = r"P:\Projects\BWM_AmericanMartenHSI\AmMarten_HSI_Python.gdb"
#Output Location - Where any outputs from this code are stored
out_location = r"P:\Projects\BWM_AmericanMartenHSI\Output"
print("os imported")
print("Folders Assigned")
os imported
```

```
AmMarten_HSI_FinalAnalysis_PDF
```

```
Folders Assigned
In [1]:#List of all Rasters included in the Analysis
   #Set Environment Workspace to data_location
   arcpy.env.workspace = data location
   # Creating lists of each dataset allows me to call these lists later in code without
   listing each layer one by one. The data was
   # downloaded by state, so it will need to be combined to one raster in code below.
   #List of all National Land Cover data
   nlcd list = arcpy.ListRasters("NLCD 2019*")
   #Using "NLCD 2019*" looks for data in the workspace listed above that starts with
    "NLCD 2019". Place the * on either side or both sides
   # depending on what text is the same in each raster
   #List of all Land fire Canopy Cover data
   cover list = arcpy.ListRasters("* evc")
   #List of all Land Fire Canopy Height data
   height list = arcpy.ListRasters("* evh")
\ln [2]:#By using the list Statements above, I am able to create Raster Mosaics to combine all
    of the rasters by state into one layer
    #Merge the National Land Cover data to one raster layer
    am nlcd = arcpy.MosaicToNewRaster management(nlcd list, out location,
                                                 "am nlcd.tif", "", "8 BIT UNSIGNED", "30",
    "1", "LAST", "FIRST")
    #This print statement shows the path of where the layer is saved
    #print(am nlcd)
    #Merge the Percent Canopy Cover data to one raster layer
    am cover = arcpy.MosaicToNewRaster management(cover list, out location,
                                                  "am cover.tif", "", "8 BIT UNSIGNED", "30",
    "1", "LAST", "FIRST")
    #print(am cover)
    #Merge the Tree Height data to one raster layer
    am height = arcpy.MosaicToNewRaster management(height list, out location,
                                                  "am height.tif", "", "8 BIT UNSIGNED",
    "30", "1", "LAST", "FIRST")
    #print(am height)
   print("Mosaics are created")
Mosaics are created
```

Average Snow Accumulation

- We downloaded a 10 year window of snow accumulation layers, but an average is needed
 - Average all of the snow accumulation layers using Cell Statistics

```
In [3]:import arcpy
```

AmMarten_HSI_FinalAnalysis_PDF

```
import os
#Set Environment Workspace to data_location
arcpy.env.workspace = data_location
#List of all Snow data
snow_list = arcpy.ListRasters("sfav2_CONUS_*")
#print(snow_list)
#Average the snow tifs from the list above
snow_avg = arcpy.sa.CellStatistics(snow_list, "MEAN")
print("Snow Average Complete")
Snow Average Complete
```

III. Defining HSI Values

What is Habitat Suitability Index (HSI)?

- HSI gives variables a numerical index
 - The numerical index is used to represent the ability for that variable to support a selected species
 - An example of the methods used to assign HSI can be seen below



Figure 1. Steps for constructing the HSI.

- An example of an HSI can be seen below
 - Typically the numberical values are decimals from 0 to 1.0
 - For our analysis, I assigned values from 0 to 100 for our HSI values





HSI Image Source: https://archive.epa.gov/aed/html/research/scallop/web/html/hsi.html

- Examples of HSI Models
 - American Marten HSI some parameters from this paper were used in our analysis
 - Understanding HSI More options for HSI analyses
 - Water Management HSI Example Another example of HSI used in South Florida

HSI Values for American marten parameters

Land Cover HSI

Attribute	HSI Value
All other Land Classes	NO DATA
Shrub	25
Deciduous	50
Mixed	75
Coniferous	100

Snow Cover HSI

Attribute	HSI Value
0 cm	NO DATA
10 - 34 cm	25
35 - 59 cm	50
60 - 90 cm	75
> 91cm	100

Percent Canopy Cover HSI

Attribute	HSI Value
0 - 25%	NO DATA
N/A	25
26 - 50%	50
N/A	75
> 50%	100

Tree Height HSI

Attribute	HSI Value
< 1m	NO DATA
1 - 7m	25
8 - 13m	50
14 - 20m	75
> 20m	100

IV. Analyses

Analysis

In analysis 2, coarse woody debris was taken out before the summation of the layers. The layer may not be accurate to what is present in the field, so an analysis was run, leaving the layer out.



Analyses

```
In [6]:import arcpy
```

```
# Set workspace and Environments
# This is important to make sure all layers are aligned
gdb = r"P:\Projects\BWM_AmericanMartenHSI\AmMarten_HSI_Python.gdb"
arcpy.env.workspace = gdb
arcpy.env.outputCoordinateSystem = am_nlcd[0]
arcpy.env.mask = am_nlcd[0]
print(arcpy.env.outputCoordinateSystem.name)
#Set up link to study area
study_states = os.path.join(gdb, "BufferedStudyArea")
# Extract by mask used to clip all rasters to the study states and sets the environments
```

```
AmMarten_HSI_FinalAnalysis_PDF
```

```
to align all rasters
    #Snow Extract to Study States
    am snow states = arcpy.sa.ExtractByMask(snow avg, study states)
    # am snow states.save(os.path.join(out location, "am snow states.tif"))
    # Using "os.path.join" allows you to save the output to a specified location in Step 1
    above. You will not need to change these for
    # future projects then
    #NLCD Extract to Study States
    am nlcd states = arcpy.sa.ExtractByMask(am nlcd, study states)
    # am nlcd states.save(os.path.join(out location, "am nlcd states.tif"))
    #Height Extract to Study States
    am height states = arcpy.sa.ExtractByMask(am height, study states)
    # am height states.save(os.path.join(out location, "am height states.tif"))
    #Canopy Cover Extract to Study States
    am cover states = arcpy.sa.ExtractByMask(am cover, study states)
    # am cover states.save(os.path.join(out location, "am cover states.tif"))
   print("Data Extracted to Study States")
Albers_Conical_Equal_Area
Data Extracted to Study States
In [7]:#Reclassify Layers to attach HSI Values to each raster cell. See tables above for list
    of hsi values.
    # Coarse Woody Debris Reclassify
    am cwd hsi rr = arcpy.sa.RemapRange([[91, 182, "NODATA"],
                                          [183, 189, 100],
                                          [201, 204, 100]])
    am cwd hsi = arcpy.sa.Reclassify(am cwd states, "VALUE", am cwd hsi rr)
    #National Land Cover Reclassify
    am nlcd hsi rr = arcpy.sa.RemapRange([[-1, 31, "NODATA"],
                                          [41, 41, 50],
                                          [42, 42, 100],
                                          [43, 43, 75],
                                          [51, 51, "NODATA"],
                                          [52, 52, 25],
                                          [71, 95, "NODATA"]])
    am nlcd hsi = arcpy.sa.Reclassify(am nlcd states, "VALUE", am nlcd hsi rr)
    #Adding in areas that are mixed as a value of 100 as well
    ## Redo Reclassify values from here down
    #Tree Height Reclassify
    am height hsi rr = arcpy.sa.RemapRange([[-1, 100, "NODATA"],
                                            [101, 107, 25],
```

```
[108, 113, 50],
                                           [114, 120, 75],
                                            [121, 199, 100],
                                            [201, 310, "NODATA"]])
    am height hsi = arcpy.sa.Reclassify(am height states, "VALUE", am height hsi rr)
    #Percent Canopy Cover Reclassify
    am cover hsi rr = arcpy.sa.RemapRange([[-1, 125, "NODATA"],
                                           [126, 150, 50],
                                           [151, 199, 100],
                                            [201, 399, "NODATA"]])
    am cover hsi = arcpy.sa.Reclassify(am cover states, "VALUE", am cover hsi rr)
    #Snow Average Reclassify failed - does not have valid statistics
    am snow hsi rr = arcpy.sa.RemapRange([[-100000, 0.999999, "NODATA"],
                                         [3.93, 13.78, 25],
                                         [13.78, 23.62, 50],
                                         [23.62, 36, 75],
                                          [36, 280, 100]])
    am snow hsi = arcpy.sa.Reclassify(am snow states, "VALUE", am snow hsi rr)
   print("HSI Values Assigned")
HSI Values Assigned
In [4]:# Set up datalists for the analysis
    # Setting up the environments as well
    gdb = r"P:\Projects\BWM AmericanMartenHSI\AmMarten HSI Python.gdb"
    arcpy.env.workspace = gdb
   arcpy.env.outputCoordinateSystem = am nlcd[0]
   arcpy.env.mask = am nlcd[0]
   arcpy.env.cellSize = am nlcd[0]
    # Below are the two options for running the final analysis. We chose to omit the coarse
    woody debris, cwd, data
    # in the final analysis because it did not seem to give us the results we wanted.
    # Option 1 - Including Coarse Woody Debris
    #am hs cwd data = [am cwd hsi, am nlcd hsi, am height hsi, am snow hsi, am cover hsi]
    # Analysis - Coarse Woody Debris, Land Cover, Tree Height, % Canopy Cover, Snow added
    together
    #am hs cwd = arcpy.sa.CellStatistics(am hs cwd data, "SUM")
    #am hs cwd.save(os.path.join(out location, "am hs cwd.tif"))
    #Option 2 - Removing Coarse Woody Debris. This is the option we chose.
    am hs no cwd data = [am nlcd hsi, am height hsi, am snow hsi, am cover hsi]
    # Analysis - Land Cover, Tree Height, % Canopy Cover, Snow added together
```

AmMarten_HSI_FinalAnalysis_PDF

```
am hs no cwd = arcpy.sa.CellStatistics(am hs no cwd data, "SUM")
   am hs no cwd.save(os.path.join(out location, "am hs no cwd.tif"))
   print("Analysis Complete")
Analysis Complete
In [13]:#Adding focal moving window analysis
    gdb = r"P:\Projects\BWM AmericanMartenHSI\AmMarten HSI Python.gdb"
    arcpy.env.workspace = gdb
    arcpy.env.outputCoordinateSystem = am nlcd[0]
    arcpy.env.mask = am nlcd[0]
    arcpy.env.cellSize = am nlcd[0]
    #Defining where each state's buffer is located for the extract by mask
    pa state = os.path.join(gdb, "Buffer10mi PA State")
    mi state = os.path.join(gdb, "Buffer10mi MI State")
    me state = os.path.join(gdb, "Buffer10mi ME State")
    nh state = os.path.join(gdb, "Buffer10mi NH State")
    vt state = os.path.join(gdb, "Buffer10mi VT State")
    ny state = os.path.join(gdb, "Buffer10mi NY State")
    am hs no cwd final = (os.path.join(out location, "am hs no cwd.tif"))
    #Extracting by mask the HSI tif for each state to run the focal statistic by state.
    There is not enough room to run the entire HSI
    # output in one round
    am hs no cwd pa = arcpy.sa.ExtractByMask(am hs no cwd final, pa state)
    am hs no cwd mi = arcpy.sa.ExtractByMask(am hs no cwd final, mi state)
    am hs no cwd me = arcpy.sa.ExtractByMask(am hs no cwd final, me state)
    am hs no cwd nh = arcpy.sa.ExtractByMask(am hs no cwd final, nh state)
    am hs no cwd vt = arcpy.sa.ExtractByMask(am hs no cwd final, vt state)
    am hs no cwd ny = arcpy.sa.ExtractByMask(am_hs_no_cwd_final, ny_state)
    # Running Focal Statistics for each State and saving the tif to my output location
    am hs no cwd pa fs norc = arcpy.sa.FocalStatistics(am hs no cwd pa, "Circle 54 CELL",
    "MEAN", "NODATA")
    am hs no cwd pa fs norc.save(os.path.join(out location, "am hs no cwd pa fs norc.tif"))
    am hs no cwd mi fs norc = arcpy.sa.FocalStatistics(am hs no cwd mi, "Circle 54 CELL",
    "MEAN", "NODATA")
    am hs no cwd mi fs norc.save(os.path.join(out location, "am hs no cwd mi fs norc.tif"))
    am hs no cwd me fs norc = arcpy.sa.FocalStatistics(am hs no cwd me, "Circle 54 CELL",
    "MEAN", "NODATA")
    am hs no cwd me fs norc.save(os.path.join(out location, "am hs no cwd me fs norc.tif"))
    am hs no cwd nh fs norc = arcpy.sa.FocalStatistics(am hs no cwd nh, "Circle 54 CELL",
    "MEAN", "NODATA")
    am hs no cwd nh fs norc.save(os.path.join(out location, "am hs no cwd nh fs norc.tif"))
```

```
am_hs_no_cwd_vt_fs_norc = arcpy.sa.FocalStatistics(am_hs_no_cwd_vt, "Circle 54 CELL",
"MEAN", "NODATA")
am_hs_no_cwd_vt_fs_norc.save(os.path.join(out_location, "am_hs_no_cwd_vt_fs_norc.tif"))
am_hs_no_cwd_ny_fs_norc = arcpy.sa.FocalStatistics(am_hs_no_cwd_ny, "Circle 54 CELL",
"MEAN", "NODATA")
am_hs_no_cwd_ny_fs_norc.save(os.path.join(out_location, "am_hs_no_cwd_ny_fs_norc.tif"))
```

Truthing the Model Against Known Marten Populations

In [7]:import os

```
# Setting Envrionments and Workspace for tools below
gdb = r"P:\Projects\BWM AmericanMartenHSI\AmMarten HSI Python.gdb"
arcpy.env.workspace = gdb
arcpy.env.outputCoordinateSystem = am nlcd[0]
arcpy.env.mask = am nlcd[0]
arcpy.env.cellSize = am nlcd[0]
#iNaturalist Point Analysis
#Converting iNaturalist points to rasters so that I can exctract the HSI value for each
raster cell where a sighting occured.
arcpy.conversion.PointToRaster("MartenObservationsINat", "id", os.path.join(gdb,
"MartenObservationsINat Raster"), "MOST FREQUENT", "NONE", 30, "BUILD")
#Extracting HSI values for each known marten sighting from iNaturalist
me inat hsi =
arcpy.sa.ExtractByMask(os.path.join(out location, "am hs no cwd me fs norc.tif"),
"MartenObservationsINat Raster")
me inat hsi.save(os.path.join(out location, "me inat hsi.tif"))
mi inat hsi = arcpy.sa.ExtractByMask(os.path.join(out location,
"am hs no cwd mi fs norc.tif"), "MartenObservationsINat Raster")
mi inat hsi.save(os.path.join(out location, "mi inat hsi.tif"))
nh inat hsi = arcpy.sa.ExtractByMask(os.path.join(out location,
"am_hs_no_cwd_nh_fs_norc.tif"), "MartenObservationsINat Raster")
nh inat hsi.save(os.path.join(out location, "nh inat hsi.tif"))
vt inat hsi = arcpy.sa.ExtractByMask(os.path.join(out location,
"am hs no cwd vt fs norc.tif"), "MartenObservationsINat Raster")
vt inat hsi.save(os.path.join(out location, "vt inat hsi.tif"))
ny inat hsi = arcpy.sa.ExtractByMask(os.path.join(out location,
"am_hs_no_cwd_ny_fs_norc.tif"), "MartenObservationsINat_Raster")
ny inat hsi.save(os.path.join(out location, "ny inat hsi.tif"))
#Converting the Rasters back to points to be able to run statistics on the feature
layer. This allowed me to average all of the HSI
```

values from each sighting point together

arcpy.conversion.RasterToPoint("me_inat_hsi", os.path.join(gdb, "me_inat_rasterpoint", "Value")

arcpy.conversion.RasterToPoint("mi_inat_hsi", os.path.join(gdb,"mi_inat_rasterpoint",
"Value")

arcpy.conversion.RasterToPoint("nh_inat_hsi", os.path.join(gdb,"nh_inat_rasterpoint", "Value")

arcpy.conversion.RasterToPoint("vt_inat_hsi", os.path.join(gdb,"vt_inat_rasterpoint",
"Value")

arcpy.conversion.RasterToPoint("ny_inat_hsi", os.path.join(gdb,"ny_inat_rasterpoint",
"Value")

Michigan Collared Marten Home Ranges

#Extracted the HSI layer by each of the marten home ranges

out_raster = arcpy.sa.ExtractByMask("am_hs_no_cwd.tif", "All_MI_MartenPops95")
#all mi marten95 hsi.save(os.path.join(out location, "all mi marten95 hsi")

Converted each of the rasters within each marten home range to individual points arcpy.conversion.RasterToPoint("all_mi_marten95_hsi", os.path.join(out_location, "all_mi_marten95"), "Value")

Summarized all of the points within each of the marten home ranges to get the average HSI within each known marten home range arcpy.analysis.SummarizeWithin("All_MI_MartenPops95", "all_mi_marten95 avg HSI: 329.4", os.path.join(out_location, "All_MI_MartenPops95_SummarizeWithin", "KEEP_ALL", "grid_code Mean;grid_code Max;grid_code Min", "ADD_SHAPE_SUM", '', None, "NO_MIN_MAJ", "NO PERCENT", None)

#Converted each of the summary polygons to a point so that I could add each homerange to the point layer to average all known marten HSIs together arcpy.management.FeatureToPoint("All_MI_MartenPops95_SummarizeWithin 336", os.path.join(gdb, "All_MI_MartenPops95_SumPts", "INSIDE")

V. Documentation References

Cell Statistics

https://pro.arcgis.com/en/pro-app/latest/tool-reference/spatial-analyst/cell-statistics.htm

Environment Settings

 https://pro.arcgis.com/en/pro-app/2.8/tool-reference/environment-settings/an-overview-of-geoprocessingenvironment-settings.htm

Euclidean Distance

https://pro.arcgis.com/en/pro-app/latest/tool-reference/spatial-analyst/euclidean-distance.htm

Extract by Mask

• https://pro.arcgis.com/en/pro-app/latest/tool-reference/spatial-analyst/extract-by-mask.htm

Focal Statistics

https://pro.arcgis.com/en/pro-app/latest/tool-reference/spatial-analyst/focal-statistics.htm

ListRasters

• https://pro.arcgis.com/en/pro-app/2.8/arcpy/functions/listrasters.htm

Mosaic to New Raster

• https://pro.arcgis.com/en/pro-app/latest/tool-reference/data-management/mosaic-to-new-raster.htm

Reclassify

https://pro.arcgis.com/en/pro-app/latest/tool-reference/spatial-analyst/reclassify.htm

Remap Range

• https://pro.arcgis.com/en/pro-app/latest/arcpy/spatial-analyst/an-overview-of-transformation-classes.htm