

Data Visualization:

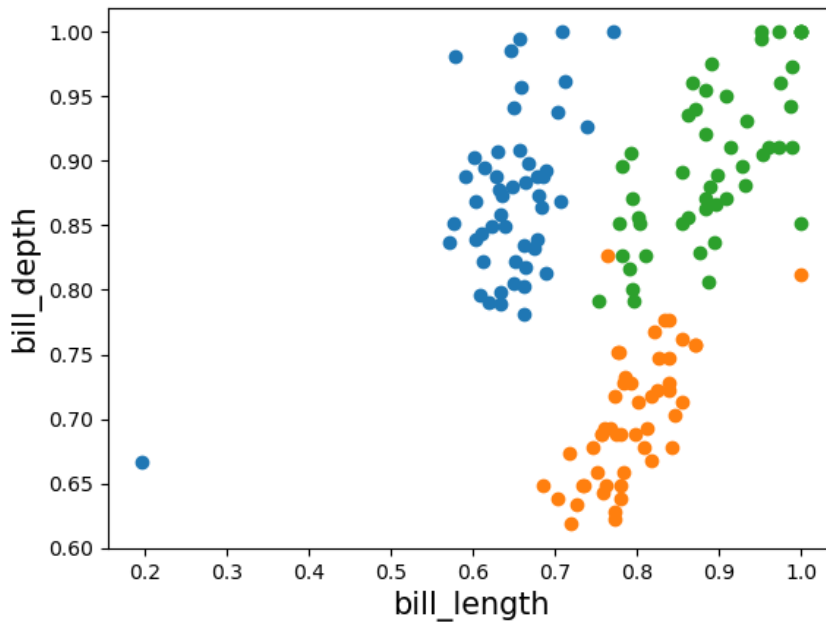
First visualize between all Penguins species by:

Blue: Adelie

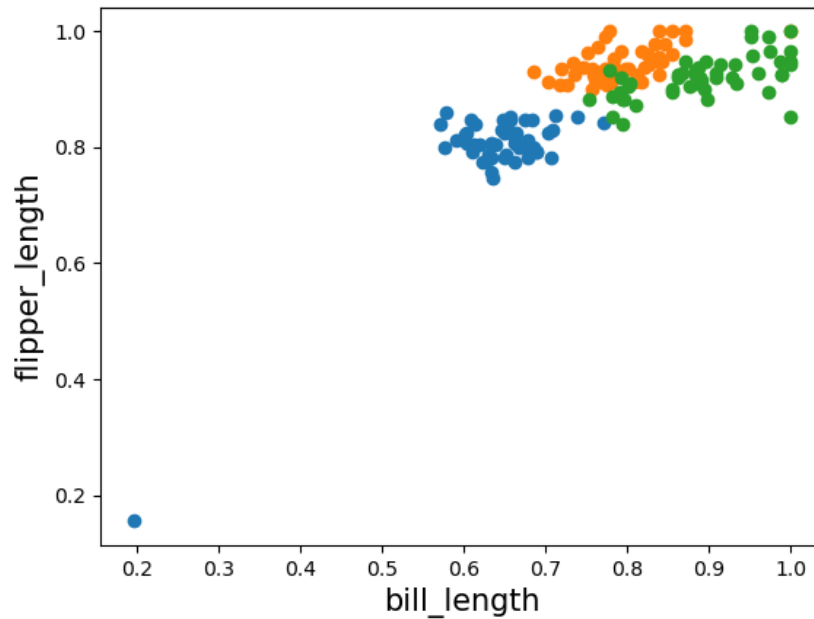
Orange: Gentoo

Green: Chinstrap

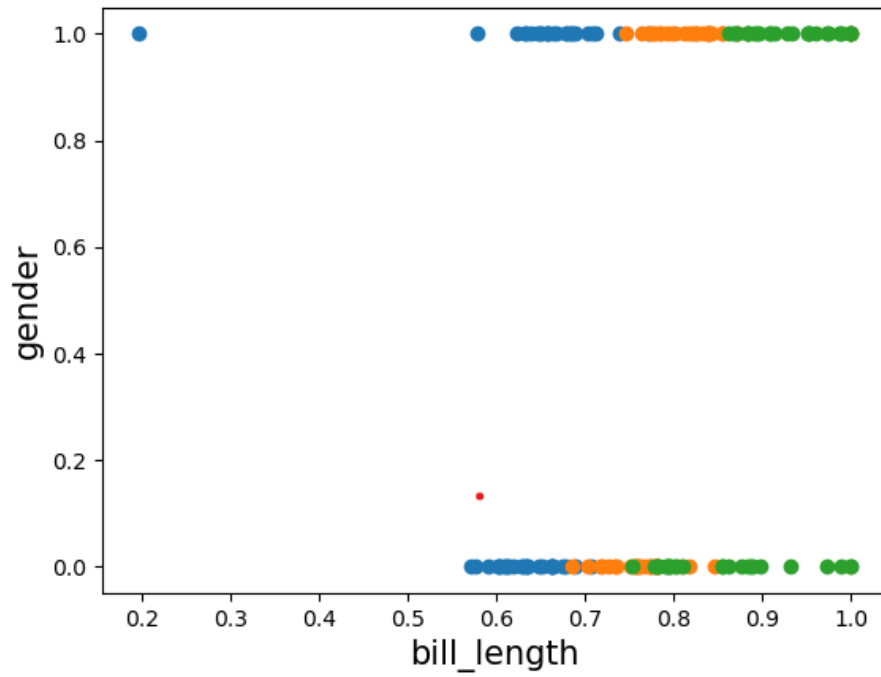
- Bill_length and bill_depth



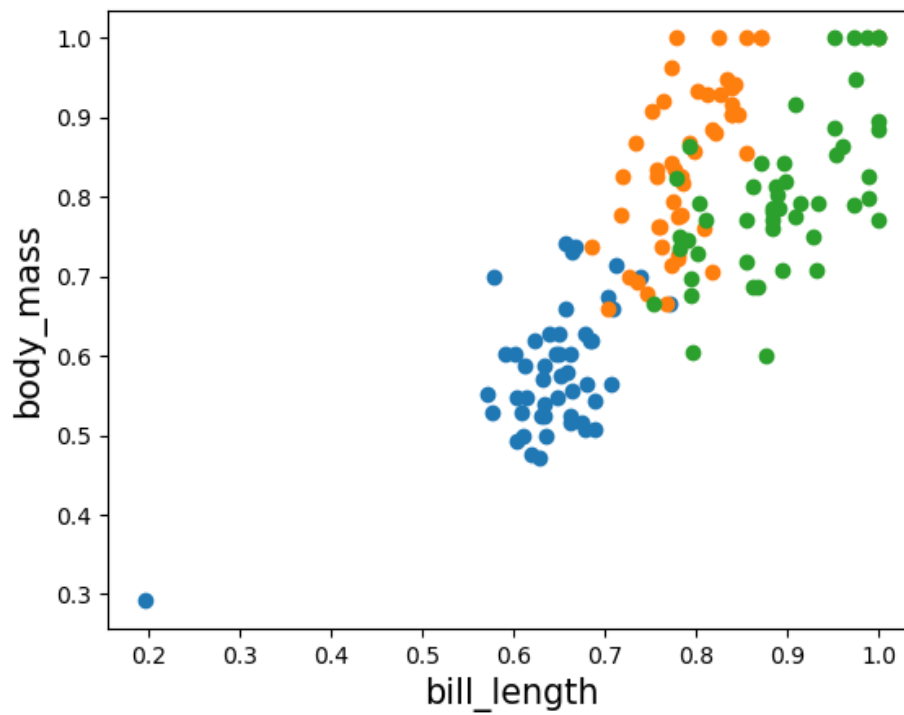
- Bill_length and flipper_length



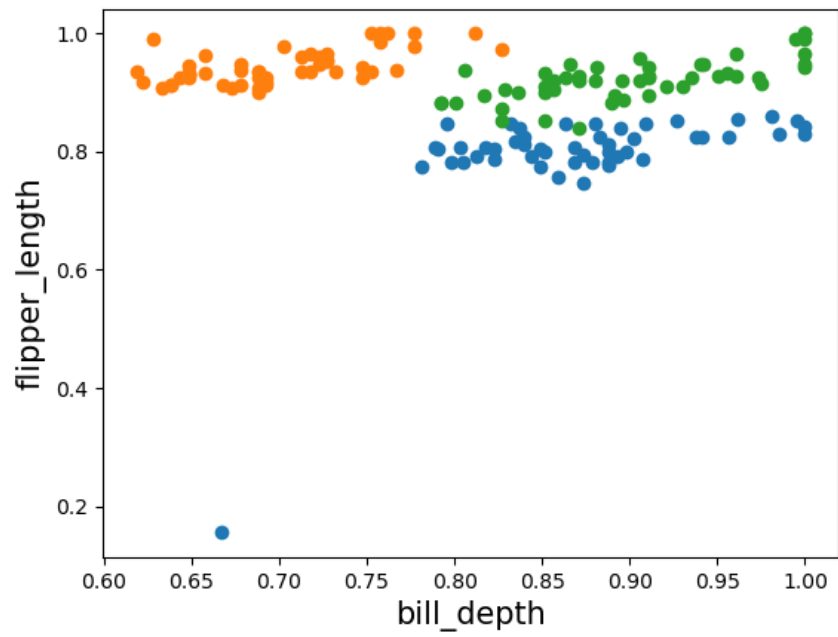
- Bill_length and gender



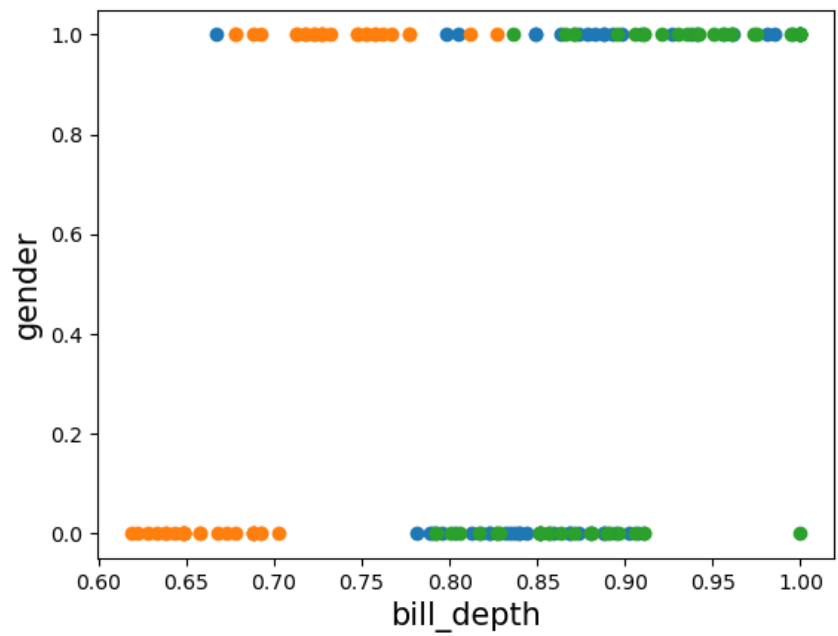
- Bill_length and body_mass



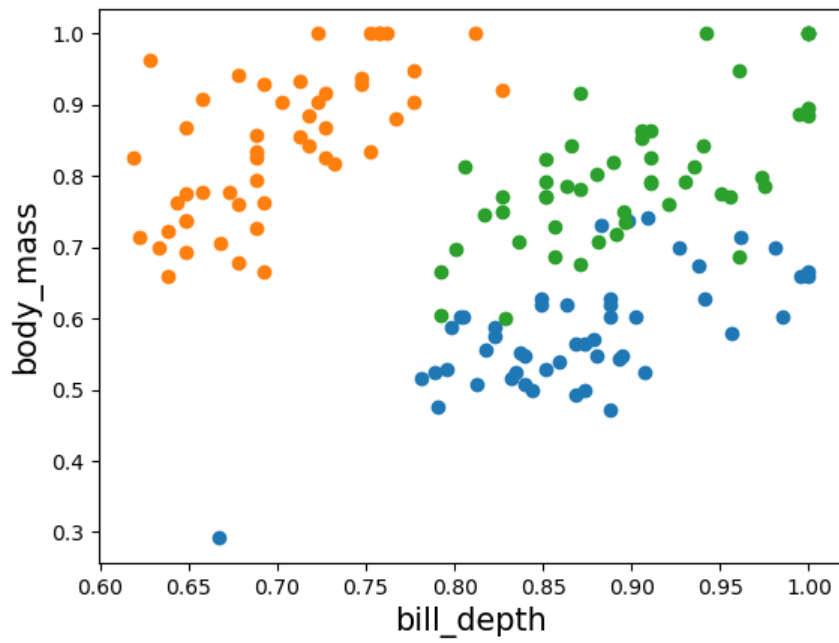
- Bill_depth and flipper_length



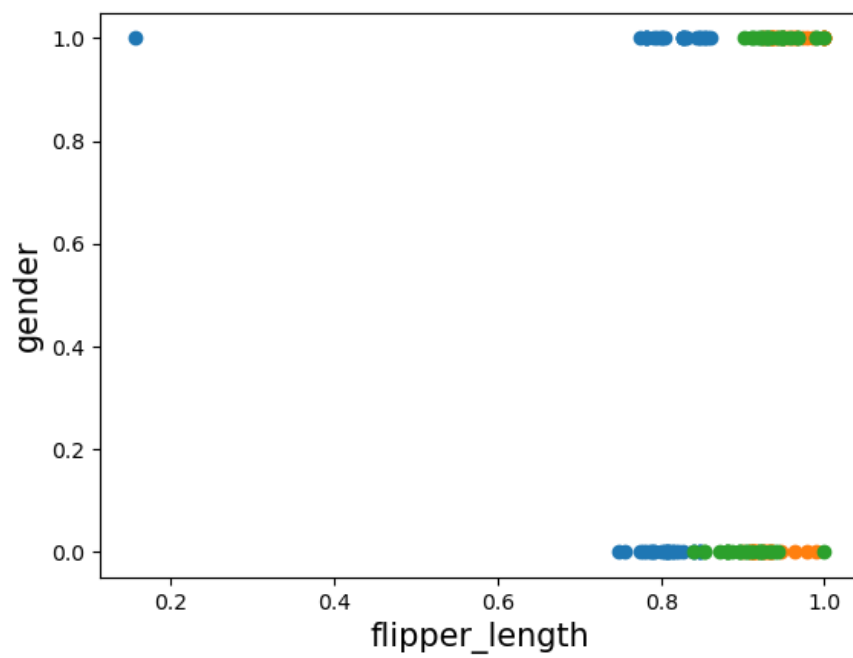
- Bill_depth and gender



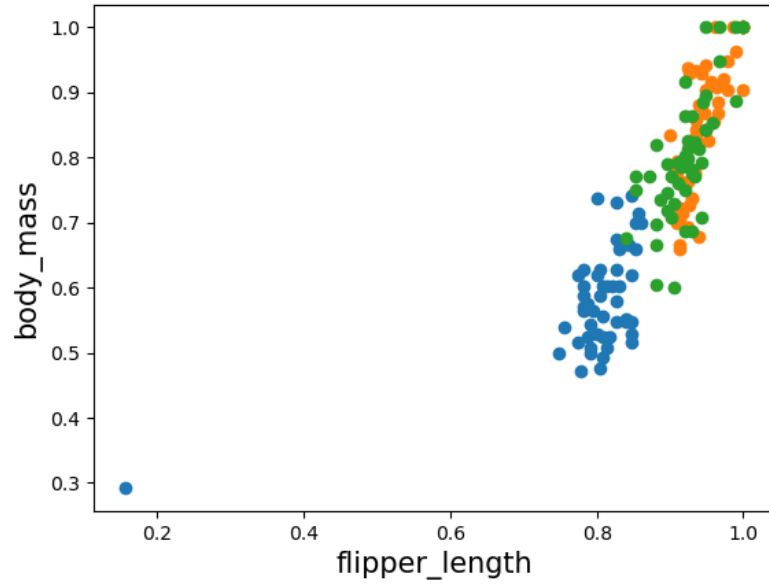
- Bill_depth and body_mass



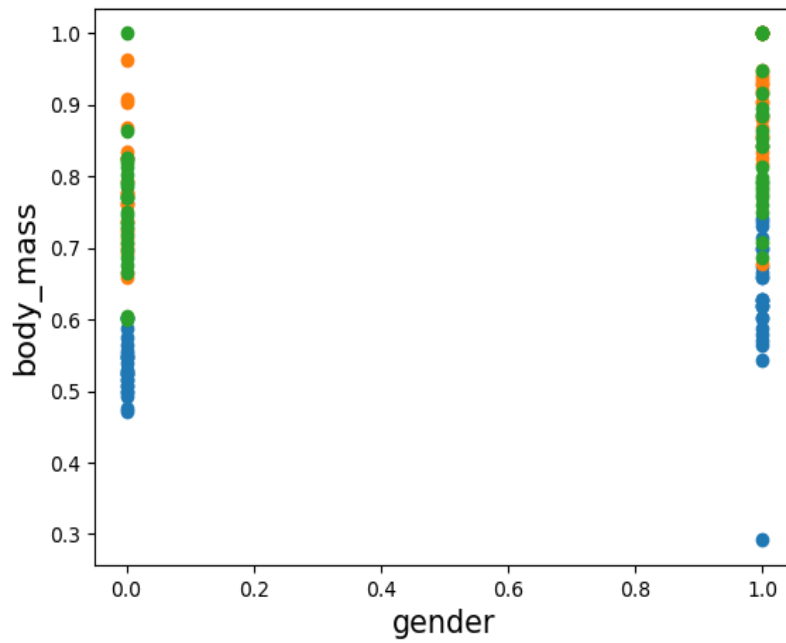
- flipper_length and gender



- flipper_length and body_mass



- gender and body_mass



- when using Bill_length and bill_depth
we find that the 3 classes are linearly separable
- when using Bill_length and flipper_length
we find that we can separate between class Adelie and class Chinstrap with line and we can separate between class Adelie and class Gentoo with line
- when using Bill_length and gender
we can not separate 3 classes with each other
- when using Bill_length and body_mass
we can separate class Adelie and class Chinstrap
- when use Bill_depth and flipper_length
we can separate 3 classes
- when using Bill_depth and gender
we can not separate 3 classes because they are not linearly separable
- when using Bill_depth and body_mass
we can separate 3 classes because they are linearly separable
- when using flipper_length and gender
we can hard separate between 3 classes because they are not linearly separable
- when using flipper_length and body_mass
we can separate class Adelie and class Chinstrap only
- when using gender and body_mass
we can not separate 3 classes because they are not linearly separable

Preprocessing techniques:

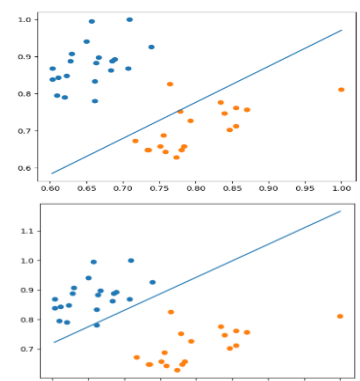
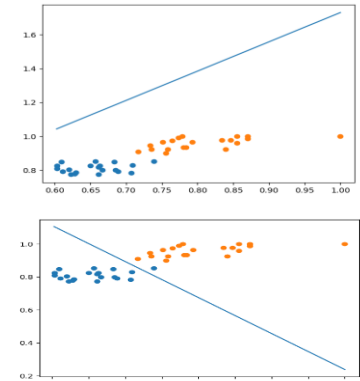
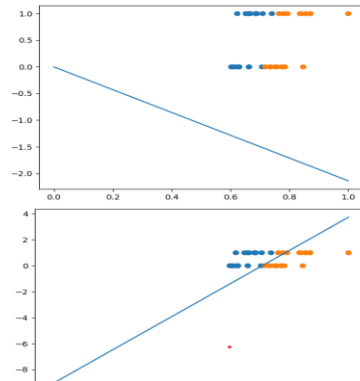
- Reading the features from the dataset and filling the NA.
- features in gender we used the normalize equation to
normalize the data from 0 to 1.
- converted the three classes into 3 numbers from (1 to 3)
to make it easier in the code.

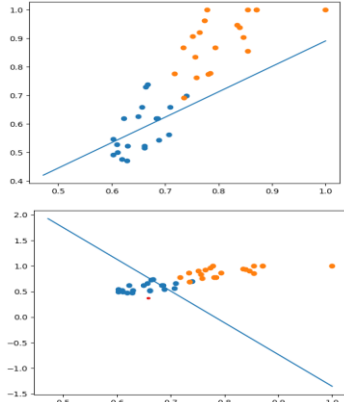
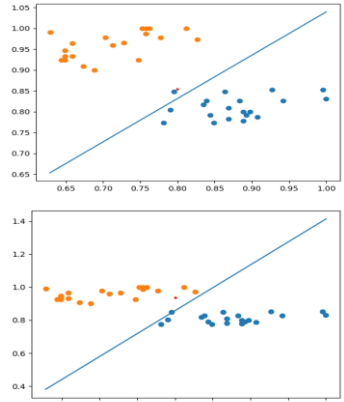
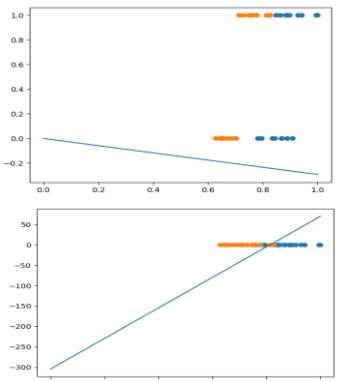
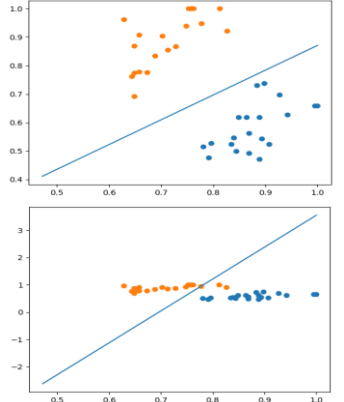
NOTE:

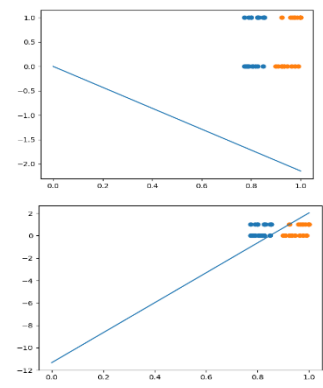
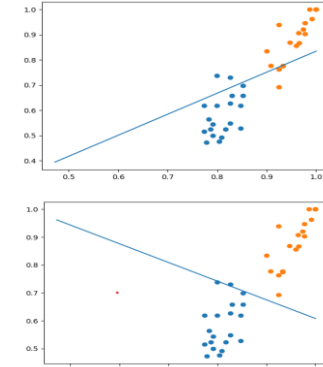
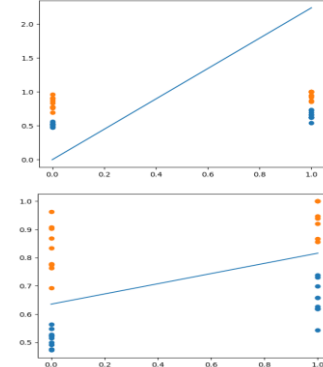
For all next result make epoch = 2000 and Learning Rate = 0.01

Visualization between each two Penguins:

“Adelie” and “Gentoo”:

Features	Accuracy	plot
bill_length and bill_depth:	Without bias:97.5% With bias: 97.5%	
bill_length and flipper length	Without bias:50.0% With bias: 97.5%	
bill_length and gender	Without bias:50.0% With bias:95.0%	

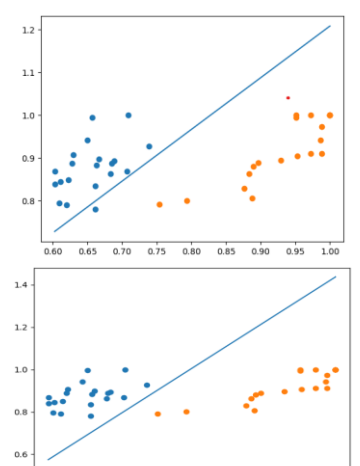
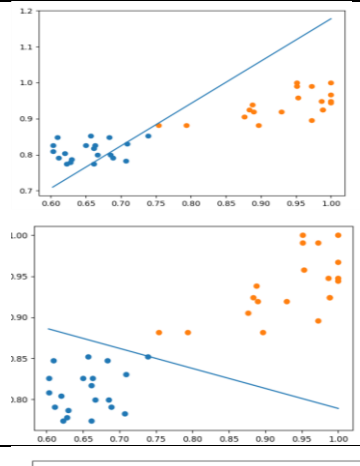
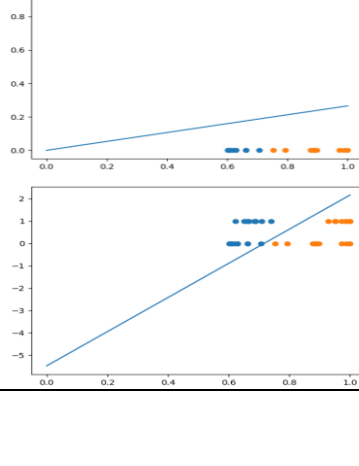
<p>bill_length and body_mass</p>	<p>Without bias: 75.0%</p> <p>With bias:92.5%</p>	
<p>Bill_dipth and flipper_length</p>	<p>Without bias: 97.5%</p> <p>With bias:97.5%</p>	
<p>Bill_dipth and gender</p>	<p>Without bias:50.0%</p> <p>With bias: 90.0%</p>	
<p>Bill_dipth and body_mass</p>	<p>Without bias:100.0%</p> <p>With bias: 95.0%</p>	

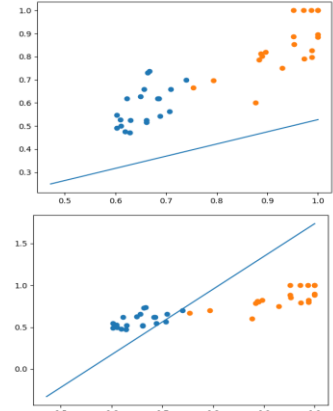
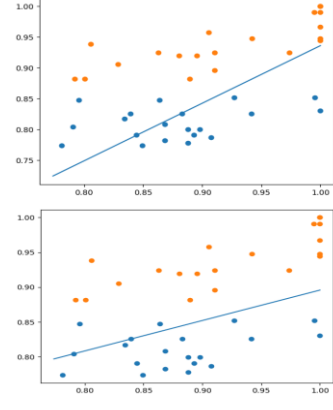
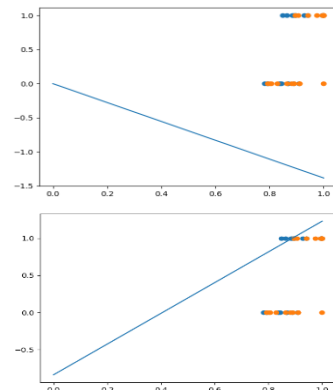
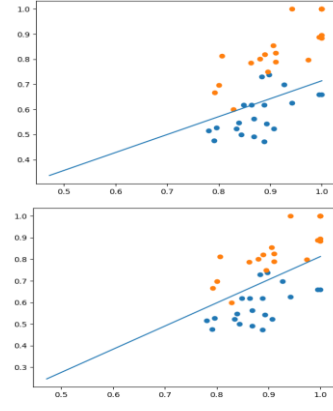
flipper_length and gender	<p>Without bias:50.0%</p> <p>With bias: 100.0%</p>	
flipper_length and body_mass	<p>Without bias:85.0%</p> <p>With bias: 97.5%</p>	
Gender and body_mass	<p>Without bios:50.0%</p> <p>With bias: 100.0%</p>	

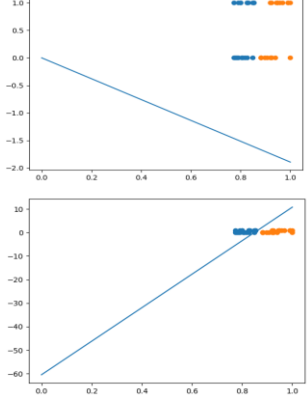
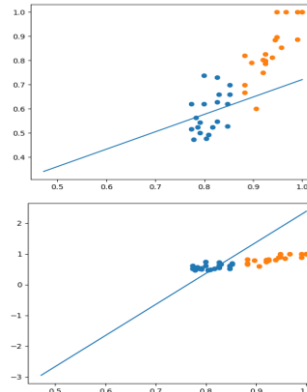
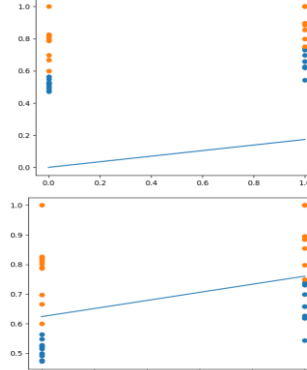
The best features to distinguish **between ‘Adelie’ and ‘Gentoo’** :

- flipper_length and gender
- bill_length and bill_depth
- Bill_dipth and body_mass

between “Adelie” and “Chinstrap”:

Features	Accuracy	plot
bill_length and bill_depth	Without bias:97.5% With bias: 100%	
bill_length and flipper length	Without bias:85.0% With bias: 100%	
bill_length and gender	Without bias:50.0% With bias:100%	

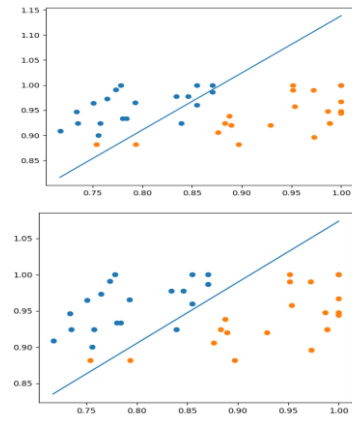
<p>bill_length and body_mass</p>	<p>Without bias: 50.0%</p> <p>With bias:95.0%</p>	
<p>Bill_dipth and flipper_length</p>	<p>Without bias: 82.5%</p> <p>With bias:87.5%~92%</p>	
<p>Bill_dipth and gender</p>	<p>Without bias:50.0%</p> <p>With bias: 60.0%</p>	
<p>Bill_dipth and body_mass</p>	<p>Without bias:87.5%</p> <p>With bias: 92.5%</p>	

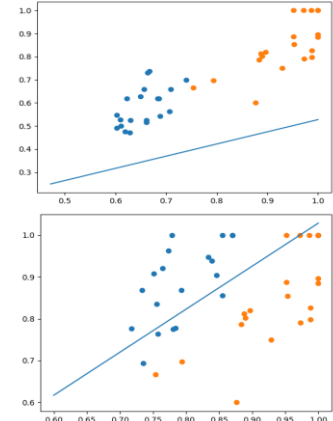
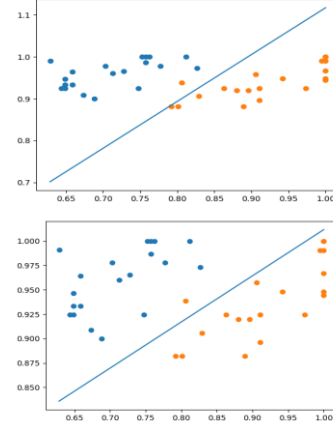
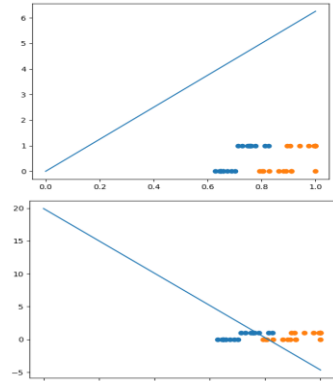
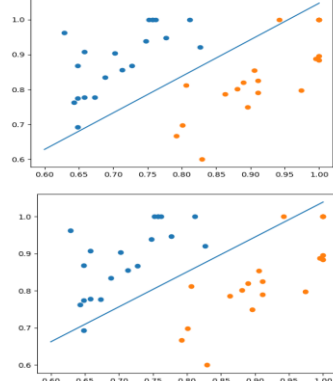
flipper_length and gender	<p>Without bias:50.0%</p> <p>With bias: 100.0%</p>	
flipper_length and body_mass	<p>Without bias:75.0%</p> <p>With bias: 80.0%</p>	
Gender and body_mass	<p>Without bias:50.0%</p> <p>With bias: 97.5%</p>	

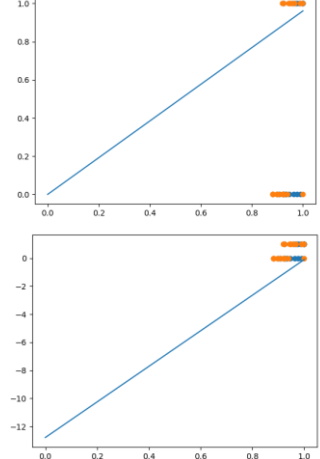
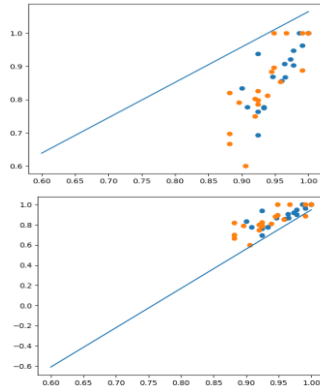
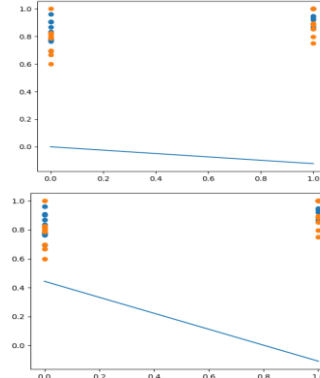
The best features to distinguish between **“Adelie”** and **“Chinstrap”**:

- flipper_length and gender
- bill_length and bill_depth
- Gender and body_mass

between “Gentoo” and “Chinstrap”:

Features	Accuracy	plot
bill_length and bill_depth:	Without bias:85.0% With bias: 90%	
bill_length and flipper length	Without bias:87.5% With bias: 92.5%	
bill_length and gender	Without bias:50.0% With bias:77.5%	

<p>bill_length and body_mass</p>	<p>Without bias: 50.0%</p> <p>With bias:80.0%</p>	
<p>Bill_dipth and flipper_length</p>	<p>Without bias: 97.5%</p> <p>With bias:92.5%</p>	
<p>Bill_dipth and gender</p>	<p>Without bias:50.0%</p> <p>With bias: 85.0%</p>	
<p>Bill_dipth and body_mass</p>	<p>Without bias:87.5%</p> <p>With bias: 95%</p>	

flipper_length and gender	<p>Without bias:50.0%</p> <p>With bias: 50.0%</p>	
flipper_length and body_mass	<p>Without bias:50.0%</p> <p>With bias: 47.5%</p>	
Gender and body_mass	<p>Without bias:50.0%</p> <p>With bias: 97.5%</p>	

The best features to distinguish **between “Adelie” and “Chinstrap”**:

- Bill_dipth and flipper_length
- Bill_dipth and body_mass
- Gender and body_mass