_	Temperature
In [1]:	<pre>import pandas as pd import numpy as np from scipy import stats from scipy.stats import fisher_exact</pre>
	<pre>data = pd.read_csv(filepath_or_buffer='///Archive/HTWTempRatios.csv') data["HTW"] = (data['DoS15km'] &gt;= 0.25)   (data['DoS20km'] &gt;= 0.25) temp = [21.7, 16.6, 13.6, 25, 18.9, 14.7, 15.1, 13.9, 20, 19.4] years = [2010,2011,2012,2013,2014,2015,2016,2017,2018,2019] runners = list(map (lambda x: len(data.loc[data['Year'] == x].index), years))</pre>
	<pre>female = list(map (lambda x: len(data.loc[(data['Year'] == x) &amp; (data['Gender'] == 'F')].index), years)) male = list(map (lambda x: len(data.loc[(data['Year'] == x) &amp; (data['Gender'] == 'M')].index), years)) avg_time = list(map (lambda x: ((data.loc[(data['Year'] == x) &amp; (data['Gender'] == 'F')])['Time']).mean(), years)) avg_time_f = list(map (lambda x: ((data.loc[(data['Year'] == x) &amp; (data['Gender'] == 'F')])['Time']).mean(), years)) avg_time_m = list(map (lambda x: ((data.loc[(data['Year'] == x) &amp; (data['Gender'] == 'M')])['Time']).mean(), years))</pre>
	<pre>htw = list(map (lambda x: len(data.loc[(data['Year'] == x) &amp; (data['HTW'] == True)].index), years)) htw_f = list(map (lambda x: len(data.loc[(data['Year'] == x) &amp; (data['HTW'] == True) &amp; (data['Gender'] == 'F')].index), years)) htw_m = list(map (lambda x: len(data.loc[(data['Year'] == x) &amp; (data['HTW'] == True) &amp; (data['Gender'] == 'M')].index), years)) neg_split = list(map (lambda x: len(data.loc[(data['Year'] == x) &amp; (data['SplitRatio'] &lt;= 1)].index), years)) neg_split_f = list(map (lambda x: len(data.loc[(data['Year'] == x) &amp; (data['SplitRatio'] &lt;= 1) &amp; (data['Gender'] == 'F')].index), years)) neg_split_m = list(map (lambda x: len(data.loc[(data['Year'] == x) &amp; (data['SplitRatio'] &lt;= 1) &amp; (data['Gender'] == 'M')].index), years))</pre>
	<pre>d = {'Year': years, 'Temp': temp, 'Runners': runners,'Female': female, 'Male': male,</pre>
	<pre>df = pd.DataFrame(data=d) df['HTW%'] = df['HTW'] / df['Runners'] df['F HTW%'] = df['HTW F'] / df['Female'] df['M HTW%'] = df['HTW M'] / df['Male'] df['Neg Split%'] = df['Neg Split'] / df['Runners'] df['F Neg Split%'] = df['Neg Split F'] / df['Female']</pre>
In [2]:	<pre>df['M Neg Split%'] = df['Neg Split M'] / df['Male']</pre>
Out[2]: -	Year         Temp         Runners         Female         Male         Avg Time         Avg Time M         HTW         HTW         Neg Split         Neg Split         Neg Split         HTW%         FHTW%         PHTW%         Neg Split         M Neg Split           0         2010         21.7         37982         10996         26986         7643.674056         8145.278192         7439.285148         5141         622         4519         2171         740         1431         0.135354         0.056566         0.167457         0.057159         0.067297         0.053027           1         2011         16.6         42838         13179         29659         7263.707736         7798.754989         7025.959068         2358         507         1851         5940         1597         4343         0.055045         0.038470         0.062409         0.138662         0.121178         0.146431
	2       2012       13.6       44094       13750       30344       7208.821881       7749.826618       6963.672423       2768       607       2161       6297       1725       4572       0.062775       0.044145       0.071217       0.142809       0.125455       0.150672         3       2013       25.0       44919       14814       30105       7747.183441       8205.571756       7521.620761       6189       1123       5066       2997       1178       1819       0.137781       0.075807       0.168278       0.066720       0.079519       0.060422         4       2014       18.9       47187       16323       30864       7458.688389       7989.805060       7177.797466       5007       1179       3828       3165       1058       2107       0.106110       0.072229       0.124028       0.067074       0.064817       0.068267
	5       2015       14.7       46207       16086       30121       7335.203281       7845.001865       7062.947379       3339       794       2545       5139       1592       3547       0.072262       0.049360       0.084493       0.111217       0.098968       0.117758         6       2016       15.1       44972       15662       29310       7343.250111       7876.157323       7058.487479       2522       573       1949       4442       1601       2841       0.056079       0.036585       0.066496       0.098773       0.102222       0.096929         7       2017       13.9       42252       14557       27695       7323.356078       7848.933915       7047.102726       2220       559       1661       5803       1834       3969       0.052542       0.038401       0.059975       0.137343       0.125987       0.143311         8       2018
	8 2018 20.0 39911 13775 26136 7519.654431 8079.968494 7224.340450 3614 814 2800 3133 1035 2098 0.090551 0.059093 0.107132 0.078500 0.075136 0.080272  9 2019 19.4 33134 11267 21867 7492.973200 8066.226502 7197.603695 3183 757 2426 2399 611 1788 0.096064 0.067187 0.110943 0.072403 0.054229 0.081767  Average times by temperature
In [3]:	<pre>slope, intercept, r_value, p_value, std_err = stats.linregress(df['Temp'],df['Avg Time']) print ("Linear regression all runners finish time") print ("r-squared:", r_value**2) print ("p: ", p_value)</pre>
	print ("intercept: ", intercept) print ("slope: ", slope) print ("std err: ", std_err)  Linear regression all runners finish time
	r-squared: 0.9111089869658573 p: 1.7715249036464247e-05 intercept: 6655.2428230633195 slope: 43.510812593271396 std err: 4.805028447513403
In [4]:	<pre>slope, intercept, r_value, p_value, std_err = stats.linregress(df['Temp'],df['Avg Time F']) print ("Linear regression female finish time") print ("r-squared:", r_value**2) print ("p: ", p_value) print ("intercept: ", intercept)</pre>
	<pre>print ("slope: ", slope) print ("std err: ", std_err)  Linear regression female finish time r-squared: 0.8979538035167087</pre>
	p: 3.094503499458341e-05 intercept: 7248.00461670852 slope: 39.82939378516619 std err: 4.74711897897793
	<pre>slope, intercept, r_value, p_value, std_err = stats.linregress(df['Temp'],df['Avg Time M']) print ("Linear regression male finish time") print ("r-squared:", r_value**2) print ("p: ", p_value) print ("intercept: ", intercept) print ("slope: ", slope)</pre>
	print ("std err: ", std_err)  Linear regression male finish time r-squared: 0.9098993795605811 p: 1.8709167068037667e-05 intercent: 6338 6711042585885
	intercept: 6338.6711042585885 slope: 46.57409475643498 std err: 5.181632825176719  HTW rates by temperature
In [6]:	<pre>slope, intercept, r_value, p_value, std_err = stats.linregress(df['Temp'],df['HTW']) print("Number of runners HTW by temperature") print ("r-squared:", r_value**2) print ("p: ", p_value)</pre>
	print ("slope: ", slope) print ("std err: ", std_err)  Number of runners HTW by temperature r-squared: 0.7463542610048088 p: 0.0012689337299441673
	<pre>slope: 312.5197011887763 std err: 64.4130233512836  #linear regression for men HTW by temperature. slope, intercept, r_value, p_value, std_err = stats.linregress(df['Temp'],df['HTW M'])</pre>
	<pre>print("Male runners HTW by temperature") print ("r-squared:", r_value**2) print ("p: ", p_value) print ("slope: ", slope) print ("std err: ", std_err)</pre>
	Male runners HTW by temperature r-squared: 0.7585656976834076 p: 0.0010352165980106564 slope: 274.76796377663305 std err: 54.80548035366965
In [8]:	<pre>#linear regression for women HTW by temperature. #It's not a good fit and women seem to not increase their risk of HTW as much as men. slope, intercept, r_value, p_value, std_err = stats.linregress(df['Temp'],df['HTW F']) print("Female runners HTW by temperature")</pre>
	<pre>print ("r-squared:", r_value**2) print ("p: ", p_value) print ("slope: ", slope) print ("std err: ", std_err)</pre> Female runners HTW by temperature
	r-squared: 0.3666048334437487 p: 0.06359434553891391 slope: 37.751737412143335 std err: 17.544065061839188
In [9]:	<pre>#linear regression for % of men HTW by temperature. slope, intercept, r_value, p_value, std_err = stats.linregress(df['Temp'],df['M HTW%']) print("% Male runners HTW by temperature") print ("r-squared:", r_value**2) print ("p: ", p_value)</pre>
	<pre>print ("slope: ", slope) print ("std err: ", std_err)  % Male runners HTW by temperature r-squared: 0.8455525035440344 p: 0.0001662461930069671</pre>
	<pre>slope: 0.010041298056111033 std err: 0.0015172774786787909  #linear regression for % of women HTW by temperature. slope, intercept, r_value, p_value, std_err = stats.linregress(df['Temp'],df['F HTW%'])</pre>
	<pre>print("% Female runners HTW by temperature") print ("r-squared:", r_value**2) print ("p: ", p_value) print ("slope: ", slope) print ("std err: ", std_err)</pre>
	<pre>% Female runners HTW by temperature r-squared: 0.66332037137115 p: 0.0041185241140052785 slope: 0.003169793192571831 std err: 0.000798422352730542</pre>
F	Negative Split Rates by Temperature  For the negative splits, the relationship isn't really linear (see plots), which will be shown below, but rather there are two clusters, below and above 18 degrees, when more runners manage a negative split in the
In [11]:	<pre>slope, intercept, r_value, p_value, std_err = stats.linregress(df['Temp'],df['Neg Split']) print ("r-squared:", r_value**2)</pre>
	<pre>print ("p: ", p_value) print ("slope: ", slope) print ("std err: ", std_err)  r-squared: 0.6841474625576467 p: 0.003153405979409332</pre>
	<pre>p: 0.003153405979409332 slope: -343.9905655168061 std err: 82.63591930164884  slope, intercept, r_value, p_value, std_err = stats.linregress(df['Temp'],df['Neg Split M'])</pre>
	<pre>print ("r-squared:", r_value**2) print ("p: ", p_value) print ("slope: ", slope) print ("std err: ", std_err)</pre> r-squared: 0.6863935996805802
	p: 0.0030608845197343205 slope: -257.58702837444486 std err: 61.55804823171513 slope, intercept, r_value, p_value, std_err = stats.linregress(df['Temp'],df['Neg Split F'])
	<pre>print ("r-squared:", r_value**2) print ("p: ", p_value) print ("slope: ", slope) print ("std err: ", std_err)</pre>
	r-squared: 0.569516782636373 p: 0.011643943985669154 slope: -86.40353714236127 std err: 26.55897911788878  As the number of pegative splits isn't linear it appear to form two clusers one for the cold years < 18 degrees, and one for the warm years > 18 degrees. Let's compute the differences between these conditions
In [14]:	As the number of negative splits isn't linear, it appear to form two clusers, one for the cold years < 18 degrees, and one for the warm years > 18 degrees. Let's compute the differences between these conditions.  ct = pd.crosstab(data['temperature'] < 18, data['SplitRatio'] <= 1)  #temperature = False means the temperature is above 18 degrees C, temperature = True that it is below  f_ct = pd.crosstab((data.loc[data["Gender"] == "F"])['temperature'] < 18, data['SplitRatio'] <= 1)
Out[14]:	<pre>m_ct = pd.crosstab((data.loc[data["Gender"] == "M"])['temperature'] &lt; 18, data['SplitRatio'] &lt;= 1) ct  SplitRatio False True temperature</pre>
	False 189268 13865 True 192742 27621
In [15]: Out[15]:	f_ct  SplitRatio False True temperature
	False 62553 4622 True 64885 8349
In [16]: Out[16]:	SplitRatio False True temperature
In [17]:	False 126715 9243  True 127857 19272
/ ] =	<pre>oddsr, p = fisher_exact(ct) oddsrF, pF = fisher_exact(f_ct) oddsrM, pM = fisher_exact(m_ct) print("Neagtive Splits warm vs cold years: \n p: ", p, "OR: ", oddsr) print("Neagtive Splits warm vs cold years (Female): \n p: ", pF, "OR: ", oddsrF) print("Neagtive Splits warm vs cold years (Men): \n p: ", pM, "OR: ", oddsrM)</pre>
	Neagtive Splits warm vs cold years: p: 0.0 OR: 1.9562319862232438  Neagtive Splits warm vs cold years (Female): p: 1.50459121297107e-190 OR: 1.7414393511243988
In [ ]:	Neagtive Splits warm vs cold years (Men): p: 0.0 OR: 2.0664140774948905