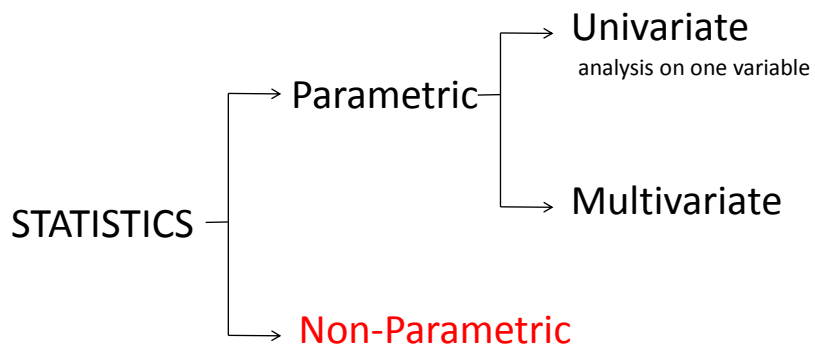


Non-Parametric Statistics for Research in Conservation

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When you need to use non-parametric statistics?

DEPEND ON:

- Scales of measurements
- Distribution of your data
 - parametric stats: data distributions are approximately normal
 - need to check the skew and kurtosis
 - for a relatively normal distribution: skew \approx 1.0, kurtosis \approx 1.0
- Number of samples: small samples \rightarrow difficult to assume normality

Scales of Measurements



higher

- Nominal: naming
- Ordinal: ordering
- Interval: equal interval without absolute zero
 - temperature, calendar year
- Ratio: equal interval with absolute zero

	Nominal	Ordinal	Interval	Ratio
Sex	x			
Hair colour	x			
Pulse				x
Temp. °C			x	
Team number	x			
Shoe size		x		
Year of birth			x	
Friends	x			
Fruit/veg	x			
Health		x		
Pulse				x
Temp.			x	
Residence	x			

When you need to use non-parametric statistics?

- If your measurement is **interval or ratio** scales → use parametric statistics
→ *Parametric stats cannot be used on rank data!*
- If you are using **nominal or ordinal** → use non-parametric statistics
- Non-parametric statistics: use less information in calculation → less powerful
→ Data on ratio scales can be analyzed by using non-parametric stats

Type of Statistic		Scale of Measurement
Nonparametric	-	Nominal
Nonparametric	-	Ordinal
Nonparametric	Parametric	Approximately Interval
-	Parametric	Interval
-	Parametric	Ratio

Non-Parametric Statistics

- “Distribution-free statistics”
- No assumptions about the distribution of the data (e.g., normality)
- Use RANKS rather than their raw values → we might lose some important info
 “How much do we want to lose information by transforming numbers into ranks?”

Non-Parametric Statistics

- Calculation: simple, fast and easy
- No need to check or assume normality
- Good for small samples

BUT:

- Tests are not as varied as parametric stats, some tests (e.g., regression/prediction) are unavailable
- Again .. Less powerful than parametric stats

Nominal Data	Ordinal Data
Chi-Square Goodness-of-Fit Test	Mann-Whitney U Test
Chi-Square Test of Independence	Wilcoxon T Test
McNemar Test	Kruskal-Wallis H Test
	Friedman ANOVA by Ranks
	Spearman's r_s

Chi-Square

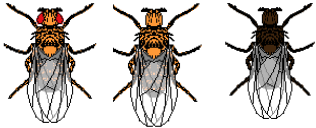
- To compare observed data with data we would expect to obtain according to a specific hypothesis
- Actually is a non-parametric statistics, although we also learn this test in parametric stats
- Chi-square → nominal data → frequency

- Formula:

$$\chi^2 = \sum_{i=1}^n \frac{(O_i - E_i)^2}{E_i}$$

O_i = observed values for each group

E_i = expected values for each group



Heritability of Fruit flies

Phenotype	Observed	Hypothesis
Wild	5610	5634
Eyeless	1881	1878
Black body	1896	1878
Eyeless, black body	622	626
Total	10009	

9

3

3

1

Mann–Whitney U test

- Also known as Wilcoxon Rank-Sum Test
- The U test is useful in the same situations as the independent samples Student's t -test
- One of the best-known non-parametric significance tests
- **Formula:** (select the highest; be careful in using critical value table)

$$U_1 = n_1 n_2 + \frac{n_1(n_1 + 1)}{2} - R_1$$

$$U_2 = n_1 n_2 + \frac{n_2(n_2 + 1)}{2} - R_2$$

Nominal Tail Probabilities: Directional H_A $\alpha = 0.025$; Non-directional H_A $\alpha = 0.05$

$n_1 \backslash n_2$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
1																
2								16	18	20	22	23	25	27	29	31
3					15	17	20	22	25	27	30	32	35	37	40	42
4				16	19	22	25	28	32	35	38	41	44	47	50	53
5			15	19	23	27	30	34	38	42	46	49	53	57	61	65
6			17	22	27	31	36	40	44	49	53	58	62	67	71	75
7			20	25	30	36	41	46	51	56	61	66	71	76	81	86
8		16	22	28	34	40	46	51	57	63	69	74	80	86	91	97
9		18	25	32	38	44	51	57	64	70	76	82	89	95	101	107
10		20	27	35	42	49	56	63	70	77	84	91	97	104	111	118
11		22	30	38	46	53	61	69	76	84	91	99	106	114	121	129
12		23	32	41	49	58	66	74	82	91	99	107	115	123	131	139
13		25	35	44	53	62	71	80	89	97	106	115	124	132	141	149
14		27	37	47	57	67	76	86	95	104	114	123	132	141	151	160
15		29	40	50	61	71	81	91	101	111	121	131	141	151	161	170
16		31	42	53	65	75	86	97	107	118	129	139	149	160	170	181
17		32	45	57	68	80	91	102	114	125	136	147	158	169	180	191
18		34	47	60	72	84	96	108	120	132	143	155	167	178	190	202
19		36	50	63	76	89	101	114	126	138	151	163	175	188	200	212
20		38	52	66	80	93	106	119	132	145	158	171	184	197	210	222

Wilcoxon T Test

- Also known as Wilcoxon paired-sample test
- To compare two related samples, matched samples, or repeated measurements on a single sample
- To assess whether their population mean ranks differ → paired difference test
- Alternative of paired sample t-test

Table W Critical Values of the Wilcoxon T Test

n	Level of Significance for a One-Tailed Test				n	Level of Significance for a One-Tailed Test			
	.05	.025	.01	.005		.05	.025	.01	.005
	Level of Significance for a Two-Tailed Test					Level of Significance for a Two-Tailed Test			
n	.10	.05	.02	.01	n	.10	.05	.02	.01
5	0	—	—	—	28	130	116	101	91
6	2	0	—	—	29	140	126	110	100
7	3	2	0	—	30	151	137	120	109
8	5	3	1	0	31	163	147	130	118
9	8	5	3	1	32	175	159	140	128
10	10	8	5	3	33	187	170	151	138
11	13	10	7	5	34	200	182	162	148
12	17	13	9	7	35	213	195	173	159
13	21	17	12	9	36	227	208	185	171
14	25	21	15	12	37	241	221	198	182
15	30	25	19	15	38	256	235	211	194
16	35	29	23	19	39	271	249	224	207
17	41	34	27	23	40	286	264	238	220
18	47	40	32	27	41	302	279	252	233
19	53	46	37	32	42	319	294	266	247
20	60	52	43	37	43	336	310	281	261
21	67	58	49	42	44	353	327	296	276
22	75	65	55	48	45	371	343	312	291
23	83	73	62	54	46	389	361	328	307
24	91	81	69	61	47	407	378	345	322
25	100	89	76	68	48	426	396	362	339
26	110	98	84	75	49	446	415	379	355
27	119	107	92	83	50	466	434	397	373

Kruskal-Wallis

- Identical to a one-way ANOVA with the data replaced by their ranks
- to decide whether k independent samples are from different populations
- An extension of the Mann-Whitney U test to 3 or more groups
- Formula:

$$H = \frac{12}{N(N+1)} \sum_{j=1}^k \frac{R_j^2}{n_j} - 3(N+1)$$

Sample Sizes	K = 3		K = 4		K = 5			
	$\alpha = 0.05$	$\alpha = 0.01$	Sample sizes	$\alpha = 0.05$	$\alpha = 0.01$	Sample sizes	$\alpha = 0.05$	$\alpha = 0.01$
2 2 2	-	-	2 2 1 1	-	-	2 2 1 1 1	-	-
3 2 1	-	-	2 2 2 1	6.679	-	2 2 2 1 1	6.750	-
3 2 2	4.714	-	2 2 2 2	6.167	6.667	2 2 2 2 1	7.133	7.633
3 3 1	5.143	-	3 1 1 1	-	-	2 2 2 2 2	7.418	8.291
3 3 2	5.361	-	3 2 1 1	-	-	3 1 1 1 1	-	-
3 3 3	5.600	7.200	3 2 2 1	6.633	-	3 2 1 1 1	6.663	-
4 2 1	-	-	3 2 2 2	6.333	7.133	3 2 2 1 1	6.900	7.600
4 2 2	5.333	-	3 3 1 1	6.333	-	3 2 2 2 1	7.309	8.127
4 3 1	5.208	-	3 3 2 1	6.244	7.200	3 2 2 2 2	7.662	8.662
4 3 2	5.444	6.444	3 3 2 2	6.527	7.636	3 3 1 1 1	7.111	-
4 3 3	5.791	6.745	3 3 3 1	6.600	7.400	3 3 2 1 1	7.200	8.073
4 4 1	4.967	6.667	3 3 3 2	6.727	8.015	3 3 2 2 1	7.691	8.676
4 4 2	5.466	7.036	3 3 3 3	7.000	8.638	3 3 2 2 2	7.910	8.116
4 4 3	5.698	7.144	4 1 1 1 1	-	-	3 3 3 1 1	7.676	8.424
4 4 4	5.692	7.664	4 2 1 1 1	6.633	-	3 3 3 2 1	7.769	9.061
5 2 1	5.000	-	4 2 2 1	6.133	7.000	3 3 3 2 2	8.044	9.606
5 2 2	5.160	6.633	4 2 2 2	6.646	7.391	3 3 3 3 1	8.000	9.461
5 3 1	4.960	-	4 3 1 1 1	6.176	7.067	3 3 3 3 2	8.200	9.676
5 3 2	5.261	6.909	4 3 2 1	6.309	7.466	3 3 3 3 3	8.333	10.20
5 3 3	5.648	7.079	4 3 2 2	6.621	7.671			
5 4 1	4.986	6.966	4 3 3 1	6.646	7.768			
5 4 2	5.273	7.206	4 3 3 2	6.796	8.333			
5 4 3	5.666	7.446	4 3 3 3	6.964	8.669			
5 4 4	5.667	7.760	4 4 1 1 1	6.946	7.909			
5 5 1	5.127	7.309	4 4 2 1	6.366	7.909			
5 5 2	5.938	7.336	4 4 2 2	6.731	8.346			
5 5 3	5.706	7.678	4 4 3 1	6.636	8.231			
5 5 4	5.666	7.823	4 4 3 2	6.874	8.621			
5 5 5	5.780	8.000	4 4 3 3	7.036	8.676			
6 1 1	-	-	4 4 4 1	6.726	8.666			

5	5	5	5.780	8.000
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Friedman Test

- To compare observations repeated on the same subjects
- A non-parametric randomized block ANOVA
- Formula:

$$M = \frac{12}{nk(k+1)} \sum R_j^2 - 3n(k+1)$$

Critical values for the Friedman Test

$$M = \frac{12}{nk(k+1)} \sum R_j^2 - 3n(k+1)$$

n	k=3		k=4		k=5		k=6	
	$\alpha=5\%$	$\alpha=1\%$	$\alpha=5\%$	$\alpha=1\%$	$\alpha=5\%$	$\alpha=1\%$	$\alpha=5\%$	$\alpha=1\%$
2	—	—	6.000	—	7.600	8.000	9.143	9.714
3	6.000	—	7.400	9.000	8.533	10.130	9.857	11.760
4	6.500	8.000	7.800	9.600	8.800	11.200	10.290	12.710
5	6.400	8.400	7.800	9.960	8.960	11.680	10.490	13.230
6	7.000	9.000	7.600	10.200	9.067	11.870	10.570	13.620
7	7.143	8.857	7.800	10.540	9.143	12.110	10.670	13.860
8	6.250	9.000	7.650	10.500	9.200	13.200	10.710	14.000
9	6.222	9.556	7.667	10.730	9.244	12.440	10.780	14.140
10	6.200	9.600	7.680	10.680	9.280	12.480	10.800	14.230
11	6.545	9.455	7.691	10.750	9.309	12.580	10.840	14.320
12	6.500	9.500	7.700	10.800	9.333	12.600	10.860	14.380
13	6.615	9.385	7.800	10.850	9.354	12.680	10.890	14.450

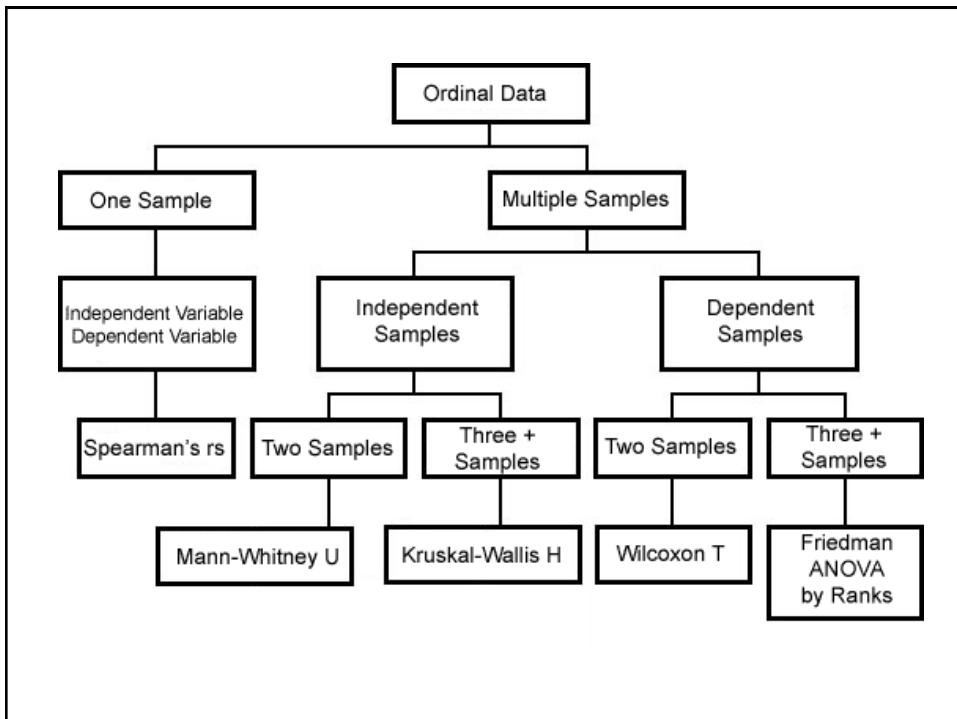
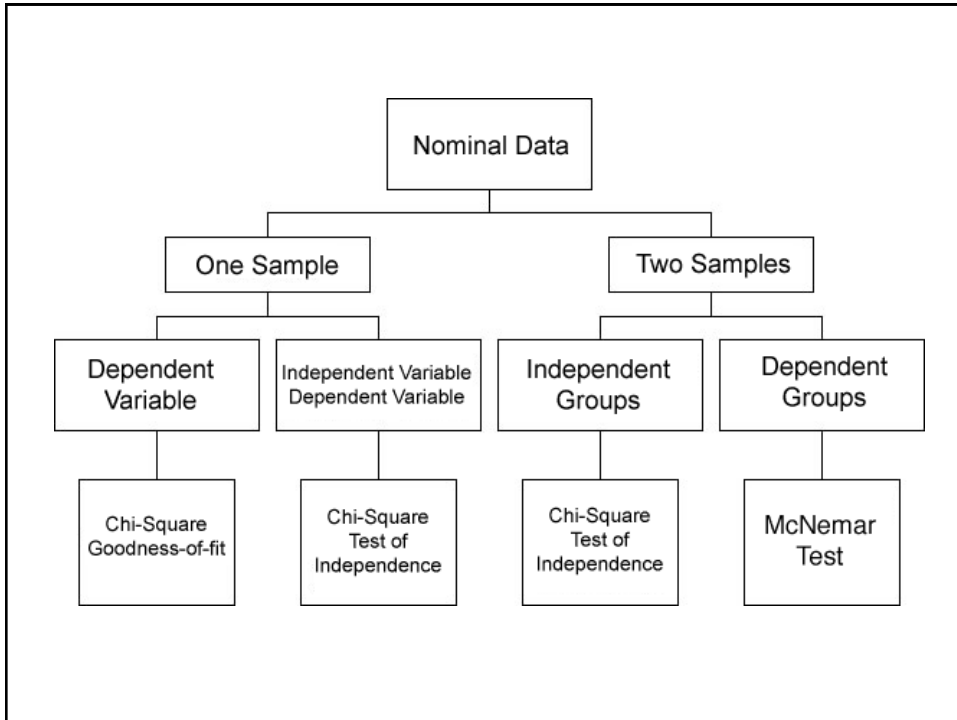
Spearman Rank Correlation Coefficient

- To test the direction and strength of the relationship between two variables
- Using ranks to calculate the correlation
- Identical to Pearson Product Moment
- Compute correlation only; unable to compute regression!

- Formula:

$$r_s = 1 - \frac{6 \sum d^2}{n(n^2 - 1)}$$

n	Nominal α					
	0.10	0.05	0.025	0.01	0.005	0.001
4	1.000	1.000	-	-	-	-
5	0.800	0.900	1.000	1.000	-	-
6	0.657	0.829	0.886	0.943	1.000	-
7	0.571	0.714	0.786	0.893	0.929	1.000
8	0.524	0.643	0.738	0.833	0.881	0.952
9	0.483	0.600	0.700	0.783	0.833	0.917
10	0.455	0.564	0.648	0.745	0.794	0.879
11	0.427	0.536	0.618	0.709	0.755	0.845
12	0.406	0.503	0.587	0.678	0.727	0.818
13	0.385	0.484	0.560	0.648	0.703	0.791
14	0.367	0.464	0.538	0.626	0.679	0.771
15	0.354	0.446	0.521	0.604	0.654	0.750
16	0.341	0.429	0.503	0.582	0.635	0.729
17	0.328	0.414	0.488	0.566	0.618	0.711
18	0.317	0.401	0.472	0.550	0.600	0.692
19	0.309	0.391	0.460	0.535	0.584	0.675
20	0.299	0.380	0.447	0.522	0.570	0.662
21	0.292	0.370	0.436	0.509	0.556	0.647
22	0.284	0.361	0.425	0.497	0.544	0.633
23	0.278	0.353	0.416	0.486	0.532	0.621
24	0.271	0.344	0.407	0.476	0.521	0.609
25	0.265	0.337	0.398	0.466	0.511	0.597
26	0.259	0.331	0.390	0.457	0.501	0.586
27	0.255	0.324	0.383	0.449	0.492	0.576
28	0.250	0.318	0.375	0.441	0.483	0.567
29	0.245	0.312	0.368	0.433	0.475	0.558



- Check your data measurements
- Your data: do they meet necessary assumptions?
- Use non-parametric tests with caution
- Prioritize parametric tests before selecting non-parametric tests

THANK YOU