PIMs as classical distribution free tests

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Contents

1 Introduction

While?) introduces PIMs, they were shown to be extensions of the known distribution free tests, like Wilcoxon in TODO:rankpaperref.

A general estimator for the variance under the null hypothesis has been created (variance estimator. H0), but for these special cases, some simplified formulas exist, which have been provided through the simplified* set of functions.

For each of the code sections below, it is easily checked that the simplified formulas give the same result as the generic code, and that the links between the known distribution free tests and PIM are confirmed.

The code also (on the final line of each block) the classical test function that has been provided. This function uses the PIM to calculate the matching test statistic. The advantage (although not shown in this vignette) is that another way of calculating the (co)variances can be specified. In particular, and as indicated in TODO ref rankpaper, the varianceestimator sandwich can be passed along to get Wald-style tests for the same null hypotheses.

2 Wilcoxon-Mann-Whitney

Code to check the equivalence (note this includes a legacy implementation):

- > library(pim)
- > set.seed(1)
- > wmw1<-demo.WilcoxonMannWhitney()</pre>
- > wmw1\$pim2<-pim(y~F(x)-1, data=wmw1\$dta, link="identity", poset=pairwisepose
- > wmw1\$pim2

```
Call:
pim(frm = y - F(x) - 1, data = wmw1$dta, link = "identity", poset = pairwiseposet,
    varianceestimator = varianceestimator.HO(), keep.data = TRUE,
    verbosity = 0)
Coefficients:
x_R_L_1_2
0.7937182
          #Simplified formulas
          simplifiedpimestimation.pairwisecoefficients(wmw1$dta, out="y", group="x")$
[1] 0.7937182
          simplifiedpimestimation.pairwisecovariance(wmw1$dta, out="y", group="x")
            1 - 2
1 - 2 0.003572439
          #From applying generic code:
          wmw1$pim2$coefficients
x_R_L_1_2
0.7937182
          wmw1$pim2$vcov[1,1]
[1] 0.003572439
          #Standardized WMW based on wilcoxon test
          wmw1$legacy<-legacy.WilcoxonMannWhitney(data=wmw1$dta, out="y", group="x")
          wmw1$legacy$statistic
       W
-4.91415
          wmw1$legacy$conversion(wmw1$pim2$coefficients, wmw1$pim2$vcov)
x_R_L_1_2
 4.91415
          classical.test(test="WilcoxonMannWhitney", data=wmw1$dta, out="y", group="x
x_R_L_1_2
  4.91415
```

3 Kruskal-Wallis

```
Code to check the equivalence (note this includes a legacy implementation):
         kw1<-demo.KruskalWallis()</pre>
         kw1$pim3<-pim(y~F(x)-1, data=kw1$dta, link="identity", poset=fullposet, for
>
         kw1$pim3
Call:
pim(frm = y ~ F(x) - 1, data = kw1$dta, link = "identity", poset = fullposet,
   force.marginal = TRUE, varianceestimator = varianceestimator.HO(),
   keep.data = TRUE, verbosity = 0)
Coefficients:
   x_R_1
             x_R_2
                      x_R_3
0.2863043 0.4330952 0.7207143
         #Simplified formulas (lemma 1)
         simplified pime stimation. \verb|marginal| coefficients (kw1\$dta, out="y", group="x")
       1
0.2863043 0.4330952 0.7207143
         simplifiedpimestimation.marginalcovariance(kw1$dta, out="y", group="x")
>
             1
1 0.0028177536 -0.0008416667 -0.0008416667
3 -0.0008416667 -0.0008416667 0.0015630952
         #From applying generic code:
         kw1$pim3$coefficients
   x_R_1
             x_R_2
                      x_R_3
0.2863043 0.4330952 0.7207143
         kw1$pim3$vcov
             x_R_1
                          x_R_2
x_R_1 0.0028177536 -0.0008416667 -0.0008416667
x_R_3 -0.0008416667 -0.0008416667 0.0015630952
         #Standardized KW based on Kruskal-Wallis test
```

kw1\$legacy\$statistic

>

kw1\$legacy<-legacy.KruskalWallis(data=kw1\$dta, out="y", group="x")</pre>

```
Kruskal-Wallis chi-squared
                  34.97031
          kw1$legacy$conversion(kw1$pim3$coefficients, kw1$pim3$vcov)
         [,1]
[1,] 34.97031
          classical.test(test="KruskalWallis", data=kw1$dta, out="y", group="x")$stat
         [,1]
[1,] 34.97031
    Mack-Skillings
4
Code to check the equivalence (note this includes a legacy implementation):
          mss1<-demo.MackSkillings()</pre>
>
          mss1pim1<-pim(y~F(x)-1, data=mss1$dta, link="identity", blocking.variables
          mss1$pim1
Call:
pim(frm = y ~ F(x) - 1, data = mss1$dta, link = "identity", blocking.variables = "b",
   poset = fullposet, force.marginal = TRUE, varianceestimator = varianceestimator.HO(
    keep.data = TRUE, verbosity = 0)
Coefficients:
    x_R_1
            x_R_2
                        x_R_3
0.2858333 0.5225000 0.6916667
          #Simplified formulas (lemma 4)
          simplifiedpimestimation.marginalcoefficients(mss1$dta, out="y", group="x",
        1
                  2
                            3
0.2858333 0.5225000 0.6916667
>
          simplifiedpimestimation.marginalcovariance(mss1$dta, out="y", group="x", bl
              1
1 0.0014351852 -0.0007175926 -0.0007175926
2 -0.0007175926  0.0014351852 -0.0007175926
3 -0.0007175926 -0.0007175926 0.0014351852
          #From applying generic code:
```

mss1\$pim1\$coefficients

>

```
0.2858333 0.5225000 0.6916667
          mss1$pim1$vcov
                            x_R_2
              x_R_1
x_R_1 0.0014351852 -0.0007175926 -0.0007175926
x_R_2 -0.0007175926  0.0014351852 -0.0007175926
x_R_3 -0.0007175926 -0.0007175926 0.0014351852
>
          #Standardized MS based on Mack-Skillings test
          mss1$legacy<-legacy.MackSkillings(data=mss1$dta, out="y", group="x", block=
          mss1$legacy$statistic
[1] 38.60581
          mss1$legacy$conversion(mss1$pim1$coefficients, mss1$pim1$vcov)
         [,1]
[1,] 38.60581
          classical.test(test="MackSkillings", data=mss1$dta, out="y", group="x", blo
         [,1]
[1,] 38.60581
    Brown-Hettmansperger
5
Code to check the equivalence (note this includes a legacy implementation):
          bh1<-demo.BrownHettmansperger()</pre>
          bh1pim1<-pim(y~F(x)-1, data=bh1$dta, link="identity", poset=fullposet, var
          bh1$pim1
>
Call:
pim(frm = y \tilde{F}(x) - 1, data = bh1\$dta, link = "identity", poset = fullposet,
    varianceestimator = varianceestimator.HO(), keep.data = TRUE,
    verbosity = 0)
Coefficients:
x_R_L_1_2 x_R_L_1_3 x_R_L_2_3
0.6454545 0.8746929 0.8081081
          #Simplified formulas (lemma 4)
          simplifiedpimestimation.pairwisecoefficients(bh1$dta, out="y", group="x")$b
>
```

x_R_1

x_R_2

 x_R_3

```
[1] 0.6454545 0.8746929 0.8081081
         simplifiedpimestimation.pairwisecovariance(bh1$dta, out="y", group="x")
            1 - 2
                       1 - 3
                                   2 - 3
1 - 2 0.005387205 0.002525253 -0.002777778
1 - 3 0.002525253 0.004845755 0.002252252
2 - 3 -0.002777778 0.002252252 0.005105105
         #From applying generic code:
         bh1$pim1$coefficients
x_R_L_1_2 x_R_L_1_3 x_R_L_2_3
0.6454545 0.8746929 0.8081081
         bh1$pim1$vcov
            x_R_L_2_3
x_R_L_1_3  0.002525253 0.004845755 0.002252252
x_R_L_2_3 -0.002777778 0.002252252 0.005105105
         #Standardized BH based on Brown-Hettmansperger test
>
         bh1$legacy<-legacy.BrownHettmansperger(data=bh1$dta, out="y", group="x")
         bh1$legacy$statistic
Kruskal-Wallis chi-squared
                62.44696
         bh1$legacy$conversion(bh1$pim1$coefficients, bh1$pim1$vcov)
        [,1]
[1,] 62.44696
         classical.test(test="BrownHettmansperger", data=bh1$dta, out="y", group="x"
        [,1]
[1,] 62.44696
```

6 Jonckheere-Terpstra

```
Code to check the equivalence (note this includes a legacy implementation):
         jt1<-demo.JonckheereTerpstra(force.balanced=FALSE)</pre>
>
         jt1$pim1<-pim(y^F(x)-1, data=jt1$dta, link="identity", poset=pairwiseposet,
>
         jt1$pim1
Call:
pim(frm = y ~ F(x) - 1, data = jt1$dta, link = "identity", poset = pairwiseposet,
   varianceestimator = varianceestimator.HO(), keep.data = TRUE,
   verbosity = 0)
Coefficients:
x_R_L_1_2 x_R_L_1_3 x_R_L_1_4 x_R_L_2_3 x_R_L_2_4 x_R_L_3_4
0.7925926 0.9025641 0.9916667 0.7065527 0.9733796 0.9314904
         #Simplified formulas (lemma 4)
         simplifiedpimestimation.pairwisecoefficients(jt1$dta, out="y", group="x")$b
[1] 0.7925926 0.9025641 0.9916667 0.7065527 0.9733796 0.9314904
>
         simplifiedpimestimation.pairwisecovariance(jt1$dta, out="y", group="x")
            1 - 2
                        1 - 3
                                    1 - 4
                                                2 - 3
                                                             2 - 4
1 \; - \; 2 \quad 0.008847737 \quad 0.005555556 \; 0.005555556 \; - 0.003086420 \; - 0.003086420
1 - 3 0.005555556 0.008974359 0.005555556 0.003205128 0.000000000
1 - 4 \quad 0.005555556 \quad 0.005555556 \quad 0.008333333 \quad 0.000000000 \quad 0.002604167
3 - 4 0.000000000 -0.003205128 0.002604167 -0.003205128 0.002604167
            3 - 4
1 - 2 0.000000000
1 - 3 -0.003205128
1 - 4 0.002604167
2 - 3 -0.003205128
2 - 4 0.002604167
3 - 4 0.005909455
         #From applying generic code:
         jt1$pim1$coefficients
```

x_R_L_1_2 x_R_L_1_3 x_R_L_1_4 x_R_L_2_3 x_R_L_2_4 x_R_L_3_4 0.7925926 0.9025641 0.9916667 0.7065527 0.9733796 0.9314904

```
>
          jt1$pim1$vcov
            x_R_L_1_2
                         x_R_L_1_3
                                     x_R_L_1_4
                                                  x_R_L_2_3
                                                               x_R_L_2_4
x_R_L_1_2 0.008847737 0.005555556 0.005555556 -0.003086420 -0.003086420
x_R_L_1_3 0.005555556 0.008974359 0.005555556 0.003205128 0.000000000
x_R_L_1_4 0.005555556 0.005555556 0.008333333 0.000000000 0.002604167
x_R_L_2_3 -0.003086420 0.003205128 0.000000000 0.006410256 0.003086420
x_R_L_2_4 -0.003086420 0.000000000 0.002604167 0.003086420 0.005787037
x_R_L_3_4 0.000000000 -0.003205128 0.002604167 -0.003205128 0.002604167
            x_R_L_3_4
x_R_L_1_2 0.00000000
x_R_L_1_3 -0.003205128
x_R_L_1_4 0.002604167
x_R_L_2_3 -0.003205128
x_R_L_2_4 0.002604167
x_R_L_3_4 0.005909455
          #Standardized JT based on Jonckheere-Terpstra test
          jt1$legacy<-legacy.JonckheereTerpstra(data=jt1$dta, out="y", group="x", ver
>
mu: 1836.5
sigsq: 26044.92
mainterm: 3261
          jt1$legacy$statistic
[1] 8.826753
          jt1$legacy$conversion(jt1$pim1$coefficients, jt1$pim1$vcov)
         [,1]
[1,] 8.826753
          classical.test(test="JonckheereTerpstra", data=jt1$dta, out="y", group="x")
         [,1]
[1,] 8.826753
```

7 Mack-Wolfe

Code to check the equivalence (note this includes a legacy implementation):

```
> mw1<-demo.MackWolfe(force.balanced=FALSE)
> mw1$pim1<-pim(y~F(x)-1, data=mw1$dta, link="identity", poset=pairwiseposet,
> mw1$pim1
```

```
Call:
pim(frm = y ~ F(x) - 1, data = mw1$dta, link = "identity", poset = pairwiseposet,
   varianceestimator = varianceestimator.HO(), keep.data = TRUE,
   verbosity = 0)
Coefficients:
x_R_L_1_2 x_R_L_1_3 x_R_L_1_4 x_R_L_2_3 x_R_L_2_4 x_R_L_3_4
0.76562500 1.00000000 0.93005952 0.98828125 0.68526786 0.02790179
         #Simplified formulas (lemma 4)
         simplifiedpimestimation.pairwisecoefficients(mw1$dta, out="y", group="x")$b
[1] 0.76562500 1.00000000 0.93005952 0.98828125 0.68526786 0.02790179
         simplifiedpimestimation.pairwisecovariance(mw1$dta, out="y", group="x")
            1 - 2
                        1 - 3
                                   1 - 4
                                               2 - 3
                                                           2 - 4
1 - 2 0.008897569 0.003472222 0.003472222 -0.005208333 -0.005208333
1 - 3 \quad 0.003472222 \quad 0.006184896 \quad 0.003472222 \quad 0.002604167 \quad 0.000000000
1 \; - \; 4 \quad 0.003472222 \quad 0.003472222 \quad 0.006572421 \quad 0.000000000 \quad 0.002976190
3 - 4 0.000000000 -0.002604167 0.002976190 -0.002604167 0.002976190
            3 - 4
1 - 2 0.000000000
1 - 3 -0.002604167
1 - 4 0.002976190
2 - 3 -0.002604167
2 - 4 0.002976190
3 - 4 0.005673363
         #From applying generic code:
         mw1$pim1$coefficients
x_R_L_{1_2} x_R_L_{1_3} x_R_L_{1_4} x_R_L_{2_3} x_R_L_{2_4} x_R_L_{3_4}
0.76562500 1.00000000 0.93005952 0.98828125 0.68526786 0.02790179
         mw1$pim1$vcov
>
            x_R_L_1_2
                        x_R_L_2_3
                                                           x_R_L_2_4
x_R_L_1_2 0.008897569 0.003472222 0.003472222 -0.005208333 -0.005208333
```

x_R_L_1_4 0.003472222 0.003472222 0.006572421 0.000000000 0.002976190 x_R_L_2_3 -0.005208333 0.002604167 0.000000000 0.007975260 0.005208333

0.00000000

x_R_L_1_3 0.003472222 0.006184896 0.003472222 0.002604167

```
x_R_L_2_4 -0.005208333 0.000000000 0.002976190 0.005208333 0.008370536
x\_R\_L\_3\_4 \quad 0.000000000 \quad -0.002604167 \quad 0.002976190 \quad -0.002604167 \quad 0.002976190
             x_R_L_3_4
x_R_L_1_2 0.000000000
x_R_L_1_3 -0.002604167
x_R_L_1_4 0.002976190
x_R_L_2_3 -0.002604167
x_R_L_2_4 0.002976190
x_R_L_3_4 0.005673363
          #Standardized MW based on Mack-Wolfe test
          mw1$legacy<-legacy.MackWolfe(data=mw1$dta, out="y", group="x", levelP=as.ch
mu: 1280
sigsq: 19626.67
leftterm: 1568
rightterm: 871
          mw1$legacy$statistic
[1] 8.272945
          mw1$legacy$conversion(mw1$pim1$coefficients, mw1$pim1$vcov)
         [,1]
[1,] 8.272945
          classical.test(test="MackWolfe", data=mw1$dta, out="y", group="x",levelP=as
         [,1]
[1,] 8.272945
```