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()
> wmw1$pim2<-pim(y~F(x)-1, data=wmw1$dta, link="identity", poset=pairwisepose
+ varianceestimator=
+ interpretation="re
> wmw1$pim2
```

```
Call:
pim(formula = y ~ F(x) - 1, data = wmw1$dta, link = "identity",
    poset = pairwiseposet, interpretation = "regular", varianceestimator = varianceesti
    keep.data = TRUE, verbosity = 0)
Coefficients:
x_R_L_1_2
0.7937182
          #Simplified formulas
          simplifiedpimestimation.pairwisecoefficients(wmw1$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$pim3<-pim(y^{r}F(x)-1, data=kw1$dta, link="identity", poset=fullposet, int
>
                                                                 varianceestimator=v
>
         kw1$pim3
Call:
pim(formula = y ~ F(x) - 1, data = kw1$dta, link = "identity",
   poset = fullposet, interpretation = "marginal", varianceestimator = varianceestimat
   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)
          simplifiedpimestimation.marginalcoefficients(kw1$dta, out="y", group="x")
0.2863043 0.4330952 0.7207143
>
          simplifiedpimestimation.marginalcovariance(kw1$dta, out="y", group="x")
1 0.0028177536 -0.0008416667 -0.0008416667
3 -0.0008416667 -0.0008416667 0.0015630952
         #From applying generic code:
         kw1$pim3$coefficients
             x_R_2
                       x_R_3
    x_R_1
0.2863043 0.4330952 0.7207143
>
         kw1$pim3$vcov
             x_R_1
                           x_R_2
                                        x_R_3
x_R_1 0.0028177536 -0.0008416667 -0.0008416667
x_R_2 -0.0008416667  0.0011623016 -0.0008416667
```

x_R_3 -0.0008416667 -0.0008416667 0.0015630952

```
>
          kw1$legacy<-legacy.KruskalWallis(data=kw1$dta, out="y", group="x")</pre>
          kw1$legacy$statistic
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$dta, link="identity", blocking.variables
                                                                      poset=fullposet, i
                                                                      varianceestimator=
          mss1$pim1
Call:
pim(formula = y ~ F(x) - 1, data = mss1$dta, link = "identity",
    blocking.variables = "b", poset = fullposet, interpretation = "marginal",
    varianceestimator = varianceestimator.HO(), keep.data = TRUE,
    verbosity = 0)
Coefficients:
    x_R_1
             x_R_2
                        x_R_3
0.2858333 0.5225000 0.6916667
          #Simplified formulas (lemma 4)
          simplifiedpimestimation.marginalcoefficients(mss1$dta, out="y", group="x",
0.2858333 0.5225000 0.6916667
>
          simplifiedpimestimation.marginalcovariance(mss1$dta, out="y", group="x", bl
```

#Standardized KW based on Kruskal-Wallis test

```
1
1 0.0014351852 -0.0007175926 -0.0007175926
3 -0.0007175926 -0.0007175926 0.0014351852
         #From applying generic code:
         mss1$pim1$coefficients
   x_R_1
             x_R_2
                      x_R_3
0.2858333 0.5225000 0.6916667
         mss1$pim1$vcov
>
             x_R_1
                          x_R_2
                                       x_R_3
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,
                                                                      varianceestimator=v
+
                                                                       interpretation="reg
>
```

bh1\$pim1

```
Call:
pim(formula = y ~ F(x) - 1, data = bh1$dta, link = "identity",
    poset = fullposet, interpretation = "regular", varianceestimator = varianceestimato
    keep.data = TRUE, verbosity = 0)
Coefficients:
x_R_L_1_2 x_R_L_1_3 x_R_L_2_3
0.6454545 0.8746929 0.8081081
          #Simplified formulas (lemma 4)
          simplifiedpimestimation.pairwisecoefficients(bh1$dta, out="y", group="x")$b
[1] 0.6454545 0.8746929 0.8081081
          simplifiedpimestimation.pairwisecovariance(bh1$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_{1_2} x_R_L_{1_3} x_R_L_{2_3}
x_R_L_1_2 0.005387205 0.002525253 -0.002777778
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
          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$pim1<-pim(y^{r}F(x)-1, data=jt1$dta, link="identity", poset=pairwiseposet,
>
+
                                                              varianceestimator=v
                                                              interpretation="reg
         jt1$pim1
>
Call:
pim(formula = y ~ F(x) - 1, data = jt1$dta, link = "identity",
   poset = pairwiseposet, interpretation = "regular", varianceestimator = varianceesti
   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$dta, out="y", group="x")
                                   1 - 4
                                               2 - 3
1 - 2 0.008847737 0.005555556 0.005555556 -0.003086420 -0.003086420
1 - 3 0.005555556 0.008974359 0.005555556 0.003205128 0.000000000
1 - 4 \quad 0.005555556 \quad 0.005555556 \quad 0.008333333 \quad 0.000000000 \quad 0.002604167
3 - 4 0.000000000 -0.003205128 0.002604167 -0.003205128 0.002604167
            3 - 4
1 - 2 0.000000000
1 - 3 - 0.003205128
1 - 4 0.002604167
2 - 3 -0.003205128
2 - 4 0.002604167
3 - 4 0.005909455
         #From applying generic code:
         jt1$pim1$coefficients
```

```
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 \quad 0.005555556 \quad 0.005555556 \quad 0.008333333 \quad 0.000000000 \quad 0.002604167
x\_R\_L\_2\_3 \ -0.003086420 \quad 0.003205128 \ 0.000000000 \quad 0.006410256 \quad 0.003086420
 x\_R\_L\_2\_4 \ -0.003086420 \quad 0.000000000 \ 0.002604167 \quad 0.003086420 \quad 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 0.000000000
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]
```

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

[1,] 8.826753

7 Mack-Wolfe

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

```
mw1<-demo.MackWolfe(force.balanced=FALSE)</pre>
          mw1$pim1<-pim(y^{r}F(x)-1, data=mw1$dta, link="identity", poset=pairwiseposet,
>
+
                                                                 varianceestimator=v
                                                                  interpretation="reg
         mw1$pim1
>
Call:
pim(formula = y ~ F(x) - 1, data = mw1$dta, link = "identity",
   poset = pairwiseposet, interpretation = "regular", varianceestimator = varianceesti
   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 - 4
                                                  2 - 3
1 - 2 0.008897569 0.003472222 0.003472222 -0.005208333 -0.005208333
1 - 3 0.003472222 0.006184896 0.003472222 0.002604167 0.000000000
1 - 4 \quad 0.003472222 \quad 0.003472222 \quad 0.006572421 \quad 0.000000000 \quad 0.002976190
2 - 4 -0.005208333  0.000000000 0.002976190  0.005208333  0.008370536
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_1_3 $x_R_L_1_4$ $x_R_L_2_3$ $x_R_L_2_4$ x_R_L_1_2 0.008897569 0.003472222 0.003472222 -0.005208333 -0.005208333 x_R_L_1_3 0.003472222 0.006184896 0.003472222 0.002604167 0.000000000 $x_R_L_1_4 \quad 0.003472222 \quad 0.003472222 \quad 0.006572421 \quad 0.000000000 \quad 0.002976190$ $x_R_L_2_3 \ -0.005208333 \ \ 0.002604167 \ \ 0.000000000 \ \ \ 0.007975260 \ \ \ 0.005208333$ $x_R_L_2_4$ -0.005208333 0.000000000 0.002976190 0.005208333 0.008370536 $x_R_L_3_4$ 0.000000000 -0.002604167 0.002976190 -0.002604167 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",</pre> 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.characte [,1][1,] 8.272945