# Verbs PhysPsych: Graphs for Accuracy

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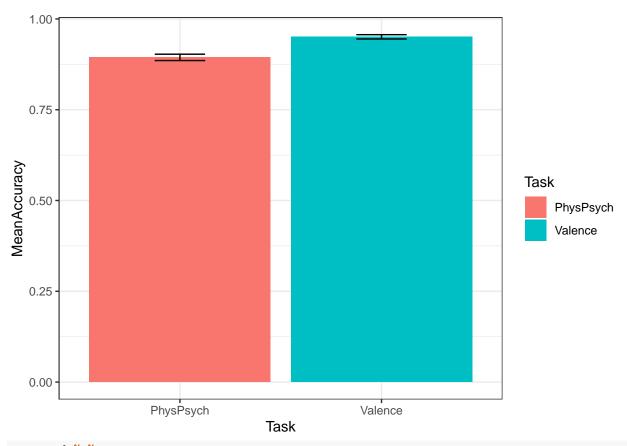
```
table(d$Task,d$Label)
##
##
                test_physpsych test_val
##
     PhysPsych
                           4800
##
     Valence
                              0
                                    4800
print(unique(d$Word))
    [1] "dishonor"
                      "adore"
                                    "like"
                                                  "respect"
                                                                "caress"
##
   [6] "decorate"
                                    "dismay"
                                                  "appreciate"
                      "content"
                                                                "stain"
## [11] "burn"
                      "embarrass"
                                    "hug"
                                                  "rust"
                                                                "discourage"
## [16] "clean"
                      "admire"
                                    "hit"
                                                  "displease"
                                                                "cuddle"
## [21] "crash"
                      "sculpt"
                                    "repair"
                                                  "corrode"
                                                                "slash"
## [26] "encourage"
                                    "twinkle"
                                                  "dread"
                      "offend"
                                                                "irritate"
## [31] "build"
                      "slaughter"
                                                  "despair"
                                                                "reassure"
                                    "tear"
## [36] "cherish"
                      "delete"
                                    "criticize"
                                                  "impress"
                                                                "produce"
```

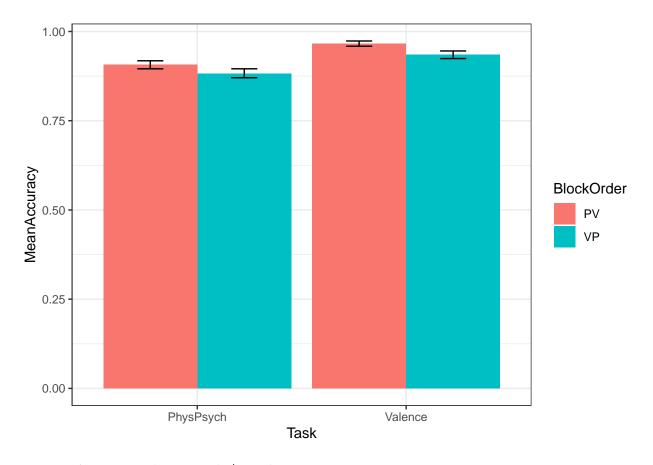
## Graph Accuracy by Word

Values for valence/concreteness were gathered/normed first from Warriner et al and Brysbaert et al. From those studies, we can establish what an Accurate response is.

A response is accurate (coded as 1) if the participant response was consistent with the norming study; innacurate (or 0) otherwise.

#### Overall Accuracy





#### Mean Accuracy by Word / Task

Looking at only the first block

```
agr <- d %>%
  group_by(Task,Word,BlockOrder) %>%
  filter((Task == "Valence") & (BlockOrder == "VP") |
           (Task == "PhysPsych") & (BlockOrder == "PV")) %>%
  mutate(MeanAccuracy = mean(Accuracy),
          CILow = ci.low(Accuracy),
          CIHigh = ci.high(Accuracy)) %>%
  mutate(YMin = MeanAccuracy - CILow,
         YMax = MeanAccuracy + CIHigh)
agrr <- agr %>%
  group_by(Word, Task) %>%
  select(Word, Task, MeanAccuracy) %>%
  unique()
dodge = position_dodge(.9)
ggplot(data=agr, aes(x=Task,y=MeanAccuracy,fill=BlockOrder)) +
  geom_bar(position=dodge,stat="identity") +
 facet_wrap(~Word,ncol=10) +
  geom_errorbar(aes(ymin=YMin,ymax=YMax),width=.25,position=position_dodge(0.9))
```

```
admire
                                          adore
                                                              appreciate
                                                                                        build
                                                                                                                                     caress
                                                                                                                                                                                                         content
                                                                                                                                                                                                                               corrode
     1.00
                                                                                                                                                                                                         Ι
     0.75
     0.50
     0.25
     0.00
                    crash
                                         criticize
                                                                 cuddle
                                                                                      decorate
                                                                                                              delete
                                                                                                                                    despair
                                                                                                                                                        discourage
                                                                                                                                                                                 dishonor
                                                                                                                                                                                                         dismay
                                                                                                                                                                                                                              displease
     1.00
     0.75
     0.50
MeanAccuracy
0.00 -
0.00 -
0.75 -
0.75 -
                                                                                                                                                                                                                                                      BlockOrder
                                                                                                                                                                                                                                                              PV
                                                                                           hit
                                                                                                                                    impress
                                                              encourage
                                                                                                                hug
                                                                                                                                                                                                                                                               VΡ
                                                                                                                                                           I
                                                                                                                                                                                  I
     0.50
     0.25
     0.00
                                                                                                                                      slash
                                                                                                                                                          slaughter
                                                                                                              sculpt
                                                                                                                                                                                                                                twinkle
                 reassure
                                          repair
                                                                respect
                                                                                         rust
     1.00
     0.75
     0.50
     0.25
     0.00
           PhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathencePhysPs)/kathen
                                                                                                                          Task
m <- lmer(MeanAccuracy ~ BlockOrder + (1|Word), data =agr)
summary(m)
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: MeanAccuracy ~ BlockOrder + (1 | Word)
##
                 Data: agr
## REML criterion at convergence: -13380.3
##
## Scaled residuals:
##
                       Min
                                                     1Q
                                                                 Median
                                                                                                                                   Max
                                                                                                           3Q
## -2.90099 -0.52047 0.00482 0.53334 2.85461
##
## Random effects:
        Groups
                                   Name
                                                                          Variance Std.Dev.
                                       (Intercept) 0.003757 0.06129
##
           Word
           Residual
                                                                          0.003450 0.05873
##
## Number of obs: 4800, groups: Word, 40
##
## Fixed effects:
##
                                                  Estimate Std. Error
                                                                                                                                   df t value Pr(>|t|)
## (Intercept) 9.067e-01 9.765e-03 3.959e+01
                                                                                                                                                  92.85
## BlockOrderVP 2.875e-02 1.695e-03 4.759e+03
                                                                                                                                                 16.96
                                                                                                                                                                         <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
                                             (Intr)
## BlockOrdrVP -0.087
```

### PropPositive and PropPhysPsych

```
val <- d %>%
  filter(Task == "Valence") %>%
  # filter(Word %in% conc$Word) %>%
  group_by(Word,ConcValCombo) %>%
  mutate(Response.n = as.numeric(factor(Response, levels = c("negative", "positive"))) - 1) %>% # Conv
  summarize(PropPositive = mean(Response.n))
## `summarise()` has grouped output by 'Word'. You can override using the
## `.groups` argument.
  # filter(PropPositive > .1 | PropPositive < .9)</pre>
dodge = position_dodge(.9)
ggplot(data=val, aes(x=reorder(Word,PropPositive),y=PropPositive,fill=ConcValCombo)) +
  geom_bar(position=dodge,stat="identity") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
  1.00
  0.75
                                                                   ConcValCombo
PropPositive
                                                                        physical-negative
  0.50
                                                                        physical-positive
                                                                        psychological-negative
                                                                        psychological-positive
  0.25
  0.00
                      reorder(Word, PropPositive)
  # guides(fill = "none")
conc <- d %>%
  filter(Task == "PhysPsych") %>%
  # filter(Word %in% conc$Word) %>%
  group_by(Word,ConcValCombo) %>%
  mutate(Response.n = as.numeric(factor(Response, levels = c("psychological", "physical"))) - 1) %>% #
  summarize(PropPhysical = mean(Response.n))
```

```
## `summarise()` has grouped output by 'Word'. You can override using the
## `.groups` argument.
  # filter(PropPositive > .1 | PropPositive < .9)</pre>
dodge = position_dodge(.9)
ggplot(data=conc, aes(x=reorder(Word,PropPhysical),y=PropPhysical,fill=ConcValCombo)) +
  geom_bar(position=dodge,stat="identity") +
  theme(axis.text.x = element text(angle = 45, hjust = 1))
   1.00
   0.75
                                                                      ConcValCombo
PropPhysical
                                                                           physical-negative
   0.50
                                                                           physical-positive
                                                                           psychological-negative
                                                                           psychological-positive
   0.25
   0.00
                      reorder(Word, PropPhysical)
  # quides(fill = "none")
```

#### Accuracy by Participant

```
geom_errorbar(aes(ymin=YMin,ymax=YMax),width=.25,position=position_dodge(0.9)) +
theme(axis.text.x = element_text(angle = 45, hjust = 1))

Task

O.25

O.25

O.25

Walence

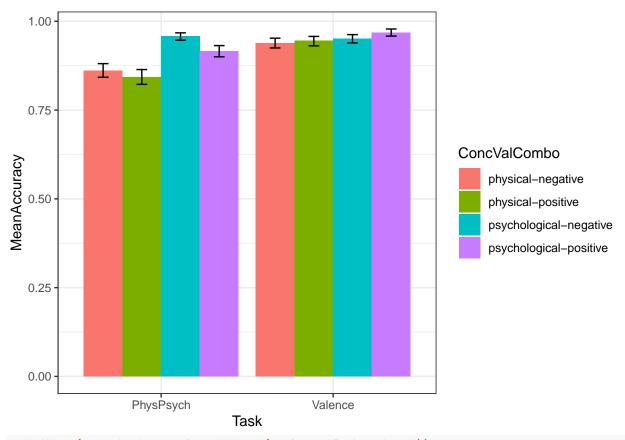
# guides(fill = "none")
```

### Mean Accuracy by ConcValCombo

```
agr <- d %>%
  group_by(Task,ConcValCombo) %>%
  summarize(MeanAccuracy = mean(Accuracy), CILow = ci.low(Accuracy), CIHigh = ci.high(Accuracy)) %>%
  mutate(YMin = MeanAccuracy - CILow, YMax = MeanAccuracy + CIHigh)

## `summarise()` has grouped output by 'Task'. You can override using the
## `.groups` argument.

dodge = position_dodge(.9)
ggplot(data=agr, aes(x=Task,y=MeanAccuracy,fill=ConcValCombo)) +
  geom_bar(position=dodge,stat="identity") +
  # facet_wrap(~Task) +
  geom_errorbar(aes(ymin=YMin,ymax=YMax),width=.25,position=position_dodge(0.9))
```



```
# theme(axis.text.x = element_text(angle = 45, hjust = 1))
# guides(fill = "none")
```