Final Report

Predicting the first four moments in particle turbulence

Introduction

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Methodology

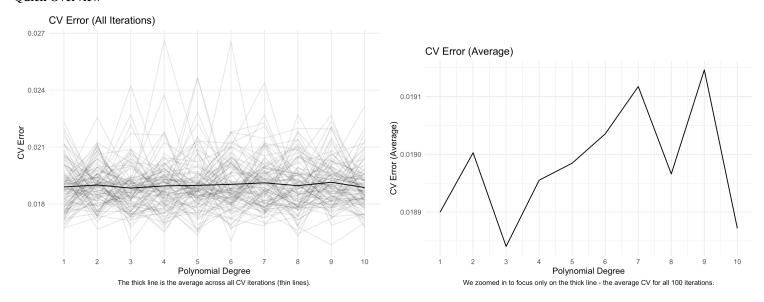
```
##
                       St
                                   Re Fr R_moment_1 R_moment_2 R_moment_3
## St
               1.00000000 -0.03169871 NaN 0.2147681 0.1479257 0.1647465
## Re
              -0.03169871 1.00000000 NaN -0.7747206 -0.3932344 -0.3844289
## Fr
                      {\tt NaN}
                                  NaN
                                        1
                                                  \mathtt{NaN}
                                                             \mathtt{NaN}
## R moment 1 0.21476813 -0.77472058 NaN 1.0000000 0.6298829 0.6217326
## R_moment_2 0.14792571 -0.39323445 NaN 0.6298829
                                                      1.0000000 0.9984335
## R moment 3 0.16474648 -0.38442895 NaN 0.6217326 0.9984335 1.0000000
## R_moment_4 0.18004537 -0.37741773 NaN 0.6150484 0.9946671 0.9988414
##
             R moment 4
              0.1800454
## St
              -0.3774177
## Re
## Fr
                     NaN
## R_moment_1 0.6150484
## R_moment_2 0.9946671
## R_moment_3 0.9988414
## R_moment_4 1.0000000
# We transform the variables using the sigmoid function so that this variable
# will be within a finite range.
train1 <- train %>%
 rename(M1 = R_moment_1, M2 = R_moment_2, M3 = R_moment_3, M4 = R_moment_4) %>%
 mutate(Fr sigmoid = 1 / (1 + exp(-Fr)),
         Re_sigmoid = 1 / (1 + \exp(-Re)),
         M1_{sigmoid} = 1 / (1 + exp(-M1)),
        M2_{sigmoid} = 1 / (1 + exp(-M2)),
        M3_{sigmoid} = 1 / (1 + exp(-M3)),
         M4_{sigmoid} = 1 / (1 + exp(-M4))) %>%
  mutate(Re_categorical = case_when(Re == 90 ~ "Low", Re == 224 ~ "Medium", Re == 398 ~ "High"),
        Fr_categorical = case_when(Fr == 0.052 ~ "Low", Fr == 0.3 ~ "Medium", Fr == Inf ~ "High"))
test1 <- test %>%
  mutate(Fr_sigmoid = 1 / (1 + exp(-Fr)),
         Re_sigmoid = 1 / (1 + \exp(-Re))) \%
 mutate(Re_categorical = case_when(Re == 90 ~ "Low", Re == 224 ~ "Medium", Re == 398 ~ "High"),
         Fr_categorical = case_when(Fr == 0.052 ~ "Low", Fr == 0.3 ~ "Medium", Fr == Inf ~ "High"))
train1
## # A tibble: 89 x 15
                  Fr
                                                          M4 Fr_sigmoid Re_sigmoid
##
         St
              Re
                                M1
                                        M2
                                                МЗ
##
      <dbl> <dbl>
                                              <dbl>
                                                                  <dbl>
                                                                             <dbl>
                    <dbl>
                             <dbl>
                                      <dbl>
                                                       <dbl>
                    0.052 0.00216 0.130
                                                     1586.
                                                                  0.513
##
   1 0.1
              224
                                           14.4
```

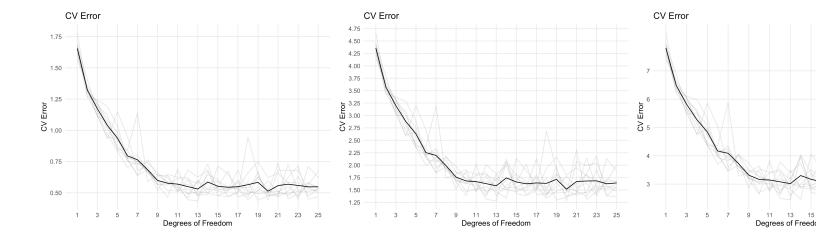
```
0.052 0.00379 0.470
##
    2
       3
              224
                                             69.9
                                                     10404
                                                                    0.513
                                                                                    1
       0.7
              224 Inf
                           0.00291
##
                                    0.0435
                                              0.822
                                                        15.6
##
       0.05
               90 Inf
                           0.0635
                                    0.0907
                                              0.467
                                                         3.27
                                                                    1
                                                                                    1
    5
       0.7
              398 Inf
                           0.000369 0.00622 0.126
                                                         2.57
                                                                    1
                                                                                    1
                           0.148
##
    6
       2
               90
                    0.3
                                    2.01
                                             36.2
                                                       672.
                                                                    0.574
                                                                                    1
##
    7
       0.2
               90 Inf
                           0.0813
                                    0.324
                                              3.04
                                                        33.0
                                                                    1
                                                                                    1
##
              224 Inf
                           0.00575
                                              2.75
                                                        63.2
    8
       3
                                    0.120
                                                                                    1
                                                                    1
##
    9
       0.9
              224 Inf
                           0.00302 0.0452
                                              0.845
                                                        15.8
                                                                    1
                                                                                    1
##
  10
      0.6
              398
                     0.052 0.000314 0.00447
                                             0.0821
                                                          1.51
                                                                    0.513
                                                                                    1
     ... with 79 more rows, and 6 more variables: M1_sigmoid <dbl>,
       M2_sigmoid <dbl>, M3_sigmoid <dbl>, M4_sigmoid <dbl>, Re_categorical <chr>,
       Fr_categorical <chr>
##
```

test1

##	# 1	A tibbl	e: 23	x 7				
##		St	Re	Fr	Fr_sigmoid	Re_sigmoid	Re_categorical	Fr_categorical
##		<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<chr></chr>	<chr></chr>
##	1	0.05	398	0.052	0.513	1	High	Low
##	2	0.2	398	0.052	0.513	1	High	Low
##	3	0.7	398	0.052	0.513	1	High	Low
##	4	1	398	0.052	0.513	1	High	Low
##	5	0.1	398	Inf	1	1	High	High
##	6	0.6	398	Inf	1	1	High	High
##	7	1	398	Inf	1	1	High	High
##	8	1.5	398	Inf	1	1	High	High
##	9	3	398	Inf	1	1	High	High
##	10	3	224	0.3	0.574	1	Medium	Medium
##	# .	wit	h 13 n	nore rows	3			

Quick Overview





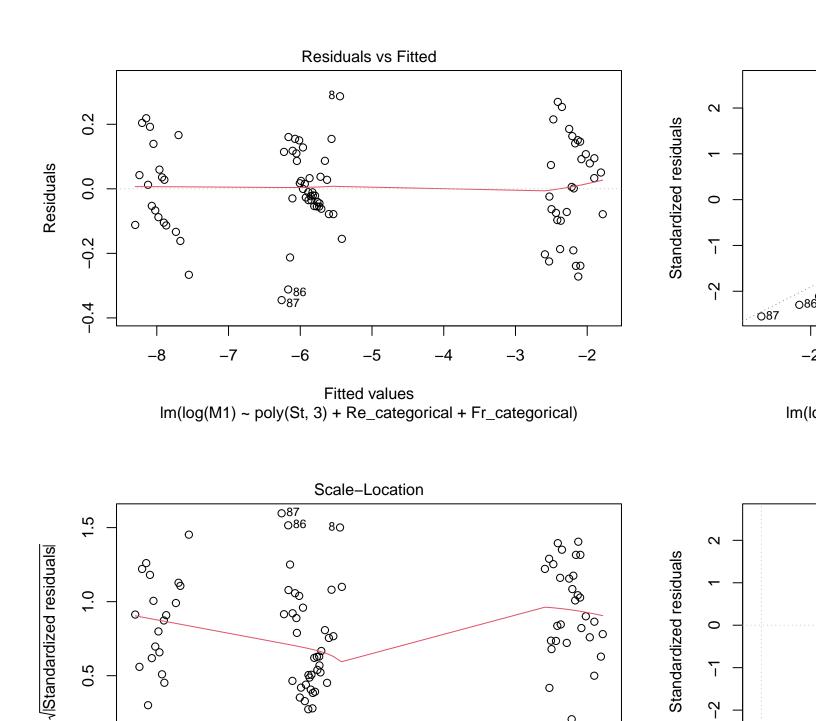
Results

Final Models

```
M1 <- lm(log(M1) ~ poly(St, 3) + Re_categorical + Fr_categorical, data = train1)
summary(M1)
```

```
##
## Call:
## lm(formula = log(M1) ~ poly(St, 3) + Re_categorical + Fr_categorical,
##
       data = train1)
##
##
  Residuals:
##
      Min
                1Q Median
                                3Q
                                       Max
##
   -0.3449 -0.0783 -0.0003 0.1073 0.2868
##
## Coefficients:
##
                       Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                        -7.98153
                                    0.03784 -210.911 < 2e-16 ***
                                              12.633 < 2e-16 ***
## poly(St, 3)1
                        1.81526
                                    0.14370
## poly(St, 3)2
                        -0.63184
                                    0.14418
                                              -4.382 3.49e-05 ***
## poly(St, 3)3
                                               1.488
                        0.21344
                                    0.14347
                                                       0.1407
                        5.76962
## Re_categoricalLow
                                    0.04355
                                            132.489
                                                      < 2e-16 ***
## Re_categoricalMedium 2.13265
                                   0.04191
                                              50.881
                                                      < 2e-16 ***
## Fr_categoricalLow
                        0.02556
                                    0.03552
                                               0.720
                                                       0.4738
## Fr_categoricalMedium -0.09182
                                    0.04155
                                              -2.210
                                                       0.0299 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1433 on 81 degrees of freedom
## Multiple R-squared: 0.9962, Adjusted R-squared: 0.9959
## F-statistic: 3054 on 7 and 81 DF, p-value: < 2.2e-16
```

plot(M1)



Fitted values Im(log(M1) ~ poly(St, 3) + Re_categorical + Fr_categorical)

-5

0

-6

 $M2 \leftarrow gam(log(M2) \sim s(St, df = 9) + Re_categorical + Fr_categorical +$ Re_categorical * Fr_categorical, data = train1) summary(M2)

-4

0

-3

0 0

-2

7

က

0.00

C

lm(lo

0

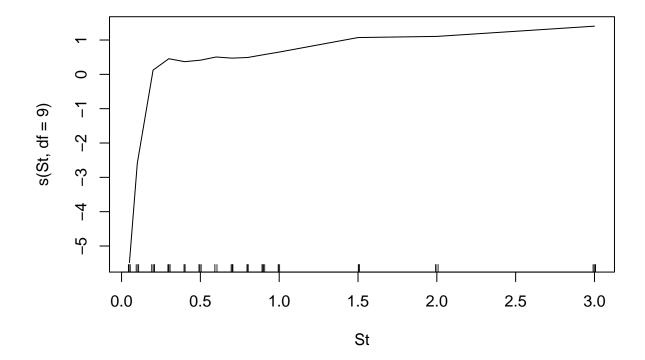
-8

-7

0.0

```
## Call: gam(formula = log(M2) ~ s(St, df = 9) + Re_categorical + Fr_categorical +
##
      Re_categorical * Fr_categorical, data = train1)
## Deviance Residuals:
        Min
              1Q
                         Median
                                                Max
## -3.100803 -0.185164 0.006113 0.231614 1.735705
##
## (Dispersion Parameter for gaussian family taken to be 0.5064)
##
##
       Null Deviance: 1212.424 on 88 degrees of freedom
## Residual Deviance: 36.461 on 72.0001 degrees of freedom
## AIC: 209.1479
## Number of Local Scoring Iterations: NA
##
## Anova for Parametric Effects
##
                                Df Sum Sq Mean Sq F value
                                                             Pr(>F)
## s(St, df = 9)
                                            49.47 97.699 4.819e-15 ***
                                 1 49.47
                                 2 722.42 361.21 713.289 < 2.2e-16 ***
## Re_categorical
## Fr categorical
                                 2 205.08 102.54 202.487 < 2.2e-16 ***
## Re_categorical:Fr_categorical 3 151.24
                                           50.41 99.550 < 2.2e-16 ***
## Residuals
                                72 36.46
                                             0.51
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Anova for Nonparametric Effects
##
                                Npar Df Npar F
                                                   Pr(F)
## (Intercept)
                                      8 20.692 6.661e-16 ***
## s(St, df = 9)
## Re_categorical
## Fr_categorical
## Re_categorical:Fr_categorical
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
plot(M2)
## Warning in preplot.Gam(x, terms = terms): No terms saved for "a:b" style
## interaction terms
## Warning in gplot.default(x = c("Medium", "Medium", "Medium", "Low", "High", :
## The "x" component of "partial for Re_categorical" has class "character"; no
## gplot() methods available
## Warning in gplot.default(x = c("Low", "Low", "High", "High", "High", "Medium", :
## The "x" component of "partial for Fr_categorical" has class "character"; no
## gplot() methods available
```

```
M3 <- gam(log(M3) ~ s(St, df = 9) + Re_categorical + Fr_categorical +
                   Re_categorical * Fr_categorical, data = train1)
summary(M3)
##
## Call: gam(formula = log(M3) ~ s(St, df = 9) + Re_categorical + Fr_categorical +
       Re_categorical * Fr_categorical, data = train1)
##
   Deviance Residuals:
##
##
        Min
                  1Q
                       Median
                                     30
                                             Max
##
   -5.97775 -0.35007 0.03625 0.34668
                                        3.06602
##
##
   (Dispersion Parameter for gaussian family taken to be 1.4732)
##
##
       Null Deviance: 2837.669 on 88 degrees of freedom
## Residual Deviance: 106.0723 on 72.0001 degrees of freedom
  AIC: 304.189
##
## Number of Local Scoring Iterations: NA
##
## Anova for Parametric Effects
##
                                 Df Sum Sq Mean Sq F value
                                                                Pr(>F)
## s(St, df = 9)
                                               94.67 64.263 1.429e-11 ***
                                       94.67
                                   1
## Re_categorical
                                  2 1162.83
                                              581.42 394.656 < 2.2e-16 ***
                                              399.24 270.994 < 2.2e-16 ***
## Fr_categorical
                                   2
                                     798.47
## Re_categorical:Fr_categorical 3
                                     542.39
                                              180.80 122.722 < 2.2e-16 ***
## Residuals
                                 72
                                     106.07
                                                1.47
##
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Anova for Nonparametric Effects
##
                                 Npar Df Npar F
                                                     Pr(F)
## (Intercept)
```



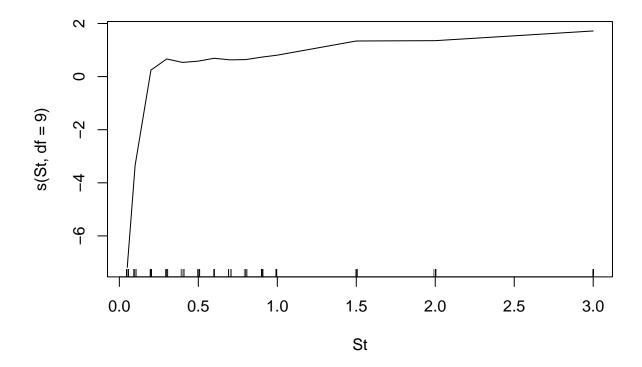
gplot() methods available

```
M4 <- gam(log(M4) ~ s(St, df = 9) + Re_categorical + Fr_categorical + Re_categorical * Fr_categorical, data = train1)
summary(M4)
```

```
##
## Call: gam(formula = log(M4) ~ s(St, df = 9) + Re_categorical + Fr_categorical +
## Re_categorical * Fr_categorical, data = train1)
## Deviance Residuals:
## Min 1Q Median 3Q Max
```

```
## -8.61481 -0.44595 0.04067 0.50573 4.15489
##
## (Dispersion Parameter for gaussian family taken to be 2.7642)
      Null Deviance: 5248.685 on 88 degrees of freedom
##
## Residual Deviance: 199.0259 on 72.0001 degrees of freedom
## AIC: 360.1979
## Number of Local Scoring Iterations: NA
##
## Anova for Parametric Effects
##
                                Df Sum Sq Mean Sq F value
## s(St, df = 9)
                                 1 146.72 146.72 53.076 3.281e-10 ***
## Re_categorical
                                 2 1732.68 866.34 313.410 < 2.2e-16 ***
                                 2 1769.38 884.69 320.048 < 2.2e-16 ***
## Fr_categorical
## Re_categorical:Fr_categorical 3 1162.78 387.59 140.217 < 2.2e-16 ***
## Residuals
                                72 199.03
                                              2.76
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Anova for Nonparametric Effects
##
                                Npar Df Npar F
                                                   Pr(F)
## (Intercept)
## s(St, df = 9)
                                      8 16.705 1.025e-13 ***
## Re_categorical
## Fr_categorical
## Re_categorical:Fr_categorical
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
plot(M4)
## Warning in preplot.Gam(x, terms = terms): No terms saved for "a:b" style
## interaction terms
## Warning in gplot.default(x = c("Medium", "Medium", "Medium", "Low", "High", :
## The "x" component of "partial for Re_categorical" has class "character"; no
## gplot() methods available
## Warning in gplot.default(x = c("Low", "Low", "High", "High", "High", "Medium", :
## The "x" component of "partial for Fr_categorical" has class "character"; no
```

gplot() methods available



Conclusion