# Journal of Science & Cycling Breakthroughs in Cycling & Triathlon Sciences



1 Article

#### Mapping whole-event drive losses: the impact of race 2 profile and rider input on transmission efficiency in 3 cycling 4

# 5

## George C Barnaby<sup>\*</sup>, Stuart Burgess and Jason Yon

Department of Mechanical Engineering, Queen's Building, University of Bristol. BS8 1TR. UK

\* Correspondence: (GB) george.barnaby@bristol.ac.uk

Received: date; Accepted: date; Published: date

8

6

7

9 Abstract: Several studies have considered the factors influencing transmission efficiency in a 10 bicycle. These conclude that the number of teeth in sprockets which are engaged with the chain 11 and the torque and cadence of the cyclist influence the frictional losses associated with 12 transmission between rider and rear wheel. These parameters may vary significantly during a 13 bicycle race since a rider modifies gear, power, and cadence to maximise physiological efficiency 14 for optimum bicycle velocity. Furthermore, gearing selection and power input varies between 15 riders, riding group and course profile. However, power models used to estimate race outcomes 16 tend to simplify efficiency to a single, arbitrary factor, describing losses which scale linearly with 17 input power regardless of expected regime. This study extends existing analytical descriptions 18 of transmission losses to the context of a road bicycle with front and rear derailleurs. The 19 calculated efficiency is considered within a cycling model to judge different regimes under 20 which the chain will typically operate and maps overall performance during an event. Efficiency 21 may vary significantly under certain loading regimes shown. In the context of highly trained 22 cyclists these differences result in small, linearly varying changes about a mean value. This 23 study shows there is limited error in assuming constant efficiency for certain race types, though 24 the efficiency value itself is dependent on several factors affecting the average loading regime. 25 Elevation profile of the racecourse and average power input from the rider are key parameters 26 affecting average efficiency. More massive riders racing at high average power input will 27 experience higher efficiency, while efficiency is higher across all riders racing courses with 28 increased elevation gain.

29 Keywords: transmission; efficiency; model; losses; bicycle; derailleur.

30

#### 31 1. Introduction

32 In cycling, the use of analytical models 33 to describe the balance of input power at the 34 crank and output power at the tyre-road 35 interface allows the engineer to identify areas 36 for improvement in rider technique or 37 equipment design. One such model is 38 described in equation (1), based on an 39 analytical model from Martin et al, 1998.

$$P_{in} = V(F_a + F_r + F_g)/\eta, \qquad \text{Eq. (1)}$$

40 where  $P_{in}$  is the input power of the rider;  $F_a$ , 41  $F_r$  and  $F_g$  are the resistive forces associated 42 with aerodynamic drag, rolling resistance 43 between tyre and road, and gravitational 44 resistance; *V* is the bicycle velocity; and  $\eta$  is 45 the transmission efficiency. 46 In deriving this model, and commonly

in literature, transmission efficiency is 47 48 assumed to be a constant value such that 49 losses scale linearly with power input, and 50 often an arbitrary estimate. Later studies, 51 however, demonstrate that the same chain in derailleur 52 а transmission system has



© 2020 fist author, licensee JSC. This is an Open Access article distributed under the terms of the Creative Commons Attribution License ((http://creativecommons.org/licenses/by/4.0/) which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.



53 measured efficiency in the range 80.9 – 98.6% 54 (Spicer et al., 2001). The changing factors 55 causing this range in efficiency are input 56 power, rotational speed, and gear 57 configuration, which may also vary greatly 58 during a bicycle race: gear shifts and non-59 consistent physiological output from the 60 rider are common consequences of varying 61 road-race profiles.

62 There is a gap in published literature for 63 a holistic consideration of the efficiency to 64 study these dependencies in the context of 65 different racecourses and riders, which may 66 be useful in determining the error in 67 assuming constant efficiency and providing 68 recommendations for what efficiency 69 estimate to use for riders and engineers based 70 on course and rider profile.

71 This study seeks to investigate the 72 variability of transmission efficiency in 73 expected regimes and defines the key factors 74 influencing the transmission efficiency in 75 usable terms, such that riders and engineers 76 might be better informed in their use of an 77 estimated efficiency in future modelling.

#### 78 2. Frictional loss model

79 The authors are unaware of а 80 comprehensive model of frictional losses in a 81 bicycle derailleur drive in literature, and so 82 have derived an analytical approach. This is 83 an extension of the work of Lodge & Burgess,

84 2001, as is used in Barnaby et al., 2020.



85

86 Figure 1. Sources of friction in a bicycle 87 transmission, including rolling element 88 bearings and points of chain articulation 89 (numbered).

90 To determine the relative contribution of 91 different sources of friction, Lodge & 92 Burgess's analysis is used in conjunction with 93 the geometry and spring rate of the rear 94 derailleur to predict bottom-span tension,

- 95 and an industrial model of bearing losses is
- 96 used to estimate friction in bearings (The SKF
- 97 model for calculating the frictional moment).
- 98 The relative losses of each of the sources

99 of friction, shown in Figure 1, is summarised 100 in Table 1. There is significant contribution to 101 losses of the high-tension span articulations 102 (71%), reduced contribution from low-103 tension span articulations (24%), and a near-104 negligible contribution from rolling element 105 bearings (3%). Friction in rolling element 106 bearings is henceforth neglected in this 107 analysis.

108 Table 1. Power losses are approximated for 109 different sources of friction in the drive 110 (300W / 90rpm)

	Power loss [W]	% of total
High-tension span <sup>1</sup>	5.5	71
Low-tension span <sup>2</sup>	2.0	26
Rolling element bearings	0.2	3
Total	7.7	100

111 <sup>1</sup> Chain links 1-2; <sup>2</sup> Chain links 3-8; <sup>3</sup> pulley

112 wheels, bottom bracket, rear hub, pedals

113 Transmission efficiency is defined as in 114 equation 2:

$$\eta = \left(P_{in} - N_s \omega_s \sum_{i=1}^8 W_i\right) / P_{in}, \quad \text{Eq. (2)}$$

115 where *W* is the work done against friction in 116 each of 8 articulating links (entry and exit to 117 each sprocket),  $\omega_s$  is the rotational frequency 118 of the chainring  $(s^{-1})$  and  $N_s$  is number of teeth 119 in the chainring. Work done against friction 120 is a function of chain geometry, articulation 121 angle and chain tension, all of which may be 122 calculated based on specific equipment and 123 rider input. A further dependency is on 124 coefficient of sliding friction within the chain 125 links, which can be accurately determined 126 experimentally using techniques such as 127 those proposed by Wragge-Morley et al., 128 2017. The calculation for work done against 129 friction is included in Appendix I.

#### 130 2.1 Transmission efficiency variation

131 The variation of transmission efficiency

- 132 is examined over a range of cycling torque
- 133 inputs and riding gears, shown in Figure 2. 134
- The low, hill climbing gears offer higher

135 efficiency due to the reduced articulation 136 angle. Positive correlation between input 137 torque at the crank and efficiency is due to 138 the relative reduction of significance of the 139 bottom-span losses, which are independent 140 of torque input. At low torque the torque-141 independent losses in the bottom-span, 142 tensioned by the derailleur arm, are 143 relatively more significant and so 144 transmission efficiency changes rapidly as a 145 function of torque.



146

Figure 2. Power efficiency [%] contour map
for varying rider torque and gear for 11-28
tooth cassette sprockets engaged with (a) 39tooth chainring; and (b) 53-tooth chainring.

#### 151 3. Variable efficiency within power model

152 To model how transmission efficiency 153 varies in a race, simulation of typical power 154 input and race profile is necessary since 155 efficiency depends on power and gear 156 selection, themselves having multivariant 157 dependencies. The power required to 158 overcome resistance at steady speed cycling 159 is given in equations (3) - (6), based on work 160 by Martin et al., 1998.

$$P_{in} = \left(P_a + P_r + P_g\right)/\eta, \quad \text{Eq. (3)}$$

161 where power to overcome aerodynamic 162 drag,  $P_a$ , is described in equation (4), power 163 to overcome rolling resistance of the tyres,  $P_r$ , 164 is described in equation (5) and power to 165 overcome gradient,  $P_g$  is described in 166 equation (6).

$$P_a = 0.5 \rho C_d A_f V^3$$
, Eq. (4)

167 where  $\rho$  is air density,  $C_d$  is coefficient of 168 aerodynamic drag,  $A_f$  is the frontal area of 169 bicycle and rider, and *V* is bicycle velocity. 170 Note that wind velocity is assumed to be zero 171 in this analysis.

$$P_r = mgC_{rr}V, \qquad \text{Eq. (5)}$$

172 where *m* is total mass of rider and bicycle, *g* 173 is the gravitational acceleration constant and 174  $C_{rr}$  is the coefficient of rolling friction 175 between tyre and road surface. Upright, 176 straight-line cycling is considered for this 177 analysis.

$$P_g = mg \sin(\theta) V$$
, Eq. (6)

178 where  $\theta$  is the angle of gradient. Typical 179 values for variables shown in equations (4) – 180 (6) are from Wilson, Papadopoulos, and 181 Whitt, 2004.

182 Steady-state velocity is calculated at183 many discrete points along a simulated184 racecourse for a typical bicycle drivetrain.

185 Gearing is selected to maintain cadence 186 within a typical range, with chosen gearing 187 influencing the calculation for efficiency 188 according to the described frictional loss 189 model. Further, a variable power input is 190 applied such that power increases with 191 positive gradient and decreases with 192 negative gradient, shown to be an effective 193 pacing strategy (Wells & Marwood, 2016).

#### 194 3.1 Efficiency variation by rider type

195 Transmission efficiency during an
196 example elite race (part of the UCI 2021
197 World Road Championships road-race from
198 Antwerp to Leuven) is simulated and
199 illustrated for four different riders in Figure
200 3, where input parameters are summarised in
201 Table 2. The spread of efficiency estimates





204 205

206

207

208

209

210

Figure 3. Transmission efficiency overlayed on Leuven 2021 road-race course profile for a (a) elite male cyclist,
(b) elite female cyclist, (c) untrained male cyclist, and (d) untrained female cyclist.



Figure 4. Histogram of simulated
transmission efficiency during example race
for elite and untrained riders.

215 Efficiency can be seen to fluctuate with 216 the gradient of the course due to the changing 217 gear and power. Hill climbing gears and a 218 marginal increase in power both result in 219 increased efficiency as has been shown 220 previously. The opposite is true for negative 221 gradients, where smaller sprocket is engaged 222 and power is slightly reduced, decreasing 223 efficiency.

Table 2. Input parameters for four modelled
cases, with estimates for elite and untrained
male and female riders.

	Elite		Untrained	
	Male	Femal e	Male	Femal e
Mass [kg]	70	60	80	65
C <sub>d</sub> A [m <sup>2</sup> ]	0.3	0.25	0.4	0.3
Average power [W]	350	250	150	100
Cadence	90±1	90±1	70±1	70±1
[rpm]	0	0	5	5
Average				
efficienc	98.0	97.9	97.8	97.6
у (S.D.) [%]	(0.20)	(0.24)	(0.33)	(0.46)

227 Comparing the proficient and untrained 228 cases, a larger variance can be seen in the 229 untrained cyclist as well as a slightly lower 230 average efficiency. This is illustrated more 231 clearly in Figure 4. The lower average 232 efficiency is largely due to the reduced power 233 input, and hence lower average torque. The 234 variance is reduced in the trained cyclist due 235 to the responsive gear changes working to 236 maintain a high cadence.

## 237 3.2 Efficiency variation in elite riders

238 In elite level racing, efficiency variance 239 during a race is low and there is little error in 240 determining average velocity, or time to 241 completion, by using a single value efficiency 242 across an entire race. This is determined in 243 simulated races by finding the ratio of total 244 energy input and total energy output, found 245 by integrating the power output and input 246 with respect to time as in equation 7.

$$\eta = \frac{\int P_{in} - P_{lost} dt}{\int P_{in} dt}, \qquad \text{Eq. (7)}$$

247 However, there is still dependency of 248 this average efficiency on power input and 249 gearing, which itself is dictated by the 250 elevation profile of a racecourse.

251 An effective average efficiency is 252 determined numerically by simulating racing 253 across 20 different grand tour events with 254 riders of varying input power (50-550W) and 255 mass (50-80kg). Dependency of average 256 efficiency on the climbing during the race 257 (measured as average metres elevation gain 258 per kilometre travel), and the average power 259 achieved by the rider during the race is 260 illustrated in Figure 5. Mass is less impactful 261 and can be accounted for by applying an 262 additional 0.1% efficiency per 20kg above 263 65kg. These results are valid for riders with a 264 power-to-weight ratio of between 2 and 6 265 W/kg.



266

Figure 5. Contour map of transmission
power efficiency [%] as function of average
power during a race and its elevation profile.

## 270 4. Discussion

271 The range in efficiency found in 272 previous research is not realised in loading 273 regimes typical in elite racing. This is largely 274 because of the narrow cadence range and 275 high torque in elite racing which leads to 276 small and linearly varying changes in 277 transmission efficiency. Provided this is 278 maintained during a race, there is little error 279 in using a single factor for efficiency. 280 However, average power and elevation

281 profile are two factors which can vary greatly282 in elite cycling between different event styles

202 In this cycling between unterent event style.

283 and rider physiologies, leading to consistent

284 changes to efficiency across a race. A 285 mountainous tour stage will see higher 286 efficiency than one which is flat by up to 287 0.5%-pts, which may be even more extreme if 288 considering specific hill climbing events. 289 Rider power input also will influence 290 average efficiency, meaning that more 291 powerful male riders racing TT courses at 292 maximal effort may experience an average 293 transmission efficiency up to 0.8%-pts higher 294 than a less powerful rider during an 295 endurance event. Female elite riders will 296 inherently experience reduced а 297 transmission efficiency than male elite riders 298 due to applying less power at the crank.

### 299 5. Practical Applications.

300 Transmission efficiency may usually be 301 experimentally examined in limited and 302 specific loading regimes. This gives limited 303 insight given the dependencies of efficiency 304 which vary during a bicycle race. This study 305 demonstrates that further applying the 306 results of such tests to contextualise 307 efficiency within the expected loading 308 regimes based on rider and course type may 309 offer additional accuracy in determining an 310 effective average efficiency. This may be 311 applied to future analytical modelling for 312 evaluating equipment upgrades or 313 determining pacing strategies.

Future research to further examine
influences on transmission efficiency is
needed to confirm the theory presented here,
including extensive practical testing which
may offer experimental validation.

319 Funding: This research was funded by320 Engineering and Physical Sciences Research321 Council and Renold Chain.

322 **Conflicts of Interest:** The authors declare no 323 conflict of interest. The funders had no role in the 324 design of the study; in the collection, analyses, or 325 interpretation of data; in the writing of the 326 manuscript, or in the decision to publish the 327 results.

## 328 References

329 1. Barnaby, G. C., Yon, J., & Burgess, S. (2020).

330 Sprocket Size Optimisation for Derailleur

- 331 Racing Bicycles. Journal of Science and
- 332 Cycling, 9(2), 36.

- 333 2. Lodge, C. J., & Burgess, S. C. (2001). A model 334 of the tension and transmission efficiency of 335 a bush roller chain. Proceedings of the 336 Institution of Mechanical Engineers, Part C: 337 Journal of Mechanical Engineering 338 Science, 216(4), 385-394.
- 339 3. Martin, J. C., Milliken, D. L., Cobb, J. E.,
  340 McFadden, K. L., & Coggan, A. R. (1998).
  341 Validation of a mathematical model for road
  342 cycling power. Journal of applied
  343 biomechanics, 14(3), 276-291.
- 344 4. Spicer, J. B., Richardson, C. J., Ehrlich, M. J.,
  345 Bernstein, J. R., Fukuda, M., & Terada, M.
  346 (2001). Effects of frictional loss on bicycle
  347 chain drive efficiency. J. Mech. Des., 123(4),
  348 598-605.
- 3495.The SKF model for calculating the frictional350moment (n.d.).Retrieved from351https://www.skf.com/binaries/pub12/Image352s/0901d1968065e9e7-The-SKF-model-for-
- 353 calculating-the-frictional-moment\_tcm\_12-354 299767.pdf
- Wells, M. S., & Marwood, S. (2016). Effects of
  power variation on cycle performance
  during simulated hilly time-trials. European
  journal of sport science, 16(8), 912-918.
- Wilson, D. G., Papadopoulos, J., & Whitt, F.
  R. (2004). Bicycling science (3rd ed.). MIT
  press.
- 362 8. Wragge-Morley, R., Yon, J., Lock, R.,
  363 Alexander, B., & Burgess, S. (2018). A novel
  364 pendulum test for measuring roller chain
  365 efficiency. Measurement Science and
  366 Technology, 29(7), 075008.
- 367

# 368 Appendix I

- 369 Work done against friction in 370 articulating chain links as derived by Lodge
- 371 & Burgess, 2001, is summarised here. Eight
- 372 articulations are considered, at entry to and
- 373 exit from each engaged sprocket. The work

- 374 done in articulating these links
- 375 simultaneously represents the energy lost for
- 376 the drive advancing by one link, which may
- 377 be multiplied by the chain speed in link pitch
- 378 per second to determine the power lost here.

#### 379 Work done in articulating chain links

- Articulation of inner and outer links
  results in relative sliding of different surfaces
  between the pin and bushing. Since they
  must alternate, an average is taken of the two
- 384 to define work done for one articulation:

$$W = (W_{pin} + W_{bush})/2$$

386 where work done during pin articulation is387 described as below:

$$W_{pin} = \frac{F_i}{\sqrt{1+\mu^2}} \mu r_{bi} \alpha_i$$

389 and work done during bush articulation is:  
300 
$$\mu F_c r_{bi} [\cos \theta_{RA} - \cos(\theta_{RA} + \alpha_m)]$$

$$W_{bush} = \frac{\mu F_c T_{bi} \cos \theta_{RA} - \cos(\theta_{RA} + \alpha_m)}{\sqrt{1 + \mu^2} \sin(\theta_{RA} + \alpha_m)}$$

$$H = \frac{\mu F_c T_{bo} (1 - \cos \alpha_m)}{\sin(\theta_{RA} + \alpha_m)}$$

392 Lodge & Burgess, 2001 should be consulted 393 for definitions of these terms. These are 394 determined by the geometry of the chain 395 components and sprocket, except for 396 coefficient of friction,  $\mu$ , which is determined 397 using accurate measurements as described in 398 Wragge-Morley et al., 2017.

# 399 Chain tension force

400 The contact force between each of 8 401 articulating links is calculated. Top span 402 contact force is from crank torque acting at 403 chainring radius and acts at articulations 404 onto the chainring and off the engaged 405 cassette sprocket. Bottom span contact force 406 is equal for all 6 remaining articulations and 407 resolved from the spring rate of the rear 408 derailleur tension arm and its geometry.