## Analysis of Weight Loss Data

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### 1 Description of Data and Research question

The primary outcome of the weight loss data set is the 12 month weight loss (in kg) [wtch]. The factors measured on each individual are treatment (1=control, 2/3=intervention), age, race (1=Black, 2=Hispanic, 3=White, 4=Other) and a biomarker for weight loss.

The objective of this analysis is to determine the factors related to 12 month weight loss and to determine whether an intervention was effective in increasing weight loss.

## 2 Exploratory Data Analysis

Histograms of wtch, age, biomarker and barplot of race are shown in Figure 1. The histogram wtch shows a slight skew. A qqplot of wtch shown in Figure 2 shows that the normality assumption is valid. A Shapiro-Wilk normality test also shows that the data is normal with a p-value of 0.03.

Histogram of age shows that all the subjects in the study are aged over 50. The distribution of bio-marker looks symmetric. Age and biomarker are positively correlated with a correlation of .68.

The barplot of race shows that most of the subjects in the study are White. There are also significant number of Blacks in the study. The number of Hispanic and other groups is very low.

A boxplot of weight loss by race in Figure 3 shows that the weight loss for White and Black population is different. The mean weight loss for Blacks is 6.186069 and for Whites is 10.878564. A boxplot of weight loss by intervention (Figure 4) shows similar distribution of weight loss for control and intervention groups. The mean weight loss for the control group is 9.340884 and for interventions 2 and 3 it is 10.291749 and 10.042560. Hypothesis tests to determine the effectiveness of weight loss interventions will be performed in the next section.

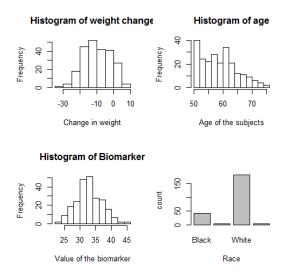


Figure 1: Histograms of weight change, age and biomarker and boxplot of race

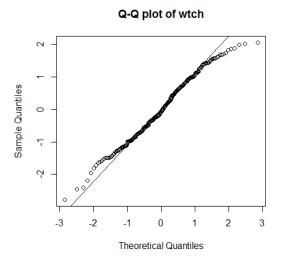


Figure 2: Q-Q plot of wtch

# 3 Frequentist Analysis

Fitting a full regression line for the data gives the following:

```
lm(formula = wtch ~ age + biomarker + as.factor(race) + as.factor(treatment),
    data = data)
```

### Weight change by Race

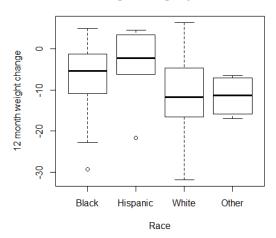


Figure 3: Weight change distribution by race

### Weight change by Treatment

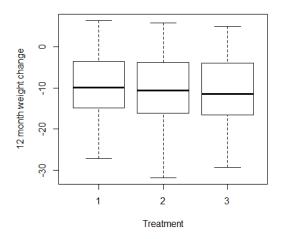


Figure 4: Weight change distribution by intervention

#### Residuals:

Min 1Q Median 3Q Max
-21.3252 -5.3880 -0.2753 5.7136 19.2159

Coefficients:

Estimate Std. Error t value Pr(>|t|)

(Intercept) -18.29184 4.99303 -3.663 0.00031 \*\*\*

age	0.17942	0.11277	1.591	0.11300	
biomarker	0.09319	0.16448	0.567	0.57158	
as.factor(race)2	0.47591	3.63383	0.131	0.89592	
as.factor(race)3	-5.46219	1.33050	-4.105	5.64e-05	***
as.factor(race)4	-5.67994	3.63573	-1.562	0.11963	
as.factor(treatment)2	-1.12395	1.21312	-0.926	0.35518	
as.factor(treatment)3	-1.79467	1.27184	-1.411	0.15960	

Residual standard error: 7.614 on 226 degrees of freedom Multiple R-squared: 0.09939, Adjusted R-squared: 0.07149 F-statistic: 3.563 on 7 and 226 DF, p-value: 0.001179

The p-values indicate that treatment is not significant. A 2-sample t-test can be performed to assess the effectiveness of the interventions. If the mean weight loss for the control group is  $\mu_1$  and for the interventions is  $\mu_2$  and  $\mu_3$ , the hypothesis being tested is as follows:

$$H_0: \mu_2 = \mu_1$$
  
 $H_a: \mu_2 \neq \mu_1$  (1)

The t-statistic for this test is 0.7513 with a p-value of 0.4536. So, the mean weight loss for intervention 1 was equal to the mean weight loss for the control group. Similarly, a t-test for intervention 2 gives a t-statistic of 0.5553 with a p-value of 0.5795. Both the interventions did not seem to effect a greater weight loss than in the control group.

The summary of full regression line shows that only race may be important. Even in race, only whether or not the subject is White matters. Let the binary dummy variable representing this be called dum2. Several reduced models were considered and the values adjusted- $R^2$  and BIC are tabulated in table 4

Model	BIC	Adjusted- $R^2$
Wtch~treatment+biomarker+race+age	1655.04	0.071
Wtch~ biomarker+race+age	1646.25	0.071
Wtch~ race+ biomarker	1642.97	0.066
Wtch~ race+age	1641.19	0.073
Wtch~dum2+age	1632.75	0.072
Wtch~age	1642.63	0.013

The model with Wtch $\sim$ dum2+age has the lowest BIC and second highest Adjusted- $R^2$ . This seems to be the best model. A summary of the fit is given below:

lm(formula = wtch ~ dum2 + age, data = data)
Residuals:

```
Min 1Q Median 3Q Max -21.2580 -5.4101 -0.2171 6.1037 18.5292 Coefficients:
```

Residual standard error: 7.612 on 231 degrees of freedom Multiple R-squared: 0.08001, Adjusted R-squared: 0.07204 F-statistic: 10.04 on 2 and 231 DF, p-value: 6.564e-05

An F-test for the remaining variables using anova shows that they are not significant (p-value of 0.43). The residual plot for this model in Figure 5 shows no pattern and most values are between (-1.5, 1.5). The q-q plot looks roughly linear. Also, a plot of Cook's distance in Figure 6 shows that there are no influential points. So, this model seems to fit alright, even if it explains only 7% of the variance in the given data.

#### Residual plot of wtch~dum2+age

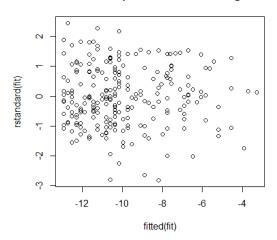


Figure 5: Distribution of standardized residuals against fitted values

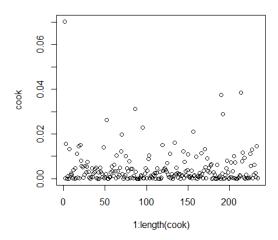


Figure 6: Cook's distance for various points

## 4 Bayesian Analysis

A full model with normal likelihood and diffuse normal priors for the betas and gamma prior for  $\tau$  is shown below:

$$Wtch \sim N(X\beta, \tau)$$
  

$$\beta_i \sim N(0, .0001)$$
  

$$\tau \sim Gamma(.001, .001)$$
(2)

An MCMC algorithm with three chains for each parameter (with burn-in of 1000 and thinning of 10, so 900 samples for each chain)was implemented using WinBUGS. The statistics for the full model are shown in Figure 7. The 95% confidence interval for  $\beta_1$  and  $\beta_3$  does not include a zero. So, only dum2 (of race) seems to be significant. The DIC and LPML values for various reduced models is shown in table below:

Model	DIC	LPML
Wtch~treatment+biomarker+race+age	1624.3	-812.3355
Wtch~ biomarker+race+age	1622.1	-811.2601
Wtch~ race+ biomarker	1622.4	-811.2701
Wtch~ race+age	1620.6	-810.5863
Wtch~dum2+age	1619.0	-809.4277
Wtch~dum2+age+dum2*age	1620.8	-810.3832

Again, the model with Wtch~dum2+age has the lowest DIC and LPML. The statistics for this model are shown below

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¥			Node	e statistic	CS				
node	mean	sd	MC error	2.5%	median	97.5%	start	sample	^
beta[1]	-18.24	4.988	0.03112	-27.98	-18.28	-8.414	1000	27003	
beta[2]	0.5075	3.636	0.02148	-6.76	0.5225	7.675	1000	27003	
beta[3]	-5.457	1.339	0.007708	-8.094	-5.452	-2.846	1000	27003	
beta[4]	-5.705	3.64	0.0222	-12.88	-5.691	1.409	1000	27003	
beta[5]	0.1791	0.1134	6.587E-4	-0.04436	0.1788	0.399	1000	27003	
beta[6]	-1.134	1.219	0.007538	-3.515	-1.125	1.241	1000	27003	
beta[7]	-1.805	1.283	0.007494	-4.319	-1.803	0.7143	1000	27003	
beta[8]	0.09213	0.1662	9.749E-4	-0.2358	0.09284	0.4179	1000	27003	~

Figure 7: Model Statistics

Node statistics

```
node mean sd MC error 2.5% median 97.5% start sample beta[1] -18.57 4.901 0.11 -28.15 -18.53 -8.984 101 2700 beta[2] -4.743 1.211 0.0208 -7.134 -4.739 -2.37 101 2700 beta[3] 0.2085 0.082 0.0018 0.05 0.207 0.371 101 2700 tau 0.01725 0.0016 2.7E-5 0.014 0.0172 0.0205 101 2700
```

This model also converges, as can be seen from the trace plots in Figures 8 and 9. The absolute fit of the model can be assessed from the posterior predictive probability. The  $\chi^2$  statistic chosen is:

$$T(y,\theta) = \frac{\sum_{i} (y_i - E(y_i|\theta))^2}{Var(y_i|\theta)}$$
(3)

The posterior predictive probability of  $P(T(y^{rep}, \theta) \ge T(y, \theta)|y) = 0.79$ . This indicates that the model does not fit very well. Models with quadratic interaction (trt\*age, dum2\*age,  $age^2$ ,etc) with the outcome were also considered in both frequentist and Bayesian analysis, but were not found to be better than the above model.

#### 5 Conclusion

In both frequentist and Bayesian analysis, the factors that are important for weight loss are age and race. The weight loss interventions are not significant as shown by the final models and hypothesis tests. The best model in frequentist analysis only explained about 7% of the variability in the data but the residual plots do not indicate a lack of fit. The best model in Bayesian analysis showed good convergence but the posterior predictive checks were not very promising. There was not a significant difference between frequentist and Bayesian methods as we did not have informative priors. But, in general Bayesian methods offer more flexibility in statistical analysis.

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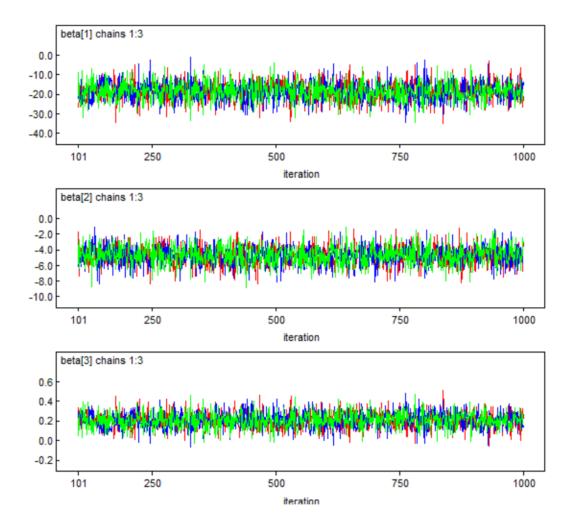


Figure 8: Model Convergence

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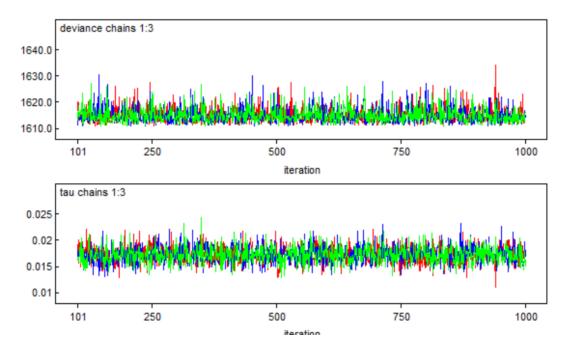


Figure 9: Model Convergence