### 1 Bin packing.

Bin packing. This is a classical application of McDiarmid's inequality. Let  $X_1, \ldots, X_n$  be i.i.d. random variables taking values in [0,1]. Each  $X_i$  is the size of a package to be shipped. The packages are shipped in bins of size 1, so each bin can hold any set of packages whose sizes sum to at most 1. Let  $B_n = f(X_1, \ldots, X_n)$  be the minimal number of bins needed to ship the packages with sizes  $X_1, \ldots, X_n$ . Computing  $B_n$  is a hard combinatorial optimization problem; however, we can say something about its mean and tail behavior.

### 1.1

Let  $\mu$  be the common mean of the  $X_i$ 's. Show that  $\mathbb{E}[B_n] \geq n\mu$ .

#### 1.2

Prove that for any  $\epsilon > 0$ ,

$$\mathbb{P}\left(n^{-1}B_n \le \mu - \epsilon\right) \le \exp\left(-2n\epsilon^2\right).$$

## 2 Orthogonal design lasso bits

Let  $y_i = x_i^{\top} \beta^* + \eta_i$  for i = 1, ..., n. Assume that the design matrix  $\mathbf{X} = (x_1, ..., x_n)^{\top}$  is not random. Also assume that  $\log \mathbb{E}\left[e^{\lambda \eta_i}\right] \leq \sigma^2 \lambda^2 / 2$ . Let

$$\widehat{\boldsymbol{\beta}} = \mathop{\rm argmin}_{\boldsymbol{\beta}} \frac{1}{2n} \| \boldsymbol{Y} - \mathbf{X} \boldsymbol{\beta} \| + \lambda \| \boldsymbol{\beta} \|_1.$$

#### 2.1

Derive the so-called "basic inequality":

$$\|\mathbf{X}\beta^* - \mathbf{X}\widehat{\beta}\| + \lambda \|\beta\|_1 \le \frac{1}{n} \eta^\top \mathbf{X}(\widehat{\beta} - \beta^*) + \lambda \|\beta^*\|_1$$

### 2.2

Let  $X_i$  be the  $j^{th}$  column of **X**. Find a constant M such that  $\mathbb{P}(\mathcal{G}) \geq 1 - 2e^{-\delta}$  for

$$\mathcal{G} = \left\{ \max_{j} |\eta^{\top} X_{j}| / n < M \right\}.$$

# 3 Empirical Rademacher Complexity

Prove the following result. Let H be a set of hypotheses bounded by M > 0 and consider an i.i.d. sample of size n. For any  $h \in H$ , with probability at least  $1 - \delta$ 

$$R_n(h) \le \widehat{R}_n(h) + \widehat{\mathcal{R}}_n(H) + 3M\sqrt{\frac{2\log(2/\delta)}{n}}.$$

You may use the result from class that

$$R_n(h) \le \widehat{R}_n(h) + \mathcal{R}_n(H) + M\sqrt{\frac{2\log(1/\delta)}{n}}.$$