1 The Event Study Method. (Text pp. 571-585)

Do returns go up when a stock splits

Mergers/acquisitions Dividend announcements New debt or equity issues Added or dropped from an index Trump's tweets

In constructing the experiment:

Assume the market is efficient (in the sense that the event is properly assessed by investors and quickly), is unanticipated and exogenous.

Let $r_{t,i}$ be rate of return at t on firm i = 1, ..., N.

$$AR_{t,i} = r_{t,i} - \underbrace{E\left(r_{t,i}|X_{t,i}\right)}_{\text{Normal Return}}$$

- Normal return is fitted value from a regression, AR is the regression residual. Candidate models:
 - Constant mean return model $E(r_{t,i}|X_{t,i}) = \mu_i$. so $AR_{t,i} = r_{t,i} \mu_i$
 - Market model: Run regression

$$r_{t,i} = \alpha_i + \beta_i r_{t,m} + A R_{t,i}$$

• CAPM: run on excess returns

$$r_{t,i}^e = \alpha_i + \beta_i r_{t,m}^e + AR_{t,i}$$

• Can get fancier say with Fama-French 3 factor model. Usually is better to stick with a simpler less parameterized model (so sampling or estimation variability doesn't mess you up).

Goal: test the null hypothesis that abnormal returns during the event window is not significantly different from 0. In what follows, we will ignore the estimation uncertainty in $\hat{\alpha}$ and $\hat{\beta}$.

1. For an individual firm i, Compute sample variance of abnormal return. Sample variance

$$\operatorname{Var}\left(AR_{t,i}\right) = \hat{\sigma}_{ar(T_{1},i)}^{2} = \frac{1}{T_{1} - 1} \sum_{t=1}^{T_{1}} AR_{t,i}^{2}$$

Hence,

$$\operatorname{SAR}_{t,i} = \frac{AR_{t,i}}{\sigma_{ar(T_1,i)}} \sim N(0,1)$$

Can check if $\text{SAR}_{t,i}$ during the event window is significantly different from zero. Say you've estimated the market model. $\hat{\alpha}_i$ and $\hat{\beta}_i$ estimated over L_1 . During L_2 , and L_3 , then use the returns over the entire event sample to construct.

$$AR_{t,i} = r_{t,i} - \hat{\alpha}_i - \hat{\beta}_i r_{t,m}$$

This is for a single firm at a single point in time.

2. Cumulate abnormal returns (CAR) over the event window.

$$CAR_{T_1,T_2,i} = \sum_{t=T_1}^{T_2} AR_{t,i}$$

The variance is

$$\operatorname{var}(CAR_{T_1,T_2,i}) = \operatorname{var}\left(\sum_{t=T_1}^{T_2} AR_{t,i}\right) = (T_2 - T_1 + 1)\operatorname{var}(AR_{t,i})$$

Sample variance

$$\hat{\sigma}_{car(T_1,T_2,i)}^2 = (T_2 - T_1 + 1) \underbrace{\hat{\sigma}_{ar(T_1,i)}^2}_{\text{computed}}$$

$$SCAR_{T_{1},T_{2},i} = \frac{CAR_{T_{1},T_{2},i}}{\hat{\sigma}_{car(T_{1},T_{2},i)}} \sim N(0,1)$$

3. Average across firms to increase sample size.

$$AR_t = \frac{1}{N} \sum_{i=1}^{N} AR_{t,i}$$

We are assuming the AR's are independent across firms. Then

$$\operatorname{Var}\left(AR_{t}\right) = \operatorname{Var}\left(\frac{1}{N}\sum_{i=1}^{N}AR_{t,i}\right) = \frac{1}{N^{2}}\sum_{i=1}^{N}\operatorname{Var}\left(AR_{t,i}\right)$$

Sample variance is

$$\hat{\sigma}_{ar(T_1)}^2 = \frac{1}{N^2} \sum_{i=1}^N \hat{\sigma}_{ar(T_1,i)}^2$$

Now standardize

$$\operatorname{SAR}_{t} = \frac{AR_{t}}{\hat{\sigma}_{ar(T_{1})}} \sim N(0, 1)$$

4. Cumulate these average returns

$$CAR_{T_{1},T_{2}} = \frac{1}{N} \sum_{i=1}^{N} CAR_{T_{1},T_{2},i}$$
$$Var\left(\frac{1}{N} \sum_{i=1}^{N} CAR_{T_{1},T_{2},i}\right) = \frac{1}{N^{2}} \sum_{i=1}^{N} Var\left(CAR_{T_{1},T_{2},i}\right)$$
$$\hat{\sigma}_{car(T_{1},T_{2})}^{2} = \frac{1}{N^{2}} \sum_{i=1}^{N} \hat{\sigma}_{car(T_{1},T_{2},i)}^{2}$$
$$SCAR_{T_{1},T_{2}} = \frac{CAR_{T_{1},T_{2}}}{\hat{\sigma}_{car(T_{1},T_{2})}} \sim N(0,1)$$