

# Empirical Finance: Week 3

## Final Consolidated Exam Study Guide

Economic Evaluation of Asset Returns & OOS Predictability

*Compiled from Lecture Captions & Handout 3*

### CRITICAL EXAM INSTRUCTIONS FROM PROFESSOR

#### **Exam Format:**

- **Pen and paper exam** with numerical exercises you can do by hand
- Multiple choice questions included — especially for Stambaugh bias implications
- “*Tutorials are mostly for the coursework. The slides are mostly for the exam.*”
- “*If you understand the basic concept, then if I give you something slightly different, you should be able to get there.*”

#### **What You MUST Memorise:**

- Henriksson-Merton Market Timing Test formula and Z-statistic
- Bootstrap algorithm steps **in words** (not mathematical details)
- Stambaugh bias **implications** (not the formula itself)
- $R^2_{OOS}$  formula and interpretation

#### **What Will Be Given:**

- Stambaugh bias formula (understand direction of bias only)
- Clark-West  $f_{t+1}$  construction
- Portfolio optimisation closed-form solutions

## 1 Predictive Regressions & Stambaugh Bias (Pages 2–14)

### Professor's Key Teaching Points

#### **Why Predictive Regressions Are Challenging:**

- “*In finance we have these problems... Beta estimates are distorted because we have tons of problems working together*”
- “*X is very persistent [e.g., interest rates with  $\rho \approx 0.99$ ], shocks to Y and X are correlated (often negatively), and we have small samples*”
- “*When you put all of them together, you get these distortions*”

#### **The Three Problems Combined:**

1. Predictors are highly persistent (nearly non-stationary,  $\rho \approx 0.99$ )
2. Returns are noisy with fast mean-reversion
3. Shocks to returns ( $u_t$ ) and predictors ( $v_t$ ) are contemporaneously correlated

#### **Why In-Sample Analysis Fails:**

- “*Bottom line: we don't work with in-sample regressions. In-sample regressions are useless to assess predictability of X on Y*”
- “*We need to combine with an out-of-sample assessment*”

### EXAM INSTRUCTIONS: Stambaugh Bias

*“For the exam you don’t have to remember the formula. However, I can ask a multiple choice question where I say: look, under what conditions do you have a bias? And what is the direction of the bias?”*

*“So in the exam, there could be a multiple choice question where I say: this is the setup, pick the right answer. For example: Does the distortion increase or decrease when T decreases? Is the bias upward or downward if the covariance between u and v is positive or negative?”*

*“You don’t remember the formula, but please try to remember how to interpret these results.”*

#### **Exam Learning Objectives — You Must Be Able To:**

1. Identify when Stambaugh bias occurs (persistent predictor + correlated shocks + small sample)
2. State the direction of bias given the sign of  $\text{Cov}(u_t, v_t)$
3. Explain how bias changes with sample size  $T$  and persistence  $\rho$
4. Justify why out-of-sample analysis is necessary in finance

### Stambaugh Bias Implications — MUST UNDERSTAND

#### **Key relationships (likely MCQ topics):**

- Higher  $\rho$  (more persistent predictor)  $\Rightarrow$  **larger bias**
- Larger  $T$  (more data)  $\Rightarrow$  **smaller bias**
- $\text{Cov}(u_t, v_t) < 0 \Rightarrow$  **upward bias** (typical in finance!)
- $\text{Cov}(u_t, v_t) > 0 \Rightarrow$  **downward bias**

**Intuition:** The minus sign in the formula means negative covariance produces positive (upward) bias.

#### **Formulas Given on Exam:**

Stambaugh (1999) bias approximation:

$$\text{Bias}(\hat{\beta}) \approx -\frac{\text{Cov}(u_t, v_t)}{\text{Var}(v_t)} \times \frac{1+3\rho}{T}$$

## 2 Out-of-Sample Predictability (Pages 15–38)

### Professor's Key Teaching Points

#### Three Key Ingredients for OOS Analysis:

1. **Real-time information:** “I can only use information that was available to an investor at the time”
2. **Benchmark model:** Compare against a reference (typically random walk)
3. **Performance evaluation:** Statistical or economic criteria

#### Why Random Walk as Benchmark:

- “There is plenty of empirical evidence suggesting that it’s very hard to beat the random walk”
- “Sometimes you can beat it, but there is no paper showing you can systematically beat the random walk in all states”

#### Macro Data Challenges:

- “Inflation for December 2025 is not available in December 2025 — it’s made available six weeks after”
- “Macro data are often revised, so you don’t know if the data today will be the same six months after”

#### *Exam Learning Objectives — You Must Be Able To:*

1. Distinguish between **expanding window** (keeps all past data) and **rolling window** (fixed sample, drops oldest)
2. Set up a proper out-of-sample forecasting exercise
3. Explain real-time information constraints
4. Apply Campbell & Thompson (2008) sign restrictions

#### *Formulas Given on Exam:*

##### OLS Estimators:

$$\hat{\beta} = \frac{\text{Cov}(y_t, x_{t-1})}{\text{Var}(x_{t-1})}, \quad \hat{\alpha} = \bar{y} - \hat{\beta}\bar{x}$$

##### One-step ahead forecast:

$$\hat{y}_{t+1|t} = \hat{\alpha}_t + \hat{\beta}_t x_t$$

##### Forecast error:

$$\hat{\varepsilon}_{t+1} = y_{t+1} - \hat{y}_{t+1|t}$$

### 3 Statistical Evaluation (Pages 39–53)

#### Professor's Key Teaching Points

##### Interpreting $R^2_{OOS}$ :

- “1% is wow. Sometimes you have less than 1%”
- “If someone shows me an out-of-sample R-squared of 2-3%, it’s amazing — you should bet everything on that model”
- “When I mark coursework and find R-squared of 20-30%, clearly something is wrong”
- “You need to have a sense of the data”

##### Why Simple MSE Comparison Fails (Clark-West):

- “The null hypothesis is incorrectly formulated”
- “Under the null that the model and benchmark are equally bad, the model has an extra parameter”
- “Adding an extra parameter means estimation noise”
- “So under the null of equal predictability, the MSE of the model is always higher than the MSE of the benchmark”
- “Your out-of-sample R-squared could be negative, but maybe your model is still better than the benchmark”

#### Interpreting OOS R-squared Values

$R^2_{OOS}$	Value	Interpretation
< 0		Model performs <b>worse</b> than benchmark
0.1% – 0.5%		Typical, potentially useful
0.5% – 1%		Good performance
1% – 2%		“Wow” — excellent
2% – 3%		“Amazing” — bet everything
> 10%		Something is wrong! Check your code

#### Exam Learning Objectives — You Must Be Able To:

1. Calculate and interpret MSE, MAE, and  $R^2_{OOS}$
2. Explain why  $R^2_{OOS}$  can be negative
3. Understand why Clark-West test is needed (extra parameters add noise under null)
4. Apply the Henriksson-Merton market timing test (see below)

#### $R^2_{OOS}$ Formula — MUST MEMORISE

$$R^2_{OOS} = 1 - \frac{MSE_{MOD}}{MSE_{BEN}}$$

#### Interpretation:

- $R^2_{OOS} > 0$ : Model beats benchmark (lower MSE)
- $R^2_{OOS} < 0$ : Benchmark beats model
- $R^2_{OOS} = 0$ : Equal performance

*Formulas Given on Exam:*

**Mean Squared Error:**

$$MSE = \frac{1}{T-t} \sum_{i=t+1}^T \hat{\varepsilon}_i^2$$

**Mean Absolute Error:**

$$MAE = \frac{1}{T-t} \sum_{i=t+1}^T |\hat{\varepsilon}_i|$$

**Clark-West statistic:**

$$f_{t+1} = \hat{\varepsilon}_{t+1,BEN}^2 - [\hat{\varepsilon}_{t+1,MOD}^2 - (\hat{y}_{t+1,BEN} - \hat{y}_{t+1,MOD})^2]$$

Reregress  $f_i$  on a constant, compute t-statistic with Newey-West standard errors.

#### 4 Henriksson-Merton Market Timing Test (Pages 48–49)

##### CRITICAL: Professor's Exam Instructions

*“For the exam, I might give you a small table with 5 or 6 data points, and I will ask you to calculate the test.”*

*“This formula must be remembered. If I tell you to remember something, it means you have to remember it.”*

*“This one could be in the exam. In the exam, I can ask you exactly to apply the formula.”*

##### Professor's Key Teaching Points

###### What Market Timing Measures:

- “You want to make sure that the model gives you the right side of the market”
- “You don’t care about the magnitude. You care about the sign”
- “If you flip a coin, you are right 50% of the time, wrong 50% of the time — that’s not skill, just luck”

###### What's a Good $p$ Value:

- “If your model is good, you are slightly better than 50%”
- “51-52% — wow, that’s very good”
- “Most of the times it’s just a little bit above 50%”

**Exam Learning Objectives — You Must Be Able To:**

1. Calculate  $p = c/n$  from a table of forecasts and actual returns
2. Compute the Z-statistic
3. Compare to critical value **1.65** and state conclusion
4. Interpret market timing ability

## Henriksson-Merton Formulas — MUST MEMORISE

**Directional accuracy:**

$$p = \frac{c}{n}$$

where  $c$  = number of correct forecasts (same sign as actual return),  $n$  = total forecasts.

**Hypotheses:**

- $H_0 : p = 0.5$  (no market timing ability — just luck)
- $H_A : p > 0.5$  (positive market timing ability)

**Test statistic** (for  $n > 30$ , binomial  $\approx$  Normal):

$$Z = \frac{p - 0.5}{\sqrt{\frac{p(1-p)}{n}}}$$

**Decision rule:** If  $Z > 1.65$ , reject  $H_0$  at 5% level (one-sided test).

## 5 Bootstrap Algorithm (Pages 50–53)

### EXAM INSTRUCTIONS: Bootstrap

*“I might ask a question about the bootstrap in the exam. However, I’m not asking you the maths. I’m asking you to explain in words what the steps are.”*

*“You don’t need to remember the details, but you have to remember the general idea.”*

*“When I say describe the bootstrap algorithm, it means you don’t remember the formula but in words you can describe how you run the recipe.”*

### Professor’s Key Teaching Points

**Why Bootstrap vs Monte Carlo:**

- “In bootstrap, you don’t randomly draw from a distribution — you resample from the actual errors”
- “This preserves the linear dependence that exists in the data”
- “Monte Carlo samples from a parametric distribution, which may miss higher-moment dependencies”

**Key Technical Points:**

- “You resample from a uniform between 1 and 100 [or  $n$ ], then use those draws to pick rows”
- “You have a matrix with two columns —  $u_1$  and  $u_2$  — you sample both at the same time”
- “How many errors do you sample? Exactly the same number as the original data”

### Exam Learning Objectives — You Must Be Able To:

1. Describe the bootstrap algorithm in words (6 steps below)
2. Explain why bootstrap preserves dependence structure
3. Calculate an empirical p-value

### Bootstrap Algorithm Steps — DESCRIBE IN WORDS

1. **Compute test statistic on actual data:** Calculate  $R^2_{OOS}$ , CW, or other statistic. Save this number  $\hat{\tau}$ .
2. **Estimate model under null of no predictability:**
  - $y_t = \alpha + u_{1,t}$  (returns are unpredictable)
  - $x_t = c + \rho_1 x_{t-1} + \dots + \rho_p x_{t-p} + u_{2,t}$  (predictor follows AR process)
  - Extract residuals  $\hat{u}_t = (\hat{u}_{1,t}, \hat{u}_{2,t})'$
3. **Generate synthetic data by resampling residuals:**
  - Draw indices from uniform distribution (with replacement)
  - Resample residual pairs (preserve correlation!)
  - Construct synthetic  $y^*$  and  $x^*$  using estimated parameters
4. **Compute test statistic on simulated data:** Run the same OOS procedure on synthetic data, get  $\hat{\tau}^*$ .
5. **Repeat steps 3–4 many times:**  $B = 1000$  or more iterations.
6. **Calculate p-value:**

$$\text{p-value} = \frac{1}{B} \sum_{j=1}^B \mathbf{1}(\hat{\tau}_j^* > \hat{\tau})$$

## 6 Economic Evaluation (Pages 54–65)

### Professor's Key Teaching Points

#### Why Statistical Measures Aren't Enough:

- “There’s a paper by Leitch and Tanner (1992) showing that statistical measures like MSE have no consistent relationship with profits”
- “You can have low MSE but low profit, or high MSE but high profit”
- “That’s why we complement statistical criteria with economic criteria”

#### Sharpe vs Sortino Ratio:

- “The Sharpe ratio only cares about mean and variance”
- “But market participants are very much concerned about downside risk — negative skewness”
- “If things go wrong, you go bankrupt”
- “That’s why we also calculate the Sortino ratio, which uses only downside volatility”

#### Sortino Calculation:

- “You only take the subset of negative portfolio returns”
- “Calculate the standard deviation on those negative returns”
- “The numerator is the same as Sharpe; only the denominator changes”

### Exam Learning Objectives — You Must Be Able To:

1. Explain why economic evaluation complements statistical evaluation
2. Define and interpret Sharpe ratio, Sortino ratio, Certainty Equivalent Return
3. Understand mean-variance optimisation setup
4. Explain the break-even transaction cost concept

*Formulas Given on Exam:*

**Sharpe Ratio:**

$$SR = \frac{E(r_{p,t+1} - r_{f,t})}{\sqrt{\text{Var}(r_{p,t+1} - r_{f,t})}}$$

**Sortino Ratio:**

$$SO = \frac{E(r_{p,t+1} - r_{f,t})}{\sqrt{\text{Var}(r_{p,t+1} - r_{f,t} \mid r_{p,t+1} - r_{f,t} < 0)}}$$

## 7 Transaction Costs & Break-Even (Pages 63–65)

### Professor's Key Teaching Points

**The Problem:**

- “You can have a great strategy, but then this strategy rebalances your portfolio so often, so quickly”
- “If you rebalance often and have large swings in your portfolio weights, you’re going to pay more transaction costs”

**Why Break-Even Approach:**

- “You don’t know what transaction costs you’ll actually pay — it depends on how much you trade”
- “If you’re a small fund you pay one cost; if you’re Blackrock you pay another cost; if you’re a pension fund you pay another cost”
- “So we calculate the break-even transaction cost: the cost that would kill the excess performance of your model relative to the benchmark”
- “If you pay less than break-even, there’s still value in your model”

*Exam Learning Objectives — You Must Be Able To:*

1. Explain why transaction costs matter for backtesting
2. Define break-even transaction cost conceptually
3. Interpret: if actual cost < break-even cost  $\Rightarrow$  model still has value

## 8 Model Combination (Pages 66–72)

### Professor's Key Teaching Points

**Why Combine Forecasts:**

- “We’ve evaluated individual models relative to the benchmark, but ex-ante we don’t know which model is true”
- “This generates model uncertainty”
- “The superior performance of combined forecasts is known since Bates and Granger (1969)”

*Exam Learning Objectives — You Must Be Able To:*

1. Explain why model combination is used (model uncertainty)
2. Describe the three types of forecast combination
3. Write the combined forecast formula

## Model Combination — UNDERSTAND CONCEPTS

**Combined forecast:**

$$\tilde{y}_{t+1} = \sum_{m=1}^M \kappa_t^m \hat{y}_{t+1}^m, \quad \text{where } \sum_{m=1}^M \kappa_t^m = 1$$

**Three Weighting Schemes:**

1. **Simple Averaging:**

- Mean rule:  $\kappa_t^m = 1/M$  (equal weights)
- Median rule: Select median forecast

2. **Statistical Averaging** (inverse MSE weighting):

$$\kappa_t^m = \frac{1/MSE_t^m}{\sum_{j=1}^M 1/MSE_t^j}$$

Lower MSE  $\Rightarrow$  higher weight

3. **Economic Averaging** (e.g., Sharpe ratio weighting):

$$\kappa_t^m = \frac{SR_t^m}{\sum_{j=1}^M SR_t^j}$$

Higher Sharpe ratio  $\Rightarrow$  higher weight

## 9 Final Exam Checklist

### Formulas You MUST Memorise

1. **Henriksson-Merton Market Timing Test:**

- $p = c/n$  (directional accuracy)
- $Z = \frac{p-0.5}{\sqrt{p(1-p)/n}}$
- Reject  $H_0$  if  $Z > 1.65$  (5% one-sided)

2. **Out-of-Sample R-squared:**

$$R_{OOS}^2 = 1 - \frac{MSE_{MOD}}{MSE_{BEN}}$$

3. **Bootstrap Algorithm** (in words):

1. Compute statistic on actual data
2. Estimate model under null (no predictability)
3. Resample residual pairs with replacement
4. Recompute statistic on synthetic data
5. Repeat 1000+ times
6. p-value = fraction of simulated statistics  $>$  actual

### Conceptual Understanding Required (Likely MCQ)

#### Stambaugh Bias:

- Higher  $\rho \Rightarrow$  larger bias
- Larger  $T \Rightarrow$  smaller bias
- $\text{Cov}(u, v) < 0 \Rightarrow$  **upward** bias
- $\text{Cov}(u, v) > 0 \Rightarrow$  **downward** bias

#### Clark-West Test:

- Why simple null is “incorrectly formulated”
- Extra parameters add estimation noise under null
- $R^2_{OOS}$  can be negative but model still better

#### $R^2_{OOS}$ Interpretation:

- 1% = “wow”, 2-3% = “amazing”, 10%+ = suspicious

#### Economic Evaluation:

- Sharpe ratio: total volatility
- Sortino ratio: downside volatility only
- Break-even cost: cost that kills model’s advantage

#### Model Combination:

- Simple (equal weights), Statistical (inverse MSE), Economic (Sharpe)

### Professor’s Final Warnings

- *“If I tell you to remember something, it means you have to remember it”* — H-M test!
- *“You need to have a sense of the data”* — know typical  $R^2_{OOS}$  values
- *“You’re not just a scientist, you’re also an artist”* — understand the data
- Bootstrap: describe in **words**, not formulas
- Stambaugh: understand **implications**, not formula