

# *R Analytics for Spread Trading: Some Interesting Bits*

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# *What the heck is a “spread”?*

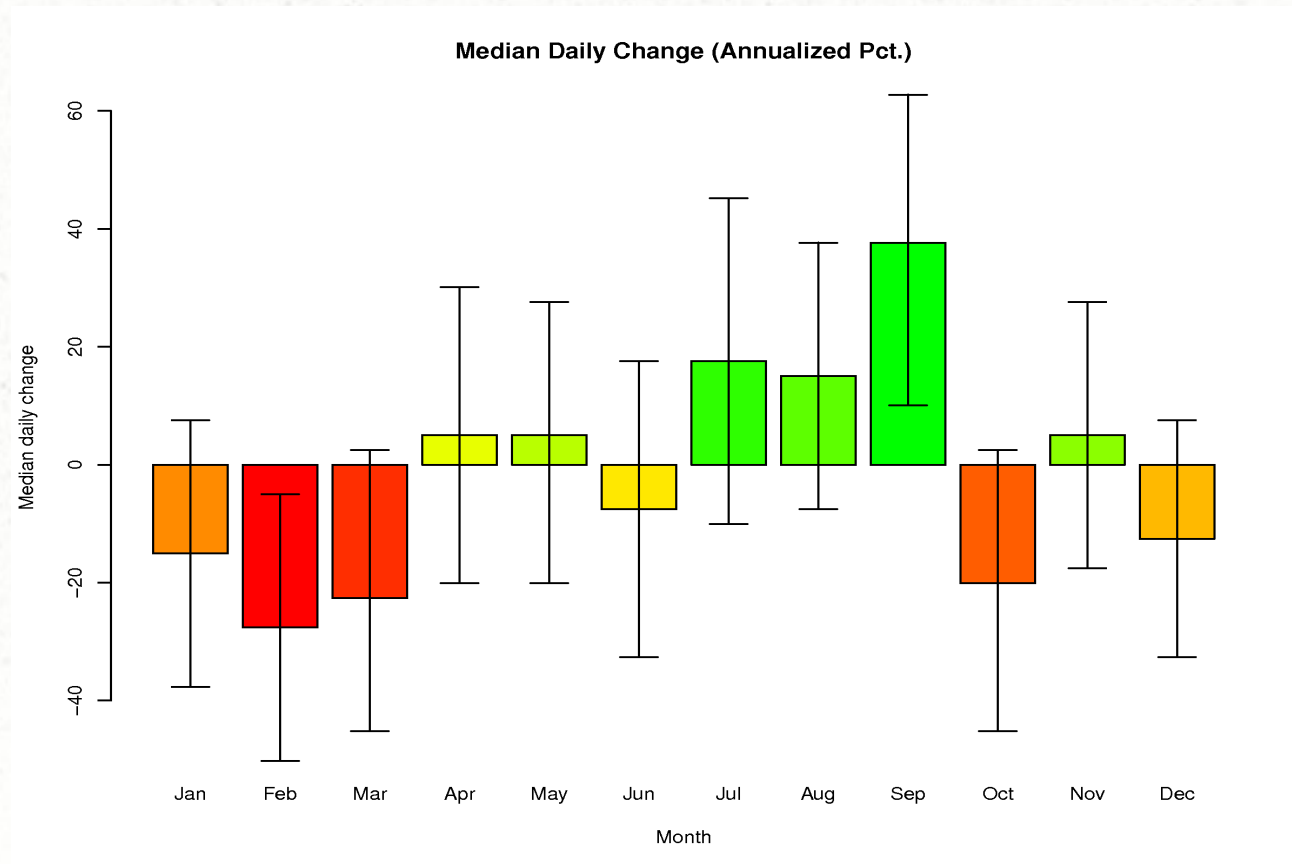
- A spread is the difference between two prices.
  - Crude oil vs. gasoline
  - Corn vs. soybeans
  - Treasury bonds vs. Treasury notes
- Some spreads have predictable properties.
  - Mean reversion: Tend back to their avg.
  - Seasonality: Price patterns by season

# *Software for analyzing spreads*

- Apache web server on an Ubuntu box
- Perl middleware (5,000 lines)
- **R for analytics** (7,000 lines) using off-the-shelf packages and a small local library
- MySQL back-end (2,800 lines)
- Batch jobs, nightly and weekly
- Accessed entirely through browser



# *Finding Seasonal Patterns: ANOVA & Bar Charts*



# *Bar Charts: Confidence Intervals and Colors*

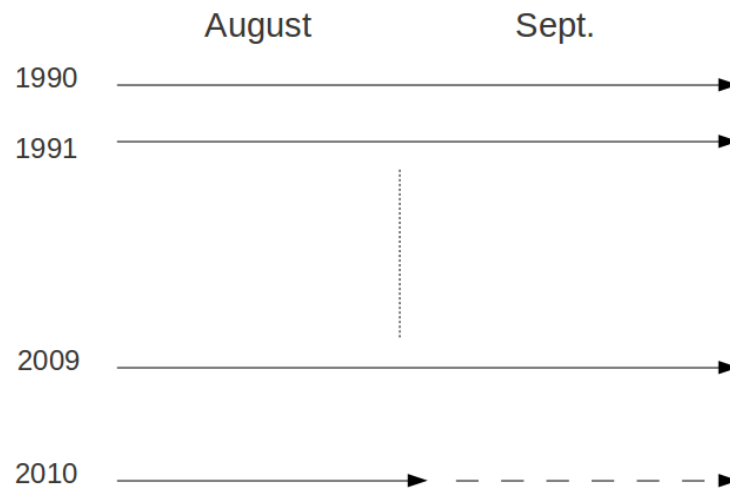
```
library(gplots)                # For barplot2 function

# . . . compute medians (bar heights) and confidence
# intervals using wilcox.test function . . .

# Build a vector of colors from red to green
cols <- rainbow(heights,start=0,end=2/6)[rank(heights)]

# Plot heights with confidence intervals and colors
barplot2(heights,
  plot.ci = TRUE, ci.u = ci.upper, ci.l = ci.lower,
  col=cols,
  . . . )
```

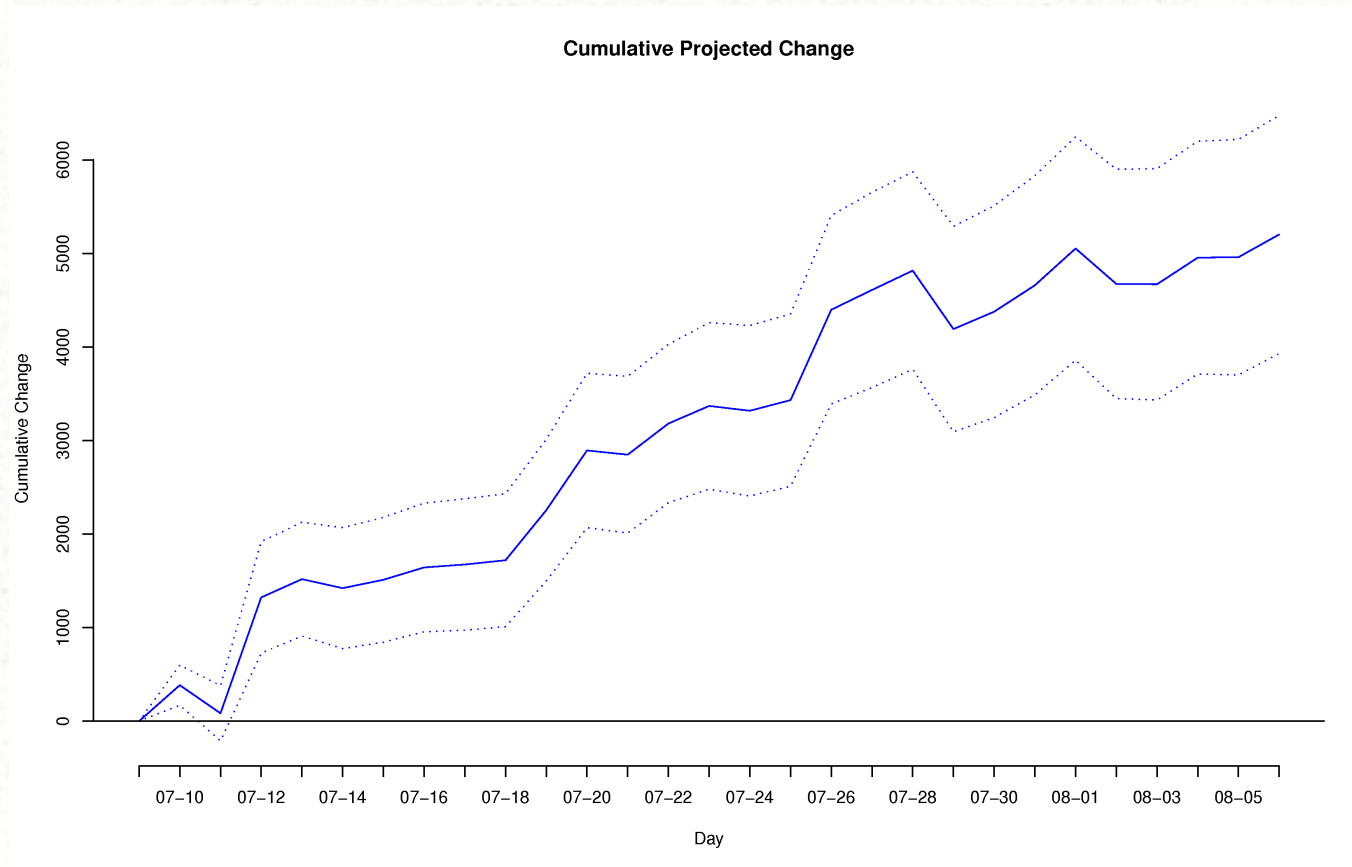
# Predicting a Seasonal Spread



- Want to project daily changes for Sept.
- Build a linear regression model of recent August changes using data from previous Augusts.
- Use backwards stepwise regr. to eliminate uninformative history.
- With that model and data from previous Septs, project spread changes for this Sept.



# Typical Seasonal Prediction



# *Outline of R Code*

```
# Build the model
#
m <- lm(recent ~ ., data=augustHistory)

# Eliminate unhelpful years
#
m.red <- step(m, direction="backward")

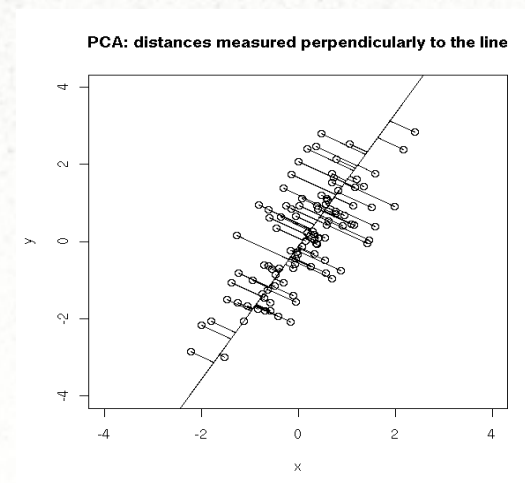
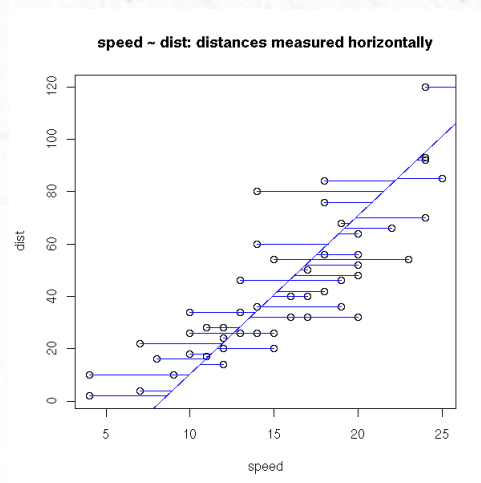
# Project into September
#
pred <- predict(m.red, newdata=septHistory)
```



# *Hedge Ratio: Finding the ratio of two random variables*

- Spread calculated as  $S_i = Y_i - \beta \times X_i + \varepsilon_i$
- $\beta$  is needed ratio between  $Y$  and  $X$ . How calculate?
- Tried extracting  $X$  coefficient from  $\text{lm}(y \sim x)$
- Doesn't work well: *Ord. least squares assumes  $Y$  random,  $X$  not random (zero variance).*
- Result is that ratio for  $X/Y$  is not simply inverse of ratio for  $Y/X$ . (Oops!)

# Ord. Least Squares vs. Total Least Squares



# *R Code for Total Least Squares (a.k.a. Orthogonal Least Squares)*

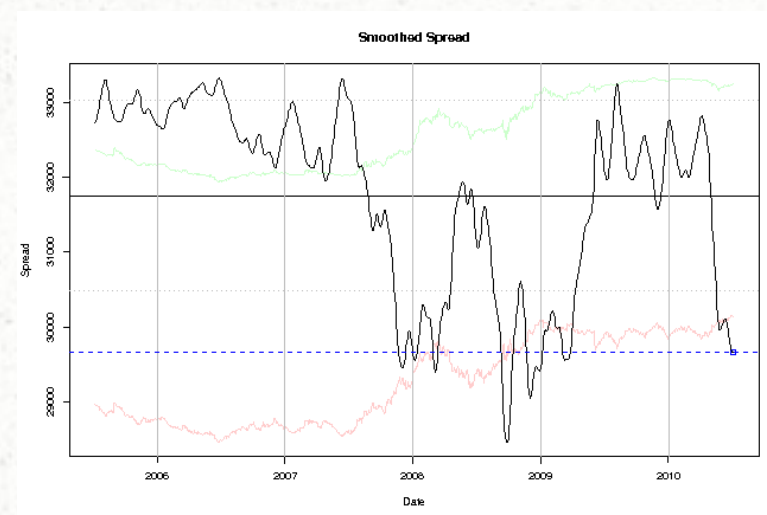
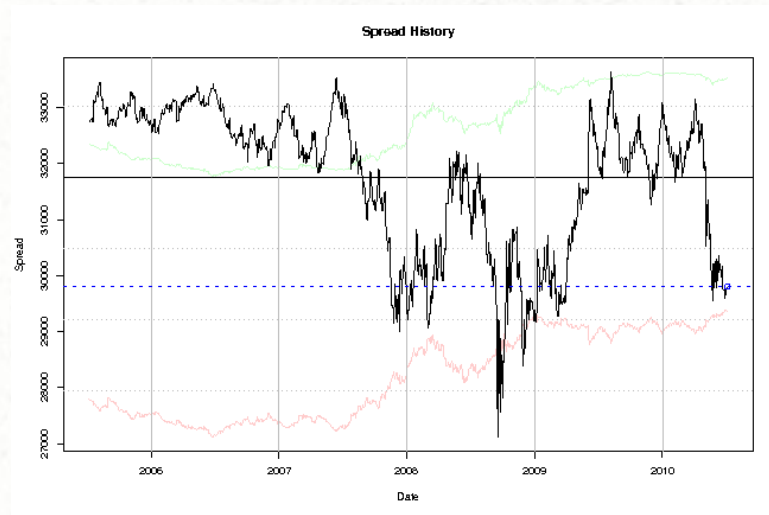
```
# Do principle components analysis (PCA)
#
r <- princomp( ~ x + y)

# Slope of orthogonal regression
#
slope <- r$loadings[2,1] / r$loadings[1,1]

# Intercept of orthogonal regression
#
inter <- r$center[2] - slope * r$center[1]
```



# Removing Noise



- “Noisy” data complicates decisions
- Smoothing reveals the underlying pattern

# *Local Polynomial Smoothing*

## *via KernSmooth Package*

```
library(KernSmooth)

# . . . given t (time) and y . . .

N <- length(t)

# Calculate bandwidth, the degree of smoothing
# (dpill = "Direct Plug-In Meth. for Local Lin. Regr.")
#
bw <- dpill(t, y, gridsize=N)

# Perform smoothing
#
smooth <- locpoly(x=t, y=y, bandwidth=bw, gridsize=N)
```

# *You, too, can build trading analytics using simple tools found around your house!*

- Built mostly from off-the-shelf R packages
- Consistent application of developed statistical tools to financial data.
- Not rocket science. *It just works.*



# *R Analytics for Spread Trading*

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