

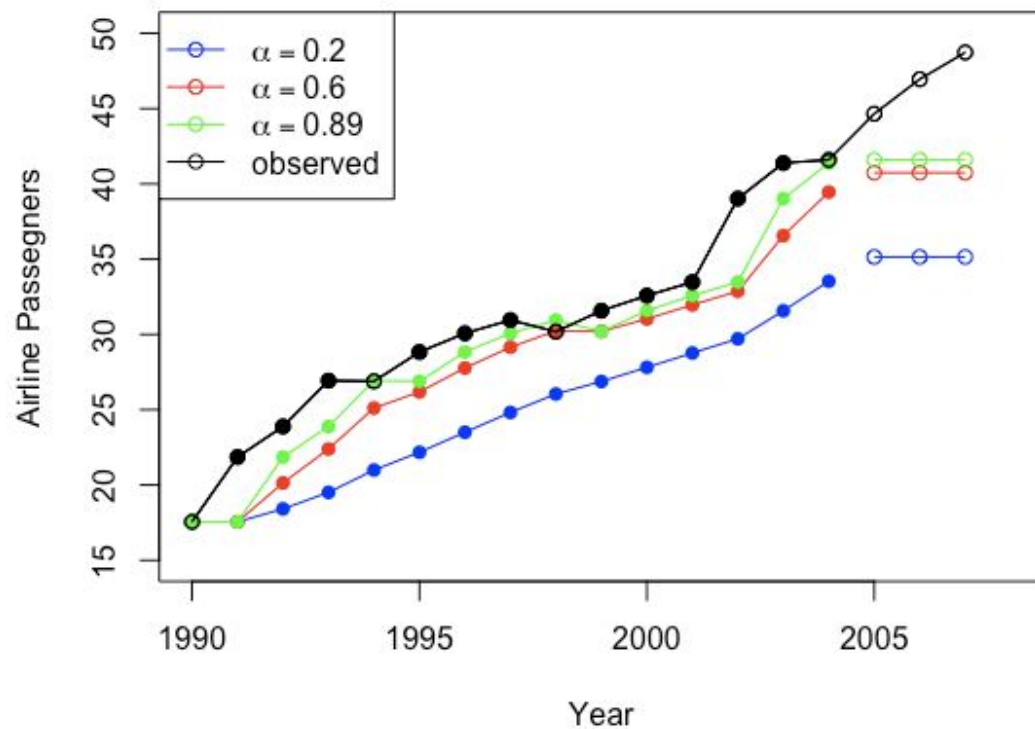
Unit11: For Live Session

DS6306

Garrity

PART 1 - Fit model

Simple Exponential Smooth



PART 2 - Holt Method

Fit both models using ETS function

```
plot(air, ylab = "Airline Passengers", xlab = "Year", type = "o", pch=16, xlim = c(1990, 2009), ylim = c(15, 60), main="ETS method from fpp2")
fit5e <- ets(air, model="ZZZ", damped=TRUE, alpha=0.8, beta = 0.2)
forecast5e <- forecast(fit5e, h=5)
lines(fitted(fit5e), col="purple", type="o", pch=16)
lines(forecast5e$mean, col="purple", type="o")

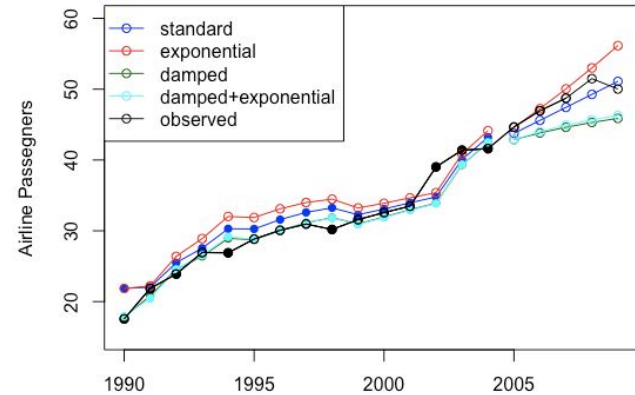
fit6e <- ets(air, model="ZZZ", damped=FALSE, alpha=0.8, beta = 0.2)
forecast6e <- forecast(fit6e, h=5)
lines(fitted(fit6e), col="red", type="o", pch=16)
lines(forecast6e$mean, col="red", type="o")

legend("topleft", lty=1, col=c("purple", "red"),
      c("damped", "not damped"), pch=1)

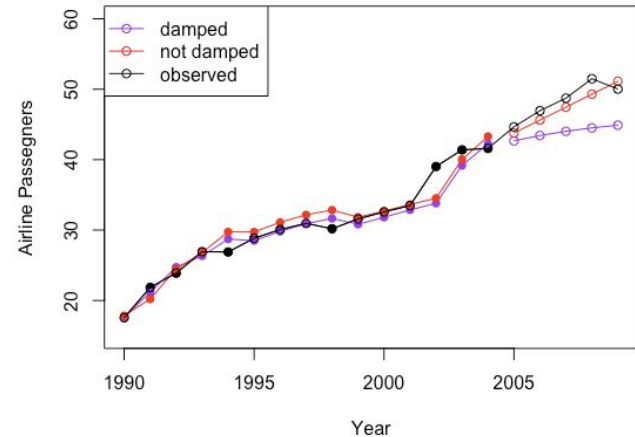
> c(fit5e$aic, fit6e$aic)
[1] 60.35480 61.99738
> c(fit5e$bic, fit6e$bic)
[1] 63.18700 64.12153
```

Model fit5e has the lowest AIC and BIC

Using Holt method from fpp2 package



ETS method from fpp2



PART 2 - Compare Holt and ETS accuracies

Holt:

```
> accuracy(fitlh, ausair)
```

	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1	Theil's U
Training set	-1.0292272	2.202869	1.772637	-4.485612	6.364749	0.9671310	0.2088940	NA
Test set	0.9405879	1.444776	1.371076	1.947628	2.808140	0.7480438	-0.3313068	0.7212626

ETS:

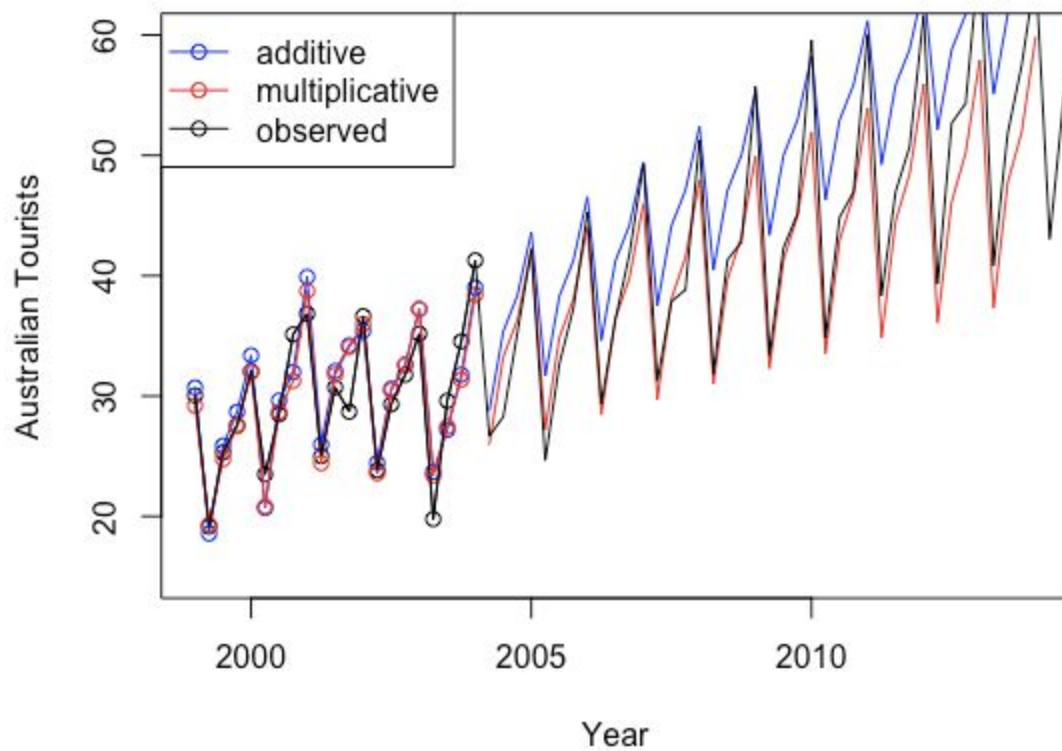
```
> accuracy(forecast6e, ausair)
```

	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1	Theil's U
Training set	-0.2713962	1.765313	1.289077	-1.112766	4.155394	0.7033060	0.2021848	NA
Test set	0.9154097	1.429896	1.358517	1.895872	2.781609	0.7411917	-0.3300821	0.7138642

ETS had slightly better performance metrics, relative to the Holt function, in the test set.

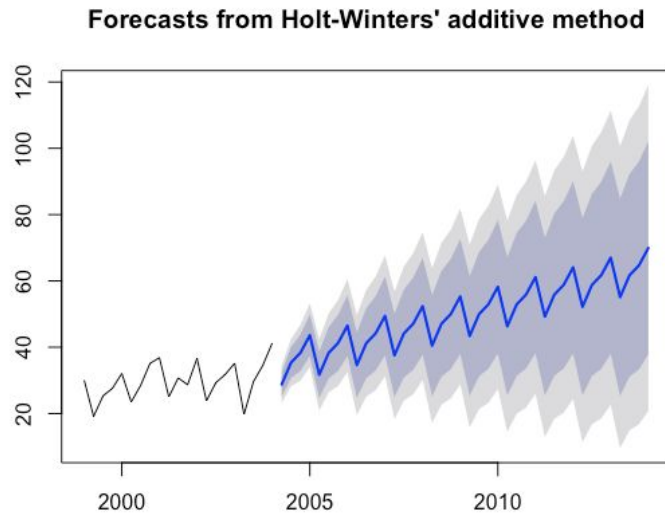
PART 3 - Holt-Winters, Seasonality

Holt-Winters

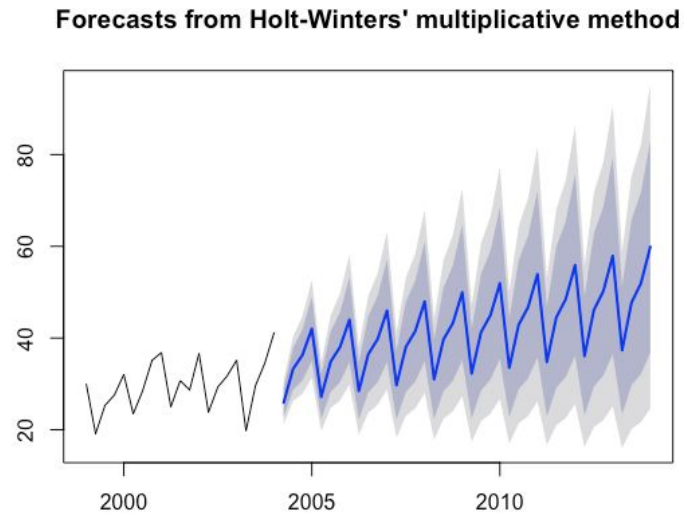


PART 3 - Holt-Winters, Cool Figures!

```
plot(fit1s)
```



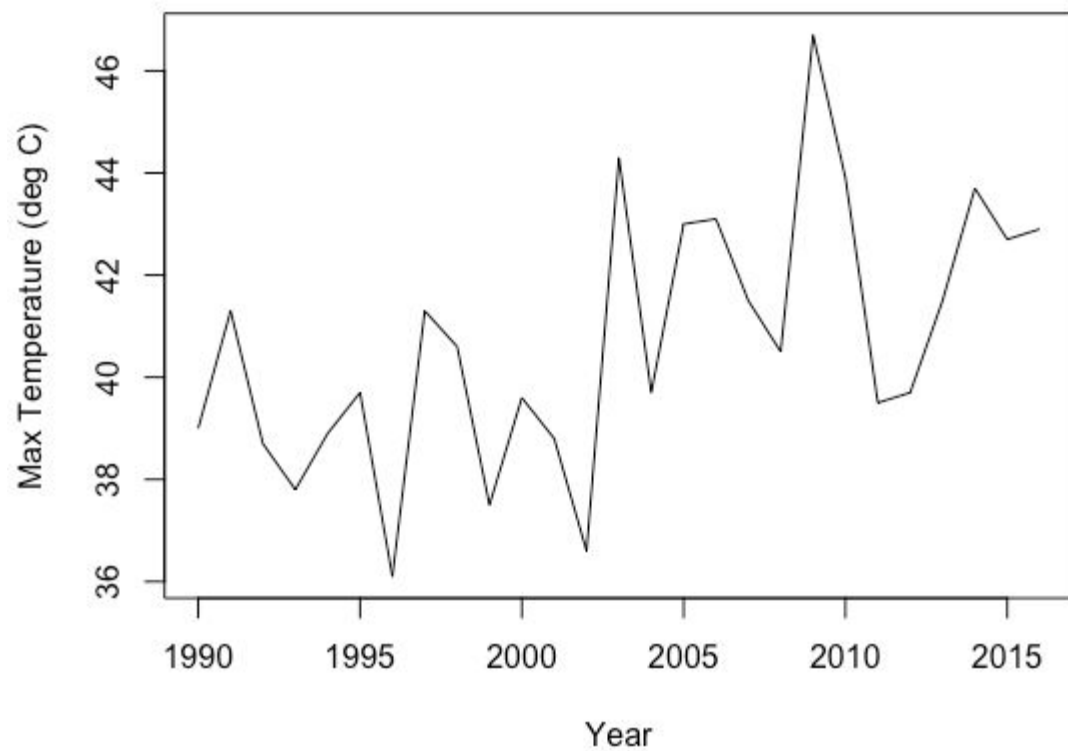
```
plot(fit2s)
```



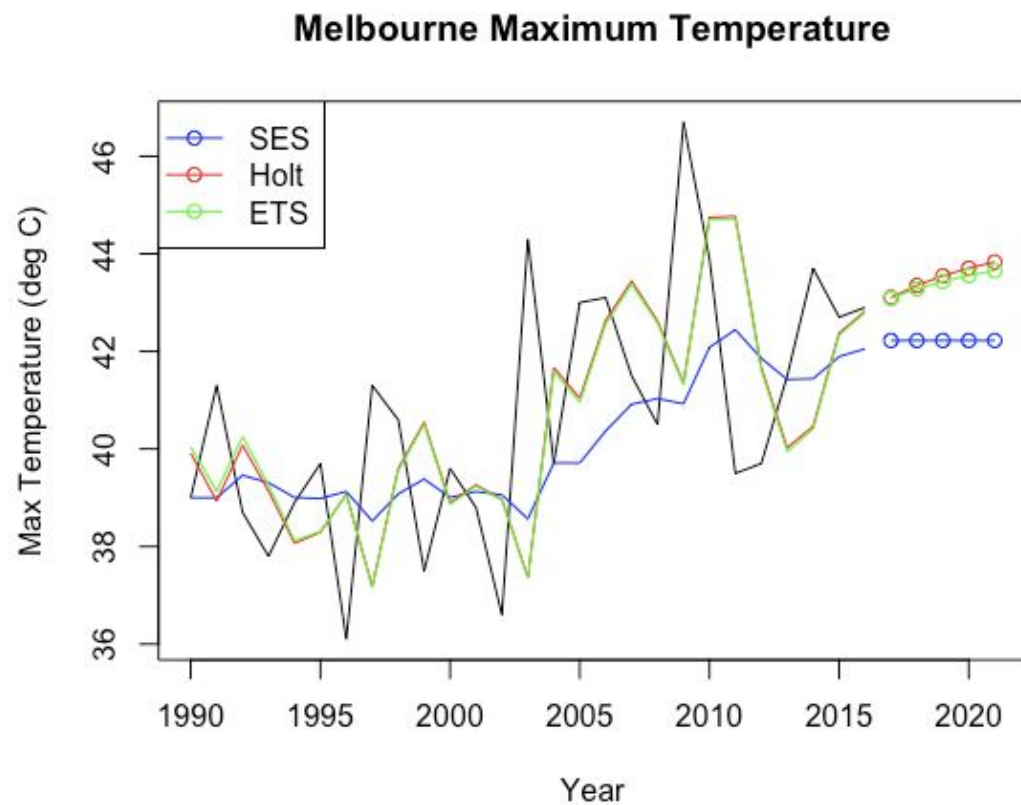
Interesting to see the differences in confidence and prediction intervals.

PART 4 - Plot the Data

Melbourne Maximum Temperature



PART 4 - Plot the Data



PART 4 - Evaluate Models

It wasn't clear to me how to get AIC, AICc, BIC from SES and Holt models. Here it is for the ETS model:

```
# ETS AIC, AICc, BIC:  
> print(c(fit3$aic, fit3$aicc, fit3$bic))  
[1] 150.1276 151.9458 155.3109
```

SES:

```
> accuracy(fit1$fitted, maxtemp)
```

	ME	RMSE	MAE	MPE	MAPE	ACF1	Theil's U
Test set	0.5967905	2.326389	1.760939	1.18444	4.253067	-0.09325844	0.7727319

Holt:

```
> accuracy(fit2$fitted, maxtemp)
```

	ME	RMSE	MAE	MPE	MAPE	ACF1	Theil's U
Test set	0.1354753	2.675408	2.105991	0.06335922	5.143567	-0.2528753	0.8876103

ETS:

```
> accuracy(forecast3$fitted, maxtemp)
```

	ME	RMSE	MAE	MPE	MAPE	ACF1	Theil's U
Test set	0.1453593	2.675333	2.114126	0.08420999	5.163047	-0.2502201	0.887498

Takeaways & Questions

This assignment is very late. I don't expect any credit. Just tying up loose ends.

I had some issues getting all the functions that Dr. McGee was using to work when I tried to replicate her code. The screens in the video were hard to read, so I don't know if I was missing something or if there was an issue with difference between versions of the same package?

I guess that is a takeaway...things change, packages get deprecated or updated, Python moves from 2 to 3. We can't expect things to stay static so we need to learn to be self-reliant.

Time series (`'ts'`) objects are pretty cool in R. It took me a minute to understand the start/end/frequency syntax, but once I did it all clicked. Managing the time component of time series can be a real pain in the ass. It looks like `ts()` help make it a little less painful.

It's looking like a lot of these models have built in plots that can be accessed by calling `plot(fitted_model)`.