```
In [1]:
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In [2]:
!pip install tf-nightly-2.0-preview
ERROR: Could not find a version that satisfies the requirement tf-nightly-2.0-preview (from versio
ns: none)
ERROR: No matching distribution found for tf-nightly-2.0-preview
In [3]:
import tensorflow as tf
import numpy as np
import matplotlib.pyplot as plt
print(tf.__version__)
2.3.0
In [4]:
def plot series(time, series, format="-", start=0, end=None):
    plt.plot(time[start:end], series[start:end], format)
    plt.xlabel("Time")
    plt.ylabel("Value")
    plt.grid(False)
def trend(time, slope=0):
   return slope * time
def seasonal pattern(season time):
    """Just an arbitrary pattern, you can change it if you wish"""
    return np.where(season time < 0.1,</pre>
                    np.cos(season time * 6 * np.pi),
                    2 / np.exp(9 * season time))
def seasonality(time, period, amplitude=1, phase=0):
    """Repeats the same pattern at each period"""
    season_time = ((time + phase) % period) / period
    return amplitude * seasonal_pattern(season_time)
def noise(time, noise level=1, seed=None):
    rnd = np.random.RandomState(seed)
    return rnd.randn(len(time)) * noise level
time = np.arange(10 * 365 + 1, dtype="float32")
baseline = 10
series = trend(time, 0.1)
baseline = 10
amplitude = 40
slope = 0.005
noise level = 3
# Create the series
series = baseline + trend(time, slope) + seasonality(time, period=365, amplitude=amplitude)
# Update with noise
```

series += noise(time, noise level, seed=51)

```
split_time = 3000
time_train = time[:split_time]
x_train = series[:split_time]
time_valid = time[split_time:]
x_valid = series[split_time:]
window_size = 20
batch_size = 32
shuffle_buffer_size = 1000
plot_series(time, series)
```

```
70 60 50 40 20 20 2500 3000 3500 Time
```

In [5]:

```
def windowed_dataset(series, window_size, batch_size, shuffle_buffer):
    dataset = tf.data.Dataset.from_tensor_slices(series)
    dataset = dataset.window(window_size + 1, shift=1, drop_remainder=True)
    dataset = dataset.flat_map(lambda window: window.batch(window_size + 1))
    dataset = dataset.shuffle(shuffle_buffer).map(lambda window: (window[:-1], window[-1]))
    dataset = dataset.batch(batch_size).prefetch(1)
    return dataset
```

In [6]:

```
dataset = windowed_dataset(x_train, window_size, batch_size, shuffle_buffer_size)

model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(100, input_shape=[window_size], activation="relu"),
    tf.keras.layers.Dense(10, activation="relu"),
    tf.keras.layers.Dense(1)
])

model.compile(loss="mse", optimizer=tf.keras.optimizers.SGD(lr=le-6, momentum=0.9))
model.fit(dataset,epochs=100,verbose=0)
```

Out[6]:

<tensorflow.python.keras.callbacks.History at 0x7f55104f7630>

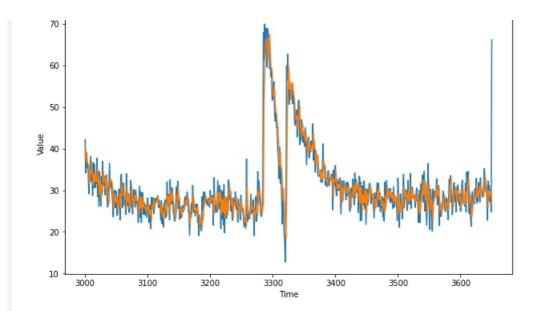
In [7]:

```
forecast = []
for time in range(len(series) - window_size):
    forecast.append(model.predict(series[time:time + window_size][np.newaxis]))

forecast = forecast[split_time-window_size:]
results = np.array(forecast)[:, 0, 0]

plt.figure(figsize=(10, 6))

plot_series(time_valid, x_valid)
plot_series(time_valid, results)
```



In [8]:

 ${\tt tf.keras.metrics.mean_absolute_error(x_valid, results).numpy()}$

Out[8]:

3.1644258